Chatgpt Documentation Latest

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Introduction

Looking for ChatGPT? Head to chat.openai.com.

The OpenAl API can be applied to virtually any task. We offer a range of <u>models</u> with different capabilities and price points, as well as the ability to <u>fine-tune</u> custom models.

Resources

- Experiment in the playground
- Read the API reference
- Visit the <u>help center</u>
- View the current API <u>status</u>
- Check out the OpenAl Developer Forum
- Learn about our usage policies

At OpenAI, protecting user data is fundamental to our mission. We do not train our models on inputs and outputs through our API. Learn more on our <u>API data privacy page</u>.

Key concepts

Text generation models

OpenAl's text generation models (often referred to as generative pre-trained transformers or "GPT" models for short), like GPT-4 and GPT-3.5, have been trained to understand natural and formal language. Models like GPT-4 allows text outputs in response to their inputs. The inputs to these models are also referred to as "prompts". Designing a prompt is essentially how you "program" a model like GPT-4, usually by providing instructions or some examples of how to successfully complete a task. Models like GPT-4 can be used across a great variety of tasks including content or code generation, summarization, conversation, creative writing, and more. Read more in our introductory text generation guide and in our prompt engineering guide.

Assistants

Assistants refer to entities, which in the case of the OpenAl API are powered by large language models like GPT-4, that are capable of performing tasks for users. These assistants operate based on the instructions embedded within the context window of the model. They also usually have access to tools which allows the assistants to perform more complex tasks like running code or retrieving information from a file. Read more about assistants in our Assistants API Overview.

Embeddings

An embedding is a vector representation of a piece of data (e.g. some text) that is meant to preserve aspects of its content and/or its meaning. Chunks of data that are similar in some way will tend to have embeddings that are closer together than unrelated data. OpenAI offers text embedding models that take as input a text string and produce as output an embedding vector. Embeddings are useful for search, clustering, recommendations, anomaly detection, classification, and more. Read more

about embeddings in our embeddings guide.

Tokens

Text generation and embeddings models process text in chunks called tokens. Tokens represent commonly occurring sequences of characters. For example, the string "tokenization" is decomposed as "token" and "ization", while a short and common word like "the" is represented as a single token. Note that in a sentence, the first token of each word typically starts with a space character. Check out our tokenizer tool to test specific strings and see how they are translated into tokens. As a rough rule of thumb, 1 token is approximately 4 characters or 0.75 words for English text.

One limitation to keep in mind is that for a text generation model the prompt and the generated output combined must be no more than the model's maximum context length. For embeddings models (which do not output tokens), the input must be shorter than the model's maximum context length. The maximum context lengths for each text generation and embeddings model can be found in the model index.

From < https://platform.openai.com/docs/introduction>

Developer quickstart

Get up and running with the OpenAl API

The OpenAl API provides a simple interface for developers to create an intelligence layer in their applications, powered by OpenAl's state of the art models. The Chat Completions endpoint powers ChatGPT and provides a simple way to take text as input and use a model like GPT-4 to generate an output.

Want to jump straight to the code?

Skip the guickstart and dive into the API reference.

This quickstart is designed to help get your local development environment setup and send your first API request. If you are an experienced developer or want to just dive into using the OpenAI API, the <u>API reference</u> of <u>GPT guide</u> are a great place to start. Throughout this quickstart, you will learn:

- How to setup your development environment
- How to install the latest SDKs
- Some of the basic concepts of the OpenAl API
- How to send your first API request

If you run into any challenges or have questions getting started, please join our <u>developer forum</u>.

Account setup

First, create an OpenAl account or sign in. Next, navigate to the API key page and "Create new secret key", optionally naming the key. Make sure to save this somewhere safe and do not share it with anyone.

Quickstart language selection

Select the tool or language you want to get started using the OpenAl API with.

curlPythonNode.js

Python is a popular programming language that is commonly used for data applications, web development, and many other programming tasks due to its ease of use. OpenAl provides a custom Python library which makes working with the OpenAl API in Python simple and efficient.

Step 1: Setup Python

Install Python

To use the OpenAl Python library, you will need to ensure you have Python installed. Some computers come with Python pre-installed while others require that you set it up yourself. To test if you have Python installed, you can navigate to your Terminal or Command line:

- MacOS: Open Terminal: You can find it in the Applications folder or search for it using Spotlight (Command + Space).
- Windows: Open Command Prompt: You can find it by searching "cmd" in the start menu.

Next, enter the word python and then press return/enter. If you enter into the Python interpreter, then you have Python installed on your computer already and you can go to the next step. If you get an error message that says something like "Error: command python not found", you likely need to install Python and make it available in your terminal / command line.

To download Python, head to the <u>official Python website</u> and download the latest version. To use the OpenAl Python library, you need at least Python 3.7.1 or newer. If you are installing Python for the first time, you can follow the <u>official Python installation guide for beginners</u>.

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Setup a virtual environment (optional)

Once you have Python installed, it is a good practice to create a virtual python environment to install the OpenAl Python library. Virtual environments provide a clean working space for your Python packages to be installed so that you do not have conflicts with other libraries you install for other projects. You are not required to use a virtual environment, so skip to step 3 if you do not want to set one up.

To create a virtual environment, Python supplies a built in <u>venv module</u> which provides the basic functionality needed for the virtual environment setup. Running the command below will create a virtual environment named "openai-env" inside the current folder you have selected in your terminal / command line:

python -m venv openai-env

Once you've created the virtual environment, you need to activate it. On Windows, run:

openai-env\Scripts\activate
On Unix or MacOS, run:

source openai-env/bin/activate

You should see the terminal / command line interface change slightly after you active the virtual environment, it should now show "openai-env" to the left of the cursor input

section. For more details on working wit virtual environments, please refer to the <u>official</u> Python documentation.

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Install the OpenAl Python library

Once you have Python 3.7.1 or newer installed and (optionally) a virtual environment setup, the OpenAl Python library can be installed. From the terminal / command line, run:

pip install --upgrade openai

Once this completes, running pip list will show you the Python libraries you have installed in your current environment, which should confirm that the OpenAl Python library was successfully installed.

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Step 2: Setup your API key

Setup your API key for all projects (recommended)

The main advantage to making your API key accessible for all projects is that the Python library will automatically detect it and use it without having to write any code.

MacOS

Windows

- 1. Open Command Prompt: You can find it by searching "cmd" in the start menu.
- 2. Set environment variable in the current session: To set the environment variable in the current session, use the command below, replacing your-api-key-here with your actual API key:

setx OPENAI_API_KEY "your-api-key-here"

This command will set the OPENAI_API_KEY environment variable for the current session.

- **3. Permanent setup**: To make the setup permanent, add the variable through the system properties as follows:
 - o Right-click on 'This PC' or 'My Computer' and select 'Properties'.
 - Click on 'Advanced system settings'.
 - Click the 'Environment Variables' button.
 - In the 'System variables' section, click 'New...' and enter OPENAI_API_KEY as the variable name and your API key as the variable value.
- **4. Verification**: To verify the setup, reopen the command prompt and type the command below. It should display your API key: echo %OPENAI_API_KEY%

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Setup your API key for a single project

If you only want your API key to be accessible to a single project, you can create a local .env file which contains the API key and then explicitly use that API key with the Python code shown in the steps to come.

Start by going to the project folder you want to create the .env file in.

In order for your **.env** file to be ignored by version control, create a **.gitignore** file in the root of your project directory. Add a line with **.env** on it which will make sure your API

key or other secrets are not accidentally shared via version control. Once you create the .gitignore and .env files using the terminal or an integrated development environment (IDE), copy your secret API key and set it as the OPENAI_API_KEY in your .env file. If you haven't created a secret key yet, you can do so on the API key page.

The .env file should look like the following:

```
# Once you add your API key below, make sure to not share it with anyone! The API key should remain private.
```

OPENAI_API_KEY=abc123

The API key can be imported by running the code below:

```
1
2
3
4
5
6
7
from openai import OpenAI
client = OpenAI()
# defaults to getting the key using os.environ.get("OPENAL API KEY")
# if you saved the key under a different environment variable name, you can do something
like:
# client = OpenAI(
# api key=os.environ.get("CUSTOM ENV NAME"),
#)
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```

Step 3: Sending your first API request

Making an API request

After you have Python configured and an API key setup, the final step is to send a request to the OpenAI API using the Python library. To do this, create a file named openai-test.py using th terminal or an IDE.

Inside the file, copy and paste one of the examples below:

```
ChatCompletions
Copy
1
2
3
4
5
6
7
8
9
10
11
```

To run the code, enter python openai-test.py into the terminal / command line.

The <u>Chat Completions</u> example highlights just one area of strength for our models: creative ability. Explaining recursion (the programming topic) in a well formatted poem is something both the best developers and best poets would struggle with. In this case, gpt-3.5-turbo does it effortlessly.

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Next steps

Now that you have made you first OpenAl API request, it is time to explore what else is possible:

- For more detailed information on our models and the API, see our GPT guide.
- Visit the <u>OpenAl Cookbook</u> for in-depth example API use-cases, as well as code snippets for common tasks.
- Wondering what OpenAl's models are capable of? Check out our library of example prompts.
- Want to try the API without writing any code? Start experimenting in the Playground.
- Keep our <u>usage policies</u> in mind as you start building.

From < https://platform.openai.com/docs/quickstart?context=python>

Models

New models launched at DevDay

We are excited to announce the preview release of <u>GPT-4 Turbo</u> (128k context window) and an updated <u>GPT-3.5 Turbo</u> (16k context window). Among other things, both models come with improved instruction following, JSON mode, more reproducible outputs, and parallel function calling. Learn more.

Overview

The OpenAl API is powered by a diverse set of models with different capabilities and price points. You can also make customizations to our models for your specific use case with fine-tuning.

MODEL DESCRIPTION

GPT-4 and GPT-4 A set of models that improve on GPT-3.5 and can understand as well as generate

<u>Turbo</u> natural language or code

GPT-3.5 A set of models that improve on GPT-3 and can understand as well as generate

natural language or code

<u>DALL-E</u> A model that can generate and edit images given a natural language prompt

A set of models that can convert text into natural sounding spoken audio

Whisper A model that can convert audio into text

<u>Embeddings</u> A set of models that can convert text into a numerical form

Moderation A fine-tuned model that can detect whether text may be sensitive or unsafe

GPT base A set of models without instruction following that can understand as well as

generate natural language or code

GPT-3 A set of models that can understand and generate natural language

Legacy

Deprecated A full list of models that have been deprecated along with the suggested

replacement

We have also published open source models including <u>Point-E</u>, <u>Whisper</u>, <u>Jukebox</u>, and CLIP.

Visit our <u>model index for researchers</u> to learn more about which models have been featured in our research papers and the differences between model series like InstructGPT and GPT-3.5.

Continuous model upgrades

gpt-3.5-turbo, gpt-4, and gpt-4-32k point to the latest model version. You can verify this by looking at the <u>response object</u> after sending a request. The response will include the specific model version used (e.g. gpt-3.5-turbo-0613).

We also offer static model versions that developers can continue using for at least three months after an updated model has been introduced. With the new cadence of model updates, we are also giving people the ability to contribute evals to help us improve the model for different use cases. If you are interested, check out the OpenAl Evals repository.

The following models are the temporary snapshots, which we have already announced the deprecation dates of along with their replacement. If you want to use the latest model version, use the standard model names like gpt-4 or gpt-3.5-turbo.

| DISCONTINUATION DATE | REPLACEMENT MODEL |
|----------------------|--------------------|
| Jun 13, 2024 | gpt-3.5-turbo-1106 |
| Jun 13, 2024 | gpt-3.5-turbo-0613 |
| Jun 13, 2024 | gpt-4-0613 |
| Jun 13, 2024 | gpt-4-32k-0613 |
| | · |

Learn more about model deprecation on our deprecation page.

GPT-4 and **GPT-4** Turbo

GPT-4 is a large multimodal model (accepting text or image inputs and outputting text)

that can solve difficult problems with greater accuracy than any of our previous models, thanks to its broader general knowledge and advanced reasoning capabilities. GPT-4 is available in the OpenAl API to <u>paying customers</u>. Like gpt-3.5-turbo, GPT-4 is optimized for chat but works well for traditional completions tasks using the <u>Chat Completions API</u>. Learn how to use GPT-4 in our GPT guide.

| DESCRIPTION | CONTEXT WINDOW | TRAINING DATA |
|---|--|--|
| GPT-4 Turbo New The latest GPT-4 model with improved instruction following, JSON mode, reproducible outputs, parallel function calling, and more. Returns a maximum of 4,096 output tokens. This preview model is not yet suited for production traffic. Learn more. | 128,000 tokens | Up to Apr 2023 |
| GPT-4 Turbo with vision New Ability to understand images, in addition to all other GPT-4 Turbo capabilties. Returns a maximum of 4,096 output tokens. This is a preview model version and not suited yet for production traffic. Learn more. | 128,000 tokens | Up to Apr 2023 |
| Currently points to gpt-4-0613. See <u>continuous model upgrades</u> . | 8,192 tokens | Up to Sep 2021 |
| Currently points to gpt-4-32k-0613. See <u>continuous model</u> <u>upgrades</u> . | 32,768 tokens | Up to Sep 2021 |
| Snapshot of gpt-4 from June 13th 2023 with improved function calling support. | 8,192 tokens | Up to Sep 2021 |
| Snapshot of gpt-4-32k from June 13th 2023 with improved function calling support. | 32,768 tokens | Up to Sep 2021 |
| Snapshot of gpt-4 from March 14th 2023 with function calling support. This model version will be <u>deprecated</u> on June 13th 2024. | 8,192 tokens | Up to Sep 2021 |
| Snapshot of gpt-4-32k from March 14th 2023 with function calling support. This model version will be <u>deprecated</u> on June 13th 2024. | 32,768 tokens | Up to Sep 2021 |
| | GPT-4 Turbo New The latest GPT-4 model with improved instruction following, JSON mode, reproducible outputs, parallel function calling, and more. Returns a maximum of 4,096 output tokens. This preview model is not yet suited for production traffic. Learn more. GPT-4 Turbo with vision New Ability to understand images, in addition to all other GPT-4 Turbo capabilties. Returns a maximum of 4,096 output tokens. This is a preview model version and not suited yet for production traffic. Learn more. Currently points to gpt-4-0613. See continuous model upgrades. Currently points to gpt-4-32k-0613. See continuous model upgrades. Snapshot of gpt-4 from June 13th 2023 with improved function calling support. Snapshot of gpt-4-32k from June 13th 2023 with function calling support. This model version will be deprecated on June 13th 2024. Snapshot of gpt-4-32k from March 14th 2023 with function calling support. This model version will be deprecated on June 13th 2024. | GPT-4 Turbo New The latest GPT-4 model with improved instruction following, JSON mode, reproducible outputs, parallel function calling, and more. Returns a maximum of 4,096 output tokens. This preview model is not yet suited for production traffic. Learn more. GPT-4 Turbo with vision New Ability to understand images, in addition to all other GPT-4 Turbo capabilties. Returns a maximum of 4,096 output tokens. This is a preview model version and not suited yet for production traffic. Learn more. Currently points to gpt-4-0613. See continuous model upgrades. Currently points to gpt-4-32k-0613. See continuous model upgrades. Currently points to gpt-4-32k-0613. See continuous model upgrades. Snapshot of gpt-4 from June 13th 2023 with improved function calling support. Snapshot of gpt-4-32k from June 13th 2023 with improved function calling support. Snapshot of gpt-4 from March 14th 2023 with function calling support. This model version will be deprecated on June 13th 2024. Snapshot of gpt-4-32k from March 14th 2023 with function calling support. This model version will be deprecated on June 13th 2024. |

For many basic tasks, the difference between GPT-4 and GPT-3.5 models is not significant. However, in more complex reasoning situations, GPT-4 is much more capable than any of our previous models.

GPT-3.5

GPT-3.5 models can understand and generate natural language or code. Our most capable and cost effective model in the GPT-3.5 family is gpt-3.5-turbo which has been optimized for chat using the Chat Completions API but works well for traditional completions tasks as well.

| MODEL | DESCRIPTION | CONTEXT WINDOW | TRAINING DATA |
|------------|--|-------------------|------------------|
| gpt-3.5- | Updated GPT 3.5 Turbo New The latest GPT-3.5 Turbo model with improved instruction following, JSON mode, reproducible outputs, parallel function | 16,385 | Up to Sep |
| turbo-1106 | | tokens | 2021 |

| | tokens. <u>Learn more</u> . | | |
|--|---|------------------|-------------------|
| gpt-3.5- turbo | Currently points to gpt-3.5-turbo-0613. Will point to gpt-3.5-turbo-1106 starting Dec 11, 2023. See <u>continuous model upgrades</u> . | 4,096 tokens | Up to Sep 2021 |
| gpt-3.5- turbo-16k | Currently points to gpt-3.5-turbo-0613. Will point to gpt-3.5-turbo-1106 starting Dec 11, 2023. See <u>continuous model upgrades</u> . | 16,385 tokens | Up to Sep 2021 |
| gpt-3.5- turbo- instruct | Similar capabilities as text-davinci-003 but compatible with legacy Completions endpoint and not Chat Completions. | 4,096 tokens | Up to Sep 2021 |
| gpt-3.5- turbo-0613 Legacy | Snapshot of gpt-3.5-turbo from June 13th 2023. Will be deprecated on June 13, 2024. | 4,096 tokens | Up to Sep 2021 |
| gpt-3.5- turbo-16k-0 613 Legacy | Snapshot of gpt-3.5-16k-turbo from June 13th 2023. Will be deprecated on June 13, 2024. | 16,385 tokens | Up to Sep 2021 |
| gpt-3.5- turbo-0301 Legacy | Snapshot of gpt-3.5-turbo from March 1st 2023. Will be deprecated on June 13th 2024. | 4,096 tokens | Up to Sep 2021 |
| text- davinci-003 Legacy | Can do language tasks with better quality and consistency than the curie, babbage, or ada models. Will be <u>deprecated</u> on Jan 4th 2024. | 4,096 tokens | Up to Jun 2021 |
| text- davinci-002 Legacy | Similar capabilities to text-davinci-003 but trained with supervised fine-tuning instead of reinforcement learning. Will be deprecated on Jan 4th 2024. | 4,096 tokens | Up to Jun 2021 |
| code- davinci-002 Legacy | Optimized for code-completion tasks. Will be <u>deprecated</u> on Jan 4th 2024. | 8,001 tokens | Up to Jun 2021 |

calling, and more. Returns a maximum of 4.096 output

We recommend using gpt-3.5-turbo over the other GPT-3.5 models because of its lower cost and improved performance.

DALL-E

DALL·E is a AI system that can create realistic images and art from a description in natural language. DALL·E 3 currently supports the ability, given a prompt, to create a new image with a specific size. DALL·E 2 also support the ability to edit an existing image, or create variations of a user provided image.

<u>DALL-E 3</u> is available through our <u>Images API</u> along with <u>DALL-E 2</u>. You can try DALL-E 3 through <u>ChatGPT Plus</u>.

MODEL DESCRIPTION dall-e-3 DALL-E 3

New

The latest DALL·E model released in Nov 2023. Learn more.

dall-e-2 The previous DALL·E model released in Nov 2022. The 2nd iteration of DALL·E with more realistic, accurate, and 4x greater resolution images than the original model.



TTS is an AI model that converts text to natural sounding spoken text. We offer two different model variates, tts-1 is optimized for real time text to speech use cases and tts-1-hd is optimized for quality. These models can be used with the Speech endpoint in the Audio API.

MODEL DESCRIPTION

tts-1 Text-to-speech 1

New

The latest text to speech model, optimized for speed.

tts-1-hd Text-to-speech 1 HD

New

The latest text to speech model, optimized for quality.

Whisper

Whisper is a general-purpose speech recognition model. It is trained on a large dataset of diverse audio and is also a multi-task model that can perform multilingual speech recognition as well as speech translation and language identification. The Whisper v2-large model is currently available through our API with the whisper-1 model name.

Currently, there is no difference between the <u>open source version of Whisper</u> and the version available through our API. However, <u>through our API</u>, we offer an optimized inference process which makes running Whisper through our API much faster than doing it through other means. For more technical details on Whisper, you can <u>read the paper</u>.

Embeddings

Embeddings are a numerical representation of text that can be used to measure the relatedness between two pieces of text. Our second generation embedding model, text-embedding-ada-002 is a designed to replace the previous 16 first-generation embedding models at a fraction of the cost. Embeddings are useful for search, clustering, recommendations, anomaly detection, and classification tasks. You can read more about our latest embedding model in the announcement blog post.

Moderation

The Moderation models are designed to check whether content complies with OpenAl's <u>usage policies</u>. The models provide classification capabilities that look for content in the following categories: hate, hate/threatening, self-harm, sexual, sexual/minors, violence, and violence/graphic. You can find out more in our <u>moderation guide</u>.

Moderation models take in an arbitrary sized input that is automatically broken up into chunks of 4,096 tokens. In cases where the input is more than 32,768 tokens, truncation is used which in a rare condition may omit a small number of tokens from the moderation check.

The final results from each request to the moderation endpoint shows the maximum value on a per category basis. For example, if one chunk of 4K tokens had a category score of 0.9901 and the other had a score of 0.1901, the results would show 0.9901 in the API response since it is higher.

MODEL DESCRIPTION MAX

| | | TOKENS |
|----------------------------|--|--------|
| text-moderation- latest | Most capable moderation model. Accuracy will be slightly higher than the stable model. | 32,768 |
| text-moderation- stable | Almost as capable as the latest model, but slightly older. | 32,768 |

GPT base

GPT base models can understand and generate natural language or code but are not trained with instruction following. These models are made to be replacements for our original GPT-3 base models and use the legacy Completions API. Most customers should use GPT-3.5 or GPT-4.

| MODEL | DESCRIPTION | MAX TOKENS | TRAINING DATA |
|-------------|--|---------------|----------------|
| babbage-002 | Replacement for the GPT-3 ada and babbage base models. | 16,384 tokens | Up to Sep 2021 |
| davinci-002 | Replacement for the GPT-3 curie and davinci base models. | 16,384 tokens | Up to Sep 2021 |

GPT-3

Legacy

GPT-3 models can understand and generate natural language. These models were superseded by the more powerful GPT-3.5 generation models. However, the original GPT-3 base models (davinci, curie, ada, and babbage) are current the only models that are available to fine-tune.

| MODEL | DESCRIPTION | MAX TOKENS | TRAINING DATA |
|----------------------|---|-----------------|-------------------|
| text-curie-001 | Very capable, faster and lower cost than Davinci. | 2,049 tokens | Up to Oct 2019 |
| text- babbage-001 | Capable of straightforward tasks, very fast, and lower cost. | 2,049 tokens | Up to Oct 2019 |
| text-ada-001 | Capable of very simple tasks, usually the fastest model in the GPT-3 series, and lowest cost. | 2,049 tokens | Up to Oct 2019 |
| davinci | Most capable GPT-3 model. Can do any task the other models can do, often with higher quality. | 2,049 tokens | Up to Oct 2019 |
| curie | Very capable, but faster and lower cost than Davinci. | 2,049 tokens | Up to Oct 2019 |
| babbage | Capable of straightforward tasks, very fast, and lower cost. | 2,049 tokens | Up to Oct 2019 |
| ada | Capable of very simple tasks, usually the fastest model in the GPT-3 series, and lowest cost. | 2,049 tokens | Up to Oct 2019 |

How we use your data

Your data is your data.

As of March 1, 2023, data sent to the OpenAl API will not be used to train or improve OpenAl models (unless you explicitly opt in). One advantage to opting in is that the models may get better at your use case over time.

To help identify abuse, API data may be retained for up to 30 days, after which it will be

deleted (unless otherwise required by law). For trusted customers with sensitive applications, zero data retention may be available. With zero data retention, request and response bodies are not persisted to any logging mechanism and exist only in memory in order to serve the request.

Note that this data policy does not apply to OpenAl's non-API consumer services like ChatGPT or DALL-E Labs.

Default usage policies by endpoint

| ENDPOINT | DATA USED FOR TRAINING | DEFAULT RETENTION | ELIGIBLE FOR ZERO RETENTION |
|---------------------------|------------------------|---------------------------|-----------------------------|
| /v1/assistants | No | Until deleted by customer | No |
| /v1/completions | No | 30 days | Yes |
| /v1/chat/completion s* | No | 30 days | Yes* |
| /v1/edits | No | 30 days | Yes |
| /v1/images/generati ons | No | 30 days | No |
| /v1/images/edits | No | 30 days | No |
| /v1/images/variatio ns | No | 30 days | No |
| /v1/embeddings | No | 30 days | Yes |
| /v1/audio/transcripti ons | No | Zero data retention | - |
| /v1/audio/translatio ns | No | Zero data retention | - |
| /v1/audio/speech | No | 30 days | - |
| /v1/files | No | Until deleted by customer | No |
| /v1/fine_tuning/jobs | No | Until deleted by customer | No |
| /v1/fine-tunes | No | Until deleted by customer | No |
| /v1/moderations | No | Zero data retention | - |
| /v1/threads | No | 30 days | No |
| /v1/threads/messag es | No | 30 days | No |
| /v1/threads/runs | No | 30 days | No |
| /v1/threads/runs/ste ps | No | 30 days | No |

Note: At launch, image inputs via the gpt-4-vision-preview model are not eligible for zero retention.

For details, see our <u>API data usage policies</u>. To learn more about zero retention, get in touch with our <u>sales team</u>.

Model endpoint compatibility

ENDPOINT LATEST MODELS

/v1/assistants All models except gpt-3.5-turbo-0301 supported. retrieval tool requires gpt-4-1106-

preview or gpt-3.5-turbo-1106.

/v1/audio/tran whisper-1

scriptions

/v1/audio/tran whisper-1

slations

/v1/audio/spe tts-1, tts-1-hd

ech

/v1/chat/com gpt-4 and dated model releases, gpt-4-1106-preview, gpt-4-vision-

pletions preview, gpt-4-32k and dated model releases, gpt-3.5-turbo and dated model

releases, gpt-3.5-turbo-16k and dated model releases, fine-tuned versions of gpt-3.5-

turbo

/v1/completio gpt-3.5-turbo-instruct, babbage-002, davinci-002

ns (Legacy)

/v1/embeddin text-embedding-ada-002

gs

/v1/fine_tunin gpt-3.5-turbo, babbage-002, davinci-002

g/jobs

/v1/moderatio text-moderation-stable, text-moderation-latest

ns

/v1/images/g dall-e-2, dall-e-3

enerations

This list excludes all of our deprecated models.

From https://platform.openai.com/docs/models/model-endpoint-compatibility

Text generation models

New capabilities launched at DevDay

Text generation models are now capable of <u>JSON mode</u> and <u>Reproducible outputs</u>. We also launched the <u>Assistants API</u> to enable you to build agent-like experiences on top of our text-generation models.

OpenAl's text generation models (often called generative pre-trained transformers or large language models) have been trained to understand natural language, code, and images. The models provide text outputs in response to their inputs. The inputs to these models are also referred to as "prompts". Designing a prompt is essentially how you "program" a large language model model, usually by providing instructions or some examples of how to successfully complete a task.

Using OpenAl's text generation models, you can build applications to:

- · Draft documents
- Write computer code
- Answer questions about a knowledge base

- Analyze texts
- Give software a natural language interface
- Tutor in a range of subjects
- Translate languages
- Simulate characters for games

With the release of gpt-4-vision-preview, you can now build systems that also process and understand images.

Explore GPT-4 with image inputs

Check out the vision guide for more detail.

To use one of these models via the OpenAl API, you'll send a request containing the inputs and your API key, and receive a response containing the model's output. Our latest models, gpt-4 and gpt-3.5-turbo, are accessed through the chat completions API endpoint.

| | MODEL FAMILIES | API ENDPOINT |
|----------------------------|--|--|
| Newer models (2023–) | gpt-4, gpt-3.5-turbo | https://api.openai.com/v1/chat/completions |
| Updated base models (2023) | babbage-002, davinci-002 | https://api.openai.com/v1/completions |
| Legacy models (2020–2022) | text-davinci-003, text-davinci-002, davinci, curie, babbage, ada | https://api.openai.com/v1/completions |

You can experiment with various models in the <u>chat playground</u>. If you're not sure which model to use, then use gpt-3.5-turbo or gpt-4.

Chat Completions API

Chat models take a list of messages as input and return a model-generated message as output. Although the chat format is designed to make multi-turn conversations easy, it's just as useful for single-turn tasks without any conversation.

An example Chat Completions API call looks like the following:

```
python
Copy
1
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7
8
9
10
11
from openai import OpenAI
client = OpenAI()
response = client.chat.completions.create(
 model="gpt-3.5-turbo",
```

```
messages=[
    {"role": "system", "content": "You are a helpful assistant."},
    {"role": "user", "content": "Who won the world series in 2020?"},
    {"role": "assistant", "content": "The Los Angeles Dodgers won the World Series in 2020."},
    {"role": "user", "content": "Where was it played?"}
]
```

To learn more, you can view the full <u>API reference documentation</u> for the Chat API.

The main input is the messages parameter. Messages must be an array of message objects, where each object has a role (either "system", "user", or "assistant") and content. Conversations can be as short as one message or many back and forth turns.

Typically, a conversation is formatted with a system message first, followed by alternating user and assistant messages.

The system message helps set the behavior of the assistant. For example, you can modify the personality of the assistant or provide specific instructions about how it should behave throughout the conversation. However note that the system message is optional and the model's behavior without a system message is likely to be similar to using a generic message such as "You are a helpful assistant."

The user messages provide requests or comments for the assistant to respond to. Assistant messages store previous assistant responses, but can also be written by you to give examples of desired behavior.

Including conversation history is important when user instructions refer to prior messages. In the example above, the user's final question of "Where was it played?" only makes sense in the context of the prior messages about the World Series of 2020. Because the models have no memory of past requests, all relevant information must be supplied as part of the conversation history in each request. If a conversation cannot fit within the model's token limit, it will need to be shortened in some way.

To mimic the effect seen in ChatGPT where the text is returned iteratively, set the <u>stream</u> parameter to true.

Chat Completions response format

An example Chat Completions API response looks as follows:

```
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13
14
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```

```
16
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21
 "choices": [
    "finish_reason": "stop",
   "index": 0,
    "message": {
     "content": "The 2020 World Series was played in Texas at Globe Life Field in Arlington.",
     "role": "assistant"
  }
 "created": 1677664795,
 "id": "chatcmpl-7QyqpwdfhqwajiclEznoc6Q47XAyW",
 "model": "gpt-3.5-turbo-0613",
 "object": "chat.completion",
 "usage": {
  "completion_tokens": 17,
  "prompt tokens": 57,
  "total tokens": 74
 }
```

The assistant's reply can be extracted with:

python Copy response['choices'][0]['message']['content']

Every response will include a finish reason. The possible values for finish reason are:

- stop: API returned complete message, or a message terminated by one of the stop sequences provided via the stop parameter
- length: Incomplete model output due to <u>max_tokens</u> parameter or token limit
- function call: The model decided to call a function
- content_filter: Omitted content due to a flag from our content filters
- null: API response still in progress or incomplete

Depending on input parameters, the model response may include different information.

JSON mode

New

A common way to use Chat Completions is to instruct the model to always return JSON in some format that makes sense for your use case, by providing a system message. This works well, but occasionally the models may generate output that does not parse to valid JSON.

To prevent these errors and improve model performance, when calling gpt-4-1106-preview or gpt-3.5-turbo-1106, you can set <u>response_format</u> to { type: "json_object" } to enable JSON mode. When JSON mode is enabled, the model is constrained to only generate strings that parse into valid JSON.

Important notes:

- When using JSON mode, always instruct the model to produce JSON via some
 message in the conversation, for example via your system message. If you don't
 include an explicit instruction to generate JSON, the model may generate an
 unending stream of whitespace and the request may run continually until it reaches
 the token limit. To help ensure you don't forget, the API will throw an error if the
 string "JSON" does not appear somewhere in the context.
- The JSON in the message the model returns may be partial (i.e. cut off)
 if finish_reason is length, which indicates the generation exceeded max_tokens or the
 conversation exceeded the token limit. To guard against this,
 check finish_reason before parsing the response.
- JSON mode will not guarantee the output matches any specific schema, only that it is valid and parses without errors.

Note that JSON mode is always enabled when the model is generating arguments as part of <u>function calling</u>.

Reproducible outputs

Beta

Chat Completions are non-deterministic by default (which means model outputs may differ from request to request). That being said, we offer some control towards deterministic outputs by giving you access to the <u>seed</u> parameter and the <u>system_fingerprint</u> response field.

To receive (mostly) deterministic outputs across API calls, you can:

- Set the <u>seed</u> parameter to any integer of your choice and use the same value across requests you'd like deterministic outputs for.
- Ensure all other parameters (like prompt or temperature) are the exact same across requests.

Sometimes, determinism may be impacted due to necessary changes OpenAI makes to model configurations on our end. To help you keep track of these changes, we expose the system_fingerprint field. If this value is different, you may see different outputs due to changes we've made on our systems.

Deterministic outputs

Explore the new seed parameter in the OpenAI cookbook

Managing tokens

Language models read and write text in chunks called tokens. In English, a token can be as short as one character or as long as one word (e.g., a or apple), and in some languages tokens can be even shorter than one character or even longer than one word.

For example, the string "ChatGPT is great!" is encoded into six tokens: ["Chat", "G", "PT", " is", " great", "!"].

The total number of tokens in an API call affects:

- How much your API call costs, as you pay per token
- How long your API call takes, as writing more tokens takes more time

 Whether your API call works at all, as total tokens must be below the model's maximum limit (4097 tokens for gpt-3.5-turbo)

Both input and output tokens count toward these quantities. For example, if your API call used 10 tokens in the message input and you received 20 tokens in the message output, you would be billed for 30 tokens. Note however that for some models the price per token is different for tokens in the input vs. the output (see the <u>pricing</u> page for more information).

To see how many tokens are used by an API call, check the usage field in the API response (e.g., response['usage']['total_tokens']).

Chat models like gpt-3.5-turbo and gpt-4 use tokens in the same way as the models available in the completions API, but because of their message-based formatting, it's more difficult to count how many tokens will be used by a conversation.

DEEP DIVE

Counting tokens for chat API calls

To see how many tokens are in a text string without making an API call, use OpenAI's <u>tiktoken</u> Python library. Example code can be found in the OpenAI Cookbook's guide on <u>how to count tokens with tiktoken</u>.

Each message passed to the API consumes the number of tokens in the content, role, and other fields, plus a few extra for behind-the-scenes formatting. This may change slightly in the future.

If a conversation has too many tokens to fit within a model's maximum limit (e.g., more than 4097 tokens for gpt-3.5-turbo), you will have to truncate, omit, or otherwise shrink your text until it fits. Beware that if a message is removed from the messages input, the model will lose all knowledge of it.

Note that very long conversations are more likely to receive incomplete replies. For example, a gpt-3.5-turbo conversation that is 4090 tokens long will have its reply cut off after just 6 tokens.

Parameter details

Frequency and presence penalties

The frequency and presence penalties found in the <u>Chat Completions API</u> and <u>Legacy Completions API</u> can be used to reduce the likelihood of sampling repetitive sequences of tokens. They work by directly modifying the logits (un-normalized log-probabilities) with an additive contribution.

mu[j] -> mu[j] - c[j] * alpha_frequency - float(c[j] > 0) * alpha_presence Where:

- mu[j] is the logits of the j-th token
- c[j] is how often that token was sampled prior to the current position
- float(c[j] > 0) is 1 if c[j] > 0 and 0 otherwise
- alpha_frequency is the frequency penalty coefficient
- alpha_presence is the presence penalty coefficient

As we can see, the presence penalty is a one-off additive contribution that applies to all

tokens that have been sampled at least once and the frequency penalty is a contribution that is proportional to how often a particular token has already been sampled.

Reasonable values for the penalty coefficients are around 0.1 to 1 if the aim is to just reduce repetitive samples somewhat. If the aim is to strongly suppress repetition, then one can increase the coefficients up to 2, but this can noticeably degrade the quality of samples. Negative values can be used to increase the likelihood of repetition.

Completions API

The completions API endpoint received its final update in July 2023 and has a different interface than the new chat completions endpoint. Instead of the input being a list of messages, the input is a freeform text string called a prompt.

An example API call looks as follows:

```
python
Copy
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from openai import OpenAl
client = OpenAI()
response = client.completions.create(
 model="gpt-3.5-turbo-instruct",
 prompt="Write a tagline for an ice cream shop."
See the full API reference documentation to learn more.
```

Token log probabilities

The completions API can provide a limited number of log probabilities associated with the most likely tokens for each output token. This feature is controlled by using the logprobs field. This can be useful in some cases to assess the confidence of the model in its output.

Inserting text

The completions endpoint also supports inserting text by providing a suffix in addition to the standard prompt which is treated as a prefix. This need naturally arises when writing long-form text, transitioning between paragraphs, following an outline, or guiding the model towards an ending. This also works on code, and can be used to insert in the middle of a function or file.

DEEP DIVE

Inserting text

Completions response format

An example completions API response looks as follows:

2

```
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 "choices": [
   "finish_reason": "length",
   "index": 0,
   "logprobs": null,
   "text": "\n\n\"Let Your Sweet Tooth Run Wild at Our Creamy Ice Cream Shack"
 "created": 1683130927,
 "id": "cmpl-7C9Wxi9Du4j1IQjdjhxBlO22M61LD",
 "model": "gpt-3.5-turbo-instruct",
 "object": "text_completion",
 "usage": {
  "completion_tokens": 16,
  "prompt tokens": 10,
  "total_tokens": 26
 }
```

In Python, the output can be extracted with response ['choices'] [0] ['text'].

The response format is similar to the response format of the Chat Completions API but also includes the optional field logprobs.

Chat Completions vs. Completions

The Chat Completions format can be made similar to the completions format by constructing a request using a single user message. For example, one can translate from English to French with the following completions prompt:

Translate the following English text to French: "{text}" And an equivalent chat prompt would be:

[{"role": "user", "content": 'Translate the following English text to French: "{text}"'}] Likewise, the completions API can be used to simulate a chat between a user and an assistant by formatting the input accordingly.

The difference between these APIs is the underlying models that are available in each.

The chat completions API is the interface to our most capable model (gpt-4), and our most cost effective model (gpt-3.5-turbo).

Which model should I use?

We generally recommend that you use either gpt-4 or gpt-3.5-turbo. Which of these you should use depends on the complexity of the tasks you are using the models for. gpt-4 generally performs better on a wide range of <u>evaluations</u>. In particular, gpt-4 is more capable at carefully following complex instructions. By contrast gpt-3.5-turbo is more likely to follow just one part of a complex multi-part instruction. gpt-4 is less likely than gpt-3.5-turbo to make up information, a behavior known as "hallucination". gpt-4 also has a larger context window with a maximum size of 8,192 tokens compared to 4,096 tokens for gpt-3.5-turbo. However, gpt-3.5-turbo returns outputs with lower latency and costs much less per token.

We recommend experimenting in the <u>playground</u> to investigate which models provide the best price performance trade-off for your usage. A common design pattern is to use several distinct query types which are each dispatched to the model appropriate to handle them.

Prompt engineering

An awareness of the best practices for working with OpenAI models can make a significant difference in application performance. The failure modes that each exhibit and the ways of working around or correcting those failure modes are not always intuitive. There is an entire field related to working with language models which has come to be known as "prompt engineering", but as the field has progressed its scope has outgrown merely engineering the prompt into engineering systems that use model queries as components. To learn more, read our guide on prompt engineering which covers methods to improve model reasoning, reduce the likelihood of model hallucinations, and more. You can also find many useful resources including code samples in the OpenAI Cookbook.

FAQ

How should I set the temperature parameter?

Lower values for temperature result in more consistent outputs, while higher values generate more diverse and creative results. Select a temperature value based on the desired trade-off between coherence and creativity for your specific application.

Is fine-tuning available for the latest models?

Yes, for some. Currently, you can only fine-tune gpt-3.5-turbo and our updated base models (babbage-002 and davinci-002). See the <u>fine-tuning guide</u> for more details on how to use fine-tuned models.

Do you store the data that is passed into the API?

As of March 1st, 2023, we retain your API data for 30 days but no longer use your data sent via the API to improve our models. Learn more in our <u>data usage policy</u>. Some endpoints offer zero retention.

How can I make my application more safe?

If you want to add a moderation layer to the outputs of the Chat API, you can follow our moderation guide to prevent content that violates OpenAI's usage policies from

being shown.

Should I use ChatGPT or the API?

<u>ChatGPT</u> offers a chat interface to the models in the OpenAl API and a range of built-in features such as integrated browsing, code execution, plugins, and more. By contrast, using OpenAl's API provides more flexibility.

From https://platform.openai.com/docs/guides/text-generation/fag

Function calling

Learn how to connect large language models to external tools.

Introduction

In an API call, you can describe functions and have the model intelligently choose to output a JSON object containing arguments to call one or many functions. The Chat Completions API does not call the function; instead, the model generates JSON that you can use to call the function in your code.

The latest models (gpt-3.5-turbo-1006 and gpt-4-1106-preview) have been trained to both detect when a function should to be called (depending on the input) and to respond with JSON that adheres to the function signature more closely than previous models. With this capability also comes potential risks. We strongly recommend building in user confirmation flows before taking actions that impact the world on behalf of users (sending an email, posting something online, making a purchase, etc).

This guide is focused on function calling with the Chat Completions API, for details on function calling in the Assistants API, please see the <u>Assistants Tools page</u>.

Common use cases

Function calling allows you to more reliably get structured data back from the model. For example, you can:

- Create assistants that answer questions by calling external APIs (e.g. like ChatGPT Plugins)
 - e.g. define functions like send_email(to: string, body: string),
 or get_current_weather(location: string, unit: 'celsius' | 'fahrenheit')
- Convert natural language into API calls
 - e.g. convert "Who are my top customers?" to get_customers(min_revenue: int, created_before: string, limit: int) and call your internal API
- Extract structured data from text
 - e.g. define a function called extract_data(name: string, birthday: string),
 or sql_query(query: string)

...and much more!

The basic sequence of steps for function calling is as follows:

- 1. Call the model with the user query and a set of functions defined in the <u>functions</u> <u>parameter</u>.
- 2. The model can choose to call one or more functions; if so, the content will be a stringified JSON object adhering to your custom schema (note: the model may

- hallucinate parameters).
- 3. Parse the string into JSON in your code, and call your function with the provided arguments if they exist.
- 4. Call the model again by appending the function response as a new message, and let the model summarize the results back to the user.

Supported models

Not all model versions are trained with function calling data. Function calling is supported with the following models:

- gpt-4
- gpt-4-1106-preview
- gpt-4-0613
- gpt-3.5-turbo
- gpt-3.5-turbo-1106
- gpt-3.5-turbo-0613

In addition, parallel function calls is supported on the following models:

- gpt-4-1106-preview
- gpt-3.5-turbo-1106

Parallel function calling

Parallel function call is helpful for cases where you want to call multiple functions in one turn. For example, you may want to call functions to get the weather in 3 different locations at the same time. In this case, the model will call multiple functions in a single response. And you can pass back the results of each function call by referencing the tool call id in the response matching the ID of each tool call.

In this example, we define a single function <code>get_current_weather</code>. The model calls the function multiple times, and after sending the function response back to the model, we let it decide the next step. It responded with a user-facing message which was telling the user the temperature in Boston, San Francisco, and Tokyo. Depending on the query, it may choose to call a function again.

If you want to force the model to call a specific function you can do so by setting tool_choice with a specific function name. You can also force the model to generate a user-facing message by setting tool_choice: "none". Note that the default behavior (tool_choice: "auto") is for the model to decide on its own whether to call a function and if so which function to call.

Example with one function called in parallel

You can find more examples of function calling in the OpenAI cookbook:

Function calling

Learn from more examples demonstrating function calling

Tokens

Under the hood, functions are injected into the system message in a syntax the model has been trained on. This means functions count against the model's context limit and are billed as input tokens. If running into context limits, we suggest limiting the number of functions or the length of documentation you provide for function parameters.

It is also possible to use <u>fine-tuning</u> to reduce the number of tokens used if you have many functions defined.

From https://platform.openai.com/docs/guides/function-calling

Embeddings

What are embeddings?

OpenAl's text embeddings measure the relatedness of text strings. Embeddings are commonly used for:

- Search (where results are ranked by relevance to a query string)
- Clustering (where text strings are grouped by similarity)
- **Recommendations** (where items with related text strings are recommended)
- Anomaly detection (where outliers with little relatedness are identified)
- Diversity measurement (where similarity distributions are analyzed)
- Classification (where text strings are classified by their most similar label)
 An embedding is a vector (list) of floating point numbers. The <u>distance</u> between two vectors measures their relatedness. Small distances suggest high relatedness and large

vectors measures their relatedness. Small distances suggest high relatedness and large distances suggest low relatedness.

Visit our <u>pricing page</u> to learn about Embeddings pricing. Requests are billed based on the number of <u>tokens</u> in the <u>input</u> sent.

To see embeddings in action, check out our code samples

- Classification
- Topic clustering
- Search
- Recommendations

Browse Samples

How to get embeddings

To get an embedding, send your text string to the <u>embeddings API endpoint</u> along with a choice of embedding model ID (e.g., text-embedding-ada-002). The response will contain an embedding, which you can extract, save, and use.

Example requests:

```
Example: Getting embeddings

curl

Copy
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curl <a href="https://api.openai.com/v1/embeddings">https://api.openai.com/v1/embeddings</a> \
-H "Content-Type: application/json" \
```

```
-H "Authorization: Bearer $OPENAI_API_KEY" \
 -d '{
  "input": "Your text string goes here",
  "model": "text-embedding-ada-002"
Example response:
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
 "data": [
   "embedding": [
    -0.006929283495992422,
    -0.005336422007530928,
    -4.547132266452536e-05,
    -0.024047505110502243
   "index": 0,
   "object": "embedding"
  }
 "model": "text-embedding-ada-002",
 "object": "list",
 "usage": {
  "prompt_tokens": 5,
  "total_tokens": 5
See more Python code examples in the OpenAl Cookbook.
```

When using OpenAI embeddings, please keep in mind their <u>limitations and risks</u>.

Embedding models

OpenAl offers one second-generation embedding model (denoted by -002 in the model ID) and 16 first-generation models (denoted by -001 in the model ID).

We recommend using text-embedding-ada-002 for nearly all use cases. It's better, cheaper, and simpler to use. Read the blog post announcement.

| MODEL GENERATION | TOKENIZER | MAX INPUT TOKENS | KNOWLEDGE CUTOFF |
|------------------|-------------|------------------|------------------|
| V2 | cl100k base | 8191 | Sep 2021 |

V1 GPT-2/GPT-3 2046 Aug 2020

Usage is priced per input token, at a rate of \$0.0004 per 1000 tokens, or about ~3,000 pages per US dollar (assuming ~800 tokens per page):

| MODEL | ROUGH PAGES PER DOLLAR | EXAMPLE PERFORMANCE ON <u>BEIR</u> SEARCH EVAL |
|----------------------------|---------------------------|--|
| text-embedding- ada-002 | 3000 | 53.9 |
| *-davinci-*-001 | 6 | 52.8 |
| *-curie-*-001 | 60 | 50.9 |
| *-babbage-*-001 | 240 | 50.4 |
| *-ada-*-001 | 300 | 49.0 |

Second-generation models

MODEL NAME TOKENIZER MAX INPUT TOKENS OUTPUT DIMENSIONS

text-embedding-ada-002 cl100k_base 8191 1536

First-generation models (not recommended)

Use cases

Here we show some representative use cases. We will use the <u>Amazon fine-food</u> reviews dataset for the following examples.

Obtaining the embeddings

The dataset contains a total of 568,454 food reviews Amazon users left up to October 2012. We will use a subset of 1,000 most recent reviews for illustration purposes. The reviews are in English and tend to be positive or negative. Each review has a ProductId, UserId, Score, review title (Summary) and review body (Text). For example:

| PRODUCT ID | USER ID | SCORE | SUMMARY | TEXT |
|----------------|--------------------|-------|--------------------------|--|
| B001E4KFG0 | A3SGXH7AUHU8G W | 5 | Good Quality Dog Food | I have bought several of the Vitality canned |
| B00813GRG 4 | A1D87F6ZCVE5NK | 1 | Not as Advertised | Product arrived labeled as Jumbo Salted Peanut |

We will combine the review summary and review text into a single combined text. The model will encode this combined text and output a single vector embedding.

Get embeddings from dataset

1

2

3

4

5

```
6
7
8
from openai import OpenAI
client = OpenAI()
def get_embedding(text, model="text-embedding-ada-002"):
 text = text.replace("\n", " ")
 return client.embeddings.create(input = [text], model=model)['data'][0]['embedding']
df['ada_embedding'] = df.combined.apply(lambda x: get_embedding(x, model='text-
embedding-ada-002'))
df.to_csv('output/embedded_1k_reviews.csv', index=False)
To load the data from a saved file, you can run the following:
1
2
3
import pandas as pd
df = pd.read csv('output/embedded 1k reviews.csv')
df['ada_embedding'] = df.ada_embedding.apply(eval).apply(np.array)
Data visualization in 2D
Embedding as a text feature encoder for ML algorithms
Classification using the embedding features
Zero-shot classification
Obtaining user and product embeddings for cold-start
recommendation
Clustering
Text search using embeddings
Code search using embeddings
Recommendations using embeddings
```

Limitations & risks

Our embedding models may be unreliable or pose social risks in certain cases, and may cause harm in the absence of mitigations.

Social bias

Limitation: The models encode social biases, e.g. via stereotypes or negative sentiment towards certain groups.

We found evidence of bias in our models via running the SEAT (May et al, 2019) and the Winogender (Rudinger et al, 2018) benchmarks. Together, these benchmarks consist of 7 tests that measure whether models contain implicit biases when applied to gendered names, regional names, and some stereotypes.

For example, we found that our models more strongly associate (a) European American names with positive sentiment, when compared to African American names, and (b) negative stereotypes with black women.

These benchmarks are limited in several ways: (a) they may not generalize to your particular use case, and (b) they only test for a very small slice of possible social bias.

These tests are preliminary, and we recommend running tests for your specific use cases. These results should be taken as evidence of the existence of the phenomenon, not a definitive characterization of it for your use case. Please see our usage policies for more details and guidance.

Please <u>contact our support team via chat</u> if you have any questions; we are happy to advise on this.

Blindness to recent events

Limitation: Models lack knowledge of events that occurred after August 2020. Our models are trained on datasets that contain some information about real world events up until 8/2020. If you rely on the models representing recent events, then they may not perform well.

Frequently asked questions

How can I tell how many tokens a string has before I embed it?

In Python, you can split a string into tokens with OpenAl's tokenizer tiktoken.

Example code:

```
2
3
4
5
6
7
8
import tiktoken
def num tokens from string(string: str, encoding name: str) -> int:
  """Returns the number of tokens in a text string."""
  encoding = tiktoken.get_encoding(encoding_name)
  num_tokens = len(encoding.encode(string))
  return num tokens
num_tokens_from_string("tiktoken is great!", "cl100k_base")
For second-generation embedding models like text-embedding-ada-002, use
the cl100k base encoding.
```

More details and example code are in the OpenAl Cookbook guide <u>how to count tokens</u> with tiktoken.

How can I retrieve K nearest embedding vectors quickly?

For searching over many vectors quickly, we recommend using a vector database. You can find examples of working with vector databases and the OpenAl API <u>in our Cookbook</u> on GitHub.

Vector database options include:

- Chroma, an open-source embeddings store
- Elasticsearch, a popular search/analytics engine and vector database
- · Milvus, a vector database built for scalable similarity search
- <u>Pinecone</u>, a fully managed vector database
- Qdrant, a vector search engine
- Redis as a vector database
- Typesense, fast open source vector search
- Weaviate, an open-source vector search engine
- Zilliz, data infrastructure, powered by Milvus

Which distance function should I use?

We recommend <u>cosine similarity</u>. The choice of distance function typically doesn't matter much.

OpenAI embeddings are normalized to length 1, which means that:

- Cosine similarity can be computed slightly faster using just a dot product
- Cosine similarity and Euclidean distance will result in the identical rankings

Can I share my embeddings online?

Customers own their input and output from our models, including in the case of embeddings. You are responsible for ensuring that the content you input to our API does not violate any applicable law or our Terms of Use.

From https://platform.openai.com/docs/guides/embeddings/limitations-risks

Fine-tuning

Learn how to customize a model for your application.

Introduction

This guide is intended for users of the new OpenAl fine-tuning API. If you are a legacy fine-tuning user, please refer to our legacy fine-tuning guide.

Fine-tuning lets you get more out of the models available through the API by providing:

- Higher quality results than prompting
- Ability to train on more examples than can fit in a prompt
- Token savings due to shorter prompts
- Lower latency requests

OpenAl's text generation models have been pre-trained on a vast amount of text. To use the models effectively, we include instructions and sometimes several examples in a prompt. Using demonstrations to show how to perform a task is often called "few-shot learning."

Fine-tuning improves on few-shot learning by training on many more examples than can fit in the prompt, letting you achieve better results on a wide number of tasks. **Once a**

model has been fine-tuned, you won't need to provide as many examples in the prompt. This saves costs and enables lower-latency requests.

At a high level, fine-tuning involves the following steps:

- 1. Prepare and upload training data
- 2. Train a new fine-tuned model
- 3. Evaluate results and go back to step 1 if needed
- 4. Use your fine-tuned model

Visit our <u>pricing page</u> to learn more about how fine-tuned model training and usage are billed.

What models can be fine-tuned?

We are working on enabling fine-tuning for GPT-4 and expect this feature to be available later this year.

Fine-tuning is currently available for the following models:

- gpt-3.5-turbo-1006 (recommended)
- babbage-002
- davinci-002
- gpt-4-0613 (experimental eligible users can request <u>here</u>)

You can also fine-tune a fine-tuned model which is useful if you acquire additional data and don't want to repeat the previous training steps.

We expect gpt-3.5-turbo to be the right model for most users in terms of results and ease of use, unless you are migrating a legacy fine-tuned model.

When to use fine-tuning

Fine-tuning OpenAI text generation models can make them better for specific applications, but it requires a careful investment of time and effort. We recommend first attempting to get good results with prompt engineering, prompt chaining (breaking complex tasks into multiple prompts), and function calling, with the key reasons being:

- There are many tasks at which our models may not initially appear to perform well, but results can be improved with the right prompts - thus fine-tuning may not be necessary
- Iterating over prompts and other tactics has a much faster feedback loop than iterating with fine-tuning, which requires creating datasets and running training jobs
- In cases where fine-tuning is still necessary, initial prompt engineering work is not wasted - we typically see best results when using a good prompt in the fine-tuning data (or combining prompt chaining / tool use with fine-tuning)

Our <u>prompt engineering guide</u> provides a background on some of the most effective strategies and tactics for getting better performance without fine-tuning. You may find it helpful to iterate quickly on prompts in our <u>playground</u>.

Common use cases

Some common use cases where fine-tuning can improve results:

Setting the style, tone, format, or other qualitative aspects

- Improving reliability at producing a desired output
- Correcting failures to follow complex prompts
- Handling many edge cases in specific ways
- Performing a new skill or task that's hard to articulate in a prompt One high-level way to think about these cases is when it's easier to "show, not tell". In the sections to come, we will explore how to set up data for fine-tuning and various examples where fine-tuning improves the performance over the baseline model.

Another scenario where fine-tuning is effective is in reducing costs and / or latency, by replacing GPT-4 or by utilizing shorter prompts, without sacrificing quality. If you can achieve good results with GPT-4, you can often reach similar quality with a fine-tuned gpt-3.5-turbo model by fine-tuning on the GPT-4 completions, possibly with a shortened instruction prompt.

Preparing your dataset

Once you have determined that fine-tuning is the right solution (i.e. you've optimized your prompt as far as it can take you and identified problems that the model still has), you'll need to prepare data for training the model. You should create a diverse set of demonstration conversations that are similar to the conversations you will ask the model to respond to at inference time in production.

Each example in the dataset should be a conversation in the same format as our Completions API, specifically a list of messages where each message has a role, content, and optional name. At least some of the training examples should directly target cases where the prompted model is not behaving as desired, and the provided assistant messages in the data should be the ideal responses you want the model to provide.

Example format

In this example, our goal is to create a chatbot that occasionally gives sarcastic responses, these are three training examples (conversations) we could create for a dataset:

```
2
3
{"messages": [{"role": "system", "content": "Marv is a factual chatbot that is also sarcastic."},
{"role": "user", "content": "What's the capital of France?"}, {"role": "assistant", "content": "Paris,
as if everyone doesn't know that already."}]}
{"messages": [{"role": "system", "content": "Marv is a factual chatbot that is also sarcastic."},
{"role": "user", "content": "Who wrote 'Romeo and Juliet'?"}, {"role": "assistant", "content": "Oh,
just some guy named William Shakespeare. Ever heard of him?"}]}
{"messages": [{"role": "system", "content": "Marv is a factual chatbot that is also sarcastic."},
{"role": "user", "content": "How far is the Moon from Earth?"}, {"role": "assistant", "content":
"Around 384,400 kilometers. Give or take a few, like that really matters."}]}
The conversational chat format is required to fine-tune gpt-3.5-turbo.
For babbage-002 and davinci-002, you can follow the prompt completion pair format used
for legacy fine-tuning as shown below.
```

```
2
3
{"prompt": "<prompt text>", "completion": "<ideal generated text>"}
{"prompt": "<prompt text>", "completion": "<ideal generated text>"}
{"prompt": "<prompt text>", "completion": "<ideal generated text>"}
```

Crafting prompts

We generally recommend taking the set of instructions and prompts that you found worked best for the model prior to fine-tuning, and including them in every training example. This should let you reach the best and most general results, especially if you have relatively few (e.g. under a hundred) training examples.

If you would like to shorten the instructions or prompts that are repeated in every example to save costs, keep in mind that the model will likely behave as if those instructions were included, and it may be hard to get the model to ignore those "baked-in" instructions at inference time.

It may take more training examples to arrive at good results, as the model has to learn entirely through demonstration and without guided instructions.

Example count recommendations

To fine-tune a model, you are required to provide at least 10 examples. We typically see clear improvements from fine-tuning on 50 to 100 training examples with gpt-3.5-turbo but the right number varies greatly based on the exact use case.

We recommend starting with 50 well-crafted demonstrations and seeing if the model shows signs of improvement after fine-tuning. In some cases that may be sufficient, but even if the model is not yet production quality, clear improvements are a good sign that providing more data will continue to improve the model. No improvement suggests that you may need to rethink how to set up the task for the model or restructure the data before scaling beyond a limited example set.

Train and test splits

After collecting the initial dataset, we recommend splitting it into a training and test portion. When submitting a fine-tuning job with both training and test files, we will provide statistics on both during the course of training. These statistics will be your initial signal of how much the model is improving. Additionally, constructing a test set early on will be useful in making sure you are able to evaluate the model after training, by generating samples on the test set.

Token limits

Each training example is limited to 4096 tokens. Examples longer than this will be truncated to the first 4096 tokens when training. To be sure that your entire training example fits in context, consider checking that the total token counts in the message contents are under 4,000.

You can compute token counts using our <u>counting tokens notebook</u> from the OpenAl cookbook.

Estimate costs

Please refer to the <u>pricing page</u> for details on cost per 1k input and output tokens (we do to charge for tokens that are part of the validation data). To estimate the costs for a specific fine-tuning job, use the following formula:

base cost per 1k tokens * number of tokens in the input file * number of epochs trained For a training file with 100,000 tokens trained over 3 epochs, the expected cost would be ~\$2.40 USD.

Check data formatting

Once you have compiled a dataset and before you create a fine-tuning job, it is important to check the data formatting. To do this, we created a simple Python script which you can use to find potential errors, review token counts, and estimate the cost of a fine-tuning job.

Fine-tuning data format validation

Learn about fine-tuning data formatting

Upload a training file

Once you have the data validated, the file needs to be uploaded using the <u>Files API</u> in order to be used with a fine-tuning jobs:

```
python
Copy
1
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from openai import OpenAl
client = OpenAl()
client.files.create(
  file=open("mydata.jsonl", "rb"),
  purpose="fine-tune"
)
```

After you upload the file, it may take some time to process. While the file is processing, you can still create a fine-tuning job but it will not start until the file processing has completed.

Create a fine-tuned model

After ensuring you have the right amount and structure for your dataset, and have uploaded the file, the next step is to create a fine-tuning job. We support creating fine-tuning jobs via the <u>fine-tuning UI</u> or programmatically.

To start a fine-tuning job using the OpenAI SDK:

```
python
Copy
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from openai import OpenAl client = OpenAl()
client.fine_tuning.jobs.create(
```

```
training_file="file-abc123",
model="gpt-3.5-turbo"
```

In this example, model is the name of the model you want to fine-tune (gpt-3.5-turbo, babbage-002, davinci-002, or an existing fine-tuned model) and training_file is the file ID that was returned when the training file was uploaded to the OpenAl API. You can customize your fine-tuned model's name using the <u>suffix parameter</u>.

To set additional fine-tuning parameters like the validation_file or hyperparameters, please refer to the API specification for fine-tuning.

After you've started a fine-tuning job, it may take some time to complete. Your job may be queued behind other jobs in our system, and training a model can take minutes or hours depending on the model and dataset size. After the model training is completed, the user who created the fine-tuning job will receive an email confirmation.

In addition to creating a fine-tuning job, you can also list existing jobs, retrieve the status of a job, or cancel a job.

```
python
Copy
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15
16
17
from openai import OpenAI
client = OpenAI()
# List 10 fine-tuning jobs
client.fine_tuning.jobs.list(limit=10)
# Retrieve the state of a fine-tune
client.fine_tuning.jobs.retrieve("ftjob-abc123")
# Cancel a job
client.fine_tuning.jobs.cancel("ftjob-abc123")
# List up to 10 events from a fine-tuning job
client.fine_tuning.jobs.list_events(id="ftjob-abc123", limit=10)
# Delete a fine-tuned model (must be an owner of the org the model was created in)
client.models.delete("ft:gpt-3.5-turbo:acemeco:suffix:abc123")
```

Use a fine-tuned model

When a job has succeeded, you will see the fine_tuned_model field populated with the name of the model when you retrieve the job details. You may now specify this model as a parameter to in the Chat Completions (for gpt-3.5-turbo) or Legacy Completions API (for babbage-002 and davinci-002), and make requests to it using the Playground.

After your job is completed, the model should be available right away for inference use. In some cases, it may take several minutes for your model to become ready to handle requests. If requests to your model time out or the model name cannot be found, it is likely because your model is still being loaded. If this happens, try again in a few minutes.

```
python
Copy
1
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11
from openai import OpenAI
client = OpenAI()
response = client.chat.completions.create(
 model="ft:gpt-3.5-turbo:my-org:custom_suffix:id",
 messages=[
  {"role": "system", "content": "You are a helpful assistant."},
  {"role": "user", "content": "Hello!"}
1
)
```

print(completion.choices[0].message)

You can start making requests by passing the model name as shown above and in our GPT guide.

Analyzing your fine-tuned model

We provide the following training metrics computed over the course of training: training loss, training token accuracy, test loss, and test token accuracy. These statistics are meant to provide a sanity check that training went smoothly (loss should decrease, token accuracy should increase). While an active fine-tuning jobs is running, you can view an event object which contains some useful metrics:

```
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6
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8
9
10
11
12
13
{
"object": "fine_tuning.job.event",
```

```
"id": "ftevent-abc-123",
"created_at": 1693582679,
"level": "info",
"message": "Step 100/100: training loss=0.00",
"data": {
    "step": 100,
    "train_loss": 1.805623287509661e-5,
    "train_mean_token_accuracy": 1.0
},
"type": "metrics"
```

After a fine-tuning job has finished, you can also see metrics around how the training process went by <u>querying a fine-tuning job</u>, extracting a file ID from the result_files, and then <u>retrieving that files content</u>. Each results CSV file has the following columns: step, train_loss, train_accuracy, valid_loss, and valid_mean_token_accuracy.

```
1 2 3 4 4 5 6 5 5 6 5 5 5 5 6 5 5 5 7 19,0.0,, 2,0.57719,0.0,, 3,3.63525,0.0,, 4,1.72257,0.0,, 5,1.52379,0.0,
```

While metrics can he helpful, evaluating samples from the fine-tuned model provides the most relevant sense of model quality. We recommend generating samples from both the base model and the fine-tuned model on a test set, and comparing the samples side by side. The test set should ideally include the full distribution of inputs that you might send to the model in a production use case. If manual evaluation is too time-consuming, consider using our Evals library to automate future evaluations.

Iterating on data quality

If the results from a fine-tuning job are not as good as you expected, consider the following ways to adjust the training dataset:

- Collect examples to target remaining issues
 - If the model still isn't good at certain aspects, add training examples that directly show the model how to do these aspects correctly
- · Scrutinize existing examples for issues
 - If your model has grammar, logic, or style issues, check if your data has any
 of the same issues. For instance, if the model now says "I will schedule this
 meeting for you" (when it shouldn't), see if existing examples teach the model
 to say it can do new things that it can't do
- Consider the balance and diversity of data
 - If 60% of the assistant responses in the data says "I cannot answer this", but at inference time only 5% of responses should say that, you will likely get an overabundance of refusals
- Make sure your training examples contain all of the information needed for the

response

- If we want the model to compliment a user based on their personal traits and a training example includes assistant compliments for traits not found in the preceding conversation, the model may learn to hallucinate information
- Look at the agreement / consistency in the training examples
 - If multiple people created the training data, it's likely that model performance will be limited by the level of agreement / consistency between people. For instance, in a text extraction task, if people only agreed on 70% of extracted snippets, the model would likely not be able to do better than this
- Make sure your all of your training examples are in the same format, as expected for inference

<u>Iterating on data quantity</u>

Once you're satisfied with the quality and distribution of the examples, you can consider scaling up the number of training examples. This tends to help the model learn the task better, especially around possible "edge cases". We expect a similar amount of improvement every time you double the number of training examples. You can loosely estimate the expected quality gain from increasing the training data size by:

- Fine-tuning on your current dataset
- Fine-tuning on half of your current dataset
- Observing the quality gap between the two

In general, if you have to make a trade-off, a smaller amount of high-quality data is generally more effective than a larger amount of low-quality data.

<u>Iterating on hyperparameters</u>

We allow you to specify the following hyperparameters:

- epochs
- learning rate multiplier
- batch size

We recommend initially training without specifying any of these, allowing us to pick a default for you based on dataset size, then adjusting if you observe the following:

- If the model does not follow the training data as much as expected increase the number of epochs by 1 or 2
 - This is more common for tasks for which there is a single ideal completion (or a small set of ideal completions which are similar). Some examples include classification, entity extraction, or structured parsing. These are often tasks for which you can compute a final accuracy metric against a reference answer.
- If the model becomes less diverse than expected decrease the number of epochs by 1 or 2
 - This is more common for tasks for which there are a wide range of possible good completions
- If the model does not appear to be converging, increase the learning rate multiplier You can set the hyperparameters as is shown below:

python Copy

```
1
2
3
4
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8
9
10
from openai import OpenAI
client = OpenAI()
client.fine_tuning.jobs.create(
 training_file="file-abc123",
 model="gpt-3.5-turbo",
 hyperparameters={
  "n_epochs":2
 }
)
```

Fine-tuning examples

Now that we have explored the basics of the fine-tuning API, let's look at going through the fine-tuning lifecycle for a few different use cases.

Style and tone

In this example, we will explore how to build a fine-tuned model which gets the model follow specific style and tone guidance beyond what is possible with prompting alone.

To begin, we create a sample set of messages showing what the model should which in this case is misspelled words.

```
1
2
3
{"messages": [{"role": "system", "content": "Marv is a factual chatbot that is also sarcastic."},
{"role": "user", "content": "What's the capital of France?"}, {"role": "assistant", "content": "Paris,
as if everyone doesn't know that already."}]}
{"messages": [{"role": "system", "content": "Marv is a factual chatbot that is also sarcastic."},
{"role": "user", "content": "Who wrote 'Romeo and Juliet'?"}, {"role": "assistant", "content": "Oh,
just some guy named William Shakespeare. Ever heard of him?"}]}
{"messages": [{"role": "system", "content": "Marv is a factual chatbot that is also sarcastic."},
{"role": "user", "content": "How far is the Moon from Earth?"}, {"role": "assistant", "content":
"Around 384,400 kilometers. Give or take a few, like that really matters."}]}
If you want to follow along and create a fine-tuned model yourself, you will need at least
10 examples.
```

After getting the data that will potentially improve the model, the next step is to check if the data meets all the <u>formatting requirements</u>.

Now that we have the data formatted and validated, the final training step is to kick off a job to create the fine-tuned model. You can do this via the OpenAl CLI or one of our SDKs as shown below:

```
2
3
4
5
6
7
8
9
10
11
from openai import OpenAI
client = OpenAI()
file = client.files.create(
 file=open("marv.jsonl", "rb"),
 purpose="fine-tune"
client.fine_tuning.jobs.create(
 training_file=file.id,
 model="gpt-3.5-turbo"
Once the training job is done, you will be able to use your fine-tuned model.
```

Collapse

Structured output

Another type of use case which works really well with fine-tuning is getting the model to provide structured information, in this case about sports headlines:

```
1
2
3
4
{"messages": [{"role": "system", "content": "Given a sports headline, provide the following fields in a JSON dict, where applicable: "player" (full name)", "team", "sport", and "gender".},{"role": "user", "content": "Sources: Colts grant RB Taylor OK to seek trade"}, {"role": "assistant", "content": "{"player": "Jonathan Taylor", "team": "Colts", "sport": "football", "gender": "male" }"},]}
{"messages": [{"role": "system", "content": "Given a sports headline, provide the following fields in a JSON dict, where applicable: "player" (full name)", "team", "sport", and "gender".},{"role": "user", "content": "OSU 'split down middle' on starting QB battle"}, {"role": "assistant", "content": "{"player": null, "team": "OSU", "sport": "football", "gender": null }"},]}
If you want to follow along and create a fine-tuned model yourself, you will need at least 10 examples.
```

After getting the data that will potentially improve the model, the next step is to check if the data meets all the formatting requirements.

Now that we have the data formatted and validated, the final training step is to kick off a job to create the fine-tuned model. You can do this via the OpenAl CLI or one of our SDKs as shown below:

```
1
2
3
4
5
6
7
8
9
10
11
12
from openai import OpenAI
client = OpenAI()
file = client.files.create(
 file=open("sports-context.jsonl", "rb"),
 purpose="fine-tune"
client.fine_tuning.jobs.create(
 training_file=file.id,
 model="gpt-3.5-turbo"
Once the training job is done, you will be able to use your fine-tuned model and make a
request that looks like the following:
1
2
3
4
5
6
7
8
completion = client.chat.completions.create(
 model="ft:gpt-3.5-turbo:my-org:custom_suffix:id",
 messages=[
  {"role": "system", "content": "Given a sports headline, provide the following fields in a JSON
dict, where applicable: player (full name), team, sport, and gender"},
  {"role": "user", "content": "Richardson wins 100m at worlds to cap comeback"}
]
print(completion.choices[0].message)
Based on the formatted training data, the response should look like the following:
{"player": "Sha'Carri Richardson", "team": null", "sport": "track and field", "gender": "female"}
Collapse
Function calling
The chat completions API supports <u>function calling</u>. Including a long list of functions in
```

the completions API can consume a considerable number of prompt tokens and sometimes the model hallucinates or does not provide valid JSON output.

Fine-tuning a model with function calling examples can allow you to:

- Get similarly formatted responses even when the full function definition isn't present
- Get more accurate and consistent outputs

Format your examples as shown, with each line including a list of "messages" and an optional list of "functions":

```
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
  "messages": [
     {"role": "user", "content": "What is the weather in San Francisco?"},
     {"role": "assistant", "function_call": {"name": "get_current_weather", "arguments":
"{\"location\": \"San Francisco, USA\", \"format\": \"celcius\"}"}
  ],
"functions": [{
     "name": "get_current_weather",
     "description": "Get the current weather",
     "parameters": {
        "type": "object",
        "properties": {
          "location": {"type": "string", "description": "The city and country, eg. San Francisco,
USA"},
          "format": {"type": "string", "enum": ["celsius", "fahrenheit"]}
        "required": ["location", "format"]
  }]
```

If you want to follow along and create a fine-tuned model yourself, you will need at least 10 examples.

If your goal is to use less tokens, some useful techniques are:

Omit function and parameter descriptions: remove the description field from

function and parameters

- Omit parameters: remove the entire properties field from the parameters object
- Omit function entirely: remove the entire function object from the functions array If your goal is to maximize the correctness of the function calling output, we recommend using the same function definitions for both training and querying the fine-tuned model.

Fine-tuning on function calling can also be used to customize the model's response to function outputs. To do this you can include a function response message and an assistant message interpreting that response:

```
1
2
3
4
5
6
7
8
9
  "messages": [
     {"role": "user", "content": "What is the weather in San Francisco?"},
     {"role": "assistant", "function_call": {"name": "get_current_weather", "arguments":
"{\"location\": \"San Francisco, USA\", \"format\": \"celcius\"}"}}
     {"role": "function", "name": "get_current_weather", "content": "21.0"},
     {"role": "assistant", "content": "It is 21 degrees celsius in San Francisco, CA"}
  "functions": [...] // same as before
Collapse
```

Migration of legacy models

For users migrating from /v1/fine-tunes to the updated /v1/fine_tuning/jobs API and newer models, the main difference you can expect is the updated API. The legacy prompt completion pair data format has been retained for the updated babbage-002 and davinci-002 models to ensure a smooth transition. The new models will support fine-tuning with 4k token context and have a knowledge cutoff of September 2021.

For most tasks, you should expect to get better performance from gpt-3.5-turbo than from the GPT base models.

FAQ

When should I use fine-tuning vs embeddings with retrieval?

Embeddings with retrieval is best suited for cases when you need to have a large database of documents with relevant context and information.

By default OpenAl's models are trained to be helpful generalist assistants. Fine-tuning can be used to make a model which is narrowly focused, and exhibits specific ingrained behavior patterns. Retrieval strategies can be used to make new information available to a model by providing it with relevant context before generating its response. Retrieval strategies are not an alternative to fine-tuning and can in fact be complementary to it.

When can I fine-tune GPT-4 or GPT-3.5-Turbo-16k?

We plan to release support for fine-tuning both of these models later this year.

How do I know if my fine-tuned model is actually better than the base model?

We recommend generating samples from both the base model and the fine-tuned model on a test set of chat conversations, and comparing the samples side by side. For more comprehensive evaluations, consider using the OpenAl evals framework to create an eval specific to your use case.

Can I continue fine-tuning a model that has already been fine-tuned?

Yes, you can pass the name of a fine-tuned model into the model parameter when creating a fine-tuning job. This will start a new fine-tuning job using the fine-tuned model as the starting point.

How can I estimate the cost of fine-tuning a model?

Please refer to the <u>estimate cost</u> section above.

Does the new fine-tuning endpoint still work with Weights & Biases for tracking metrics?

No, we do not currently support this integration but are working to enable it in the near future.

How many fine-tuning jobs can I have running at once?

Please refer to our <u>rate limit guide</u> for the most up to date information on the limits.

How do rate limits work on fine-tuned models?

A fine-tuned model pulls from the same shared rate limit as the model it is based off of. For example, if you use half your TPM rate limit in a given time period with the standard gpt-3.5-turbo model, any model(s) you fine-tuned from gpt-3.5-turbo would only have the remaining half of the TPM rate limit accessible since the capacity is shared across all models of the same type.

Put another way, having fine-tuned models does not give you more capacity to use our models from a total throughput perspective.

From https://platform.openai.com/docs/guides/fine-tuning

Image generation

Learn how to generate or manipulate images with our DALL-E models.

Introduction

The Images API provides three methods for interacting with images:

1. Creating images from scratch based on a text prompt (DALL-E 3 and DALL-E 2)

- 2. Creating edited versions of images by having the model replace some areas of a pre-existing image, based on a new text prompt (DALL-E 2 only)
- 3. Creating variations of an existing image (DALL-E 2 only)

This guide covers the basics of using these three API endpoints with useful code samples. To try DALL-E 3, head to ChatGPT. To try DALL-E 2, check out the DALL-E preview app.

Usage

Generations

The <u>image generations</u> endpoint allows you to create an original image given a text prompt. When using DALL-E 3, images can have a size of 1024x1024, 1024x1792 or 1792x1024 pixels.

By default, images are generated at standard quality, but when using DALL-E 3 you can set quality: "hd" for enhanced detail. Square, standard quality images are the fastest to generate.

You can request 1 image at a time with DALL-E 3 (request more by making parallel requests) or up to 10 images at a time using DALL-E 2 with the n parameter.

When you send a generation request to DALL-E 3, we will automatically re-write it for safety reasons, and to add more detail (because more detailed prompts generally result in higher quality images).

```
Generate an image
python
Copy
2
3
4
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6
7
8
9
10
11
from openai import OpenAI
client = OpenAI()
response = client.images.generate(
 model="dall-e-3",
 prompt="a white siamese cat",
 size="1024x1024",
 quality="standard",
 n=1.
)
image_url = response.data[0].url
```

What is new with DALL-E 3

Explore what is new with DALL-E 3 in the OpenAl Cookbook

Example DALL-E 3 generations

GENERATION

A photograph of a white Siamese cat.

Each image can be returned as either a URL or Base64 data, using the <u>response_format</u> parameter. URLs will expire after an hour.

Edits (DALL-E 2 only)

Also known as "inpainting", the <u>image edits</u> endpoint allows you to edit or extend an image by uploading an image and mask indicating which areas should be replaced. The transparent areas of the mask indicate where the image should be edited, and the prompt should describe the full new image, **not just the erased area**. This endpoint can enable experiences like <u>the editor in our DALL-E preview app</u>.

```
Edit an image
python
Copy
1
2
3
4
5
6
7
8
9
10
11
12
from openai import OpenAI
client = OpenAI()
response = client.images.edit((
 model="dall-e-2",
 image=open("sunlit_lounge.png", "rb"),
 mask=open("mask.png", "rb"),
 prompt="A sunlit indoor lounge area with a pool containing a flamingo",
 n=1
 size="1024x1024"
image_url = response.data[0].url
IMAGE MASK
                  OUTPUT
```

Prompt: a sunlit indoor lounge area with a pool containing a flamingo

The uploaded image and mask must both be square PNG images less than 4MB in size, and also must have the same dimensions as each other. The non-transparent areas of the mask are not used when generating the output, so they don't necessarily need to match the original image like the example above.

Variations (DALL-E 2 only)

The image variations endpoint allows you to generate a variation of a given image.

Generate an image variation python Copy

```
1
2
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9
10
from openai import OpenAI
client = OpenAI()
response = client.images.create_variation(
 image=open("image_edit_original.png", "rb"),
 n=2
 size="1024x1024"
image_url = response.data[0].url
IMAGE
         OUTPUT
```

Similar to the edits endpoint, the input image must be a square PNG image less than 4MB in size.

Content moderation

Prompts and images are filtered based on our <u>content policy</u>, returning an error when a prompt or image is flagged.

Language-specific tips

Node.jsPython

Using in-memory image data

The Node.js examples in the guide above use the fs module to read image data from disk. In some cases, you may have your image data in memory instead. Here's an example API call that uses image data stored in a Node.js Buffer object:

```
1
2
3
4
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8
9
// This is the Buffer object that contains your image data
const buffer = [your image data];
// Set a `name` that ends with .png so that the API knows it's a PNG image
buffer.name = "image.png";
const response = await openai.createImageVariation(
 buffer,
 "1024x1024"
```

Working with TypeScript

1

If you're using TypeScript, you may encounter some quirks with image file arguments. Here's an example of working around the type mismatch by explicitly casting the argument:

```
2
3
4
5
// Cast the ReadStream to `any` to appease the TypeScript compiler
const response = await openai.createImageVariation(
 fs.createReadStream("image.png") as any,
 "1024x1024"
);
And here's a similar example for in-memory image data:
1
2
3
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8
9
10
// This is the Buffer object that contains your image data
const buffer: Buffer = [your image data];
// Cast the buffer to `any` so that we can set the `name` property
const file: any = buffer;
// Set a `name` that ends with .png so that the API knows it's a PNG image
file.name = "image.png";
const response = await openai.createImageVariation(
file,
 1,
 "1024x1024"
```

Error handling

API requests can potentially return errors due to invalid inputs, rate limits, or other issues. These errors can be handled with a try...catch statement, and the error details can be found in either error.response or error.message:

```
1
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```

```
8
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11
12
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14
15
try {
 const response = await openai.createImageVariation(
  fs.createReadStream("image.png"),
  "1024x1024"
 );
 console.log(response.data.data[0].url);
} catch (error) {
 if (error.response) {
  console.log(error.response.status);
  console.log(error.response.data);
 } else {
  console.log(error.message);
```

From https://platform.openai.com/docs/guides/images/language-specific-tips?context=node

Vision

7

Learn how to use GPT-4 to understand images

Introduction

GPT-4 with Vision, sometimes referred to as <u>GPT-4V</u> or gpt-4-vision-preview in the API, allows the model to take in images and answer questions about them. Historically, language model systems have been limited by taking in a single input modality, text. For many use cases, this constrained the areas where models like GPT-4 could be used.

GPT-4 with vision is currently available to all <u>developers who have access to GPT-4</u> via the gpt-4-vision-preview model and the Chat Completions API which has been updated to support image inputs. Note that the <u>Assistants API</u> does not currently support image inputs.

It is important to note the following:

- GPT-4 with vision is not a model that behaves differently from GPT-4, with the small exception of the system prompt we use for the model
- GPT-4 with vision is not a different model that does worse at text tasks because it has vision, it is simply GPT-4 with vision added
- GPT-4 with vision is an augmentative set of capabilities for the model

<u>Quick start</u>

Images can are made available to the model in two main ways: by passing a link to the image or by passing the base64 encoded image directly in the request. Images can be passed in the user, system and assistant messages. Currently we don't support images in

the first system message but this may change in the future.

```
Speak this text
python
Copy
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5
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11
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15
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20
21
22
from openai import OpenAI
client = OpenAI()
response = client.chat.completions.create(
  model="gpt-4-vision-preview",
  messages=[
       "role": "user",
       "content": [
         {"type": "text", "text": "What's in this image?"},
            "type": "image_url",
            "image_url": "https://upload.wikimedia.org/wikipedia/commons/thumb/d/dd/Gfp-wisconsin-
madison-the-nature-boardwalk.jpg/2560px-Gfp-wisconsin-madison-the-nature-boardwalk.jpg",
         },
       ],
    }
  ],
  max_tokens=300,
print(response.choices[0])
```

The model is best at answering general questions about what is present in the images. While it does understand the relationship between objects in images, it is not yet optimized to answer detailed questions about the location of certain objects in an image. For example, you can ask it what color a car is or what some ideas for dinner might be based on what is in you fridge, but if you show it an image of a room and ask it where the chair is, it may not answer the question correctly.

It is important to keep in mind the <u>limitations of the model</u> as you explore what use-cases visual understanding can be applied to.

Video understanding with vision

Learn how to use use GPT-4 with Vision to understand videos in the OpenAl Cookbook

Uploading base 64 encoded images

If you have an image or set of images locally, you can pass those to the model in base 64 encoded format, here is an example of this in action:

```
import base64
import requests
# OpenAl API Key
api_key = "YOUR_OPENAI_API_KEY"
# Function to encode the image
def encode_image(image_path):
  with open(image_path, "rb") as image_file:
     return base64.b64encode(image_file.read()).decode('utf-8')
# Path to your image
image_path = "path_to_your_image.jpg"
# Getting the base64 string
base64_image = encode_image(image_path)
headers = {
  "Content-Type": "application/json",
  "Authorization": f"Bearer {api_key}"
}
payload = {
  "model": "gpt-4-vision-preview",
  "messages": [
     "role": "user",
     "content": [
       "type": "text",
       "text": "What's in this image?"
      },
       "type": "image_url",
       "image_url": {
        "url": f"data:image/jpeg;base64,{base64_image}"
       }
      }
   }
  "max_tokens": 300
}
response = requests.post("https://api.openai.com/v1/chat/completions", headers=headers,
json=payload)
print(response.json())
```

Multiple image inputs

The Chat Completions API is capable of taking in and processing multiple image inputs in both base64 encoded format or as an image URL. The model will process each image and use the information from all of them to answer the question.

```
Multiple image inputs
python
Copy
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14
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16
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18
19
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21
22
23
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26
27
from openai import OpenAI
client = OpenAI()
response = client.chat.completions.create(
  model="gpt-4-vision-preview",
  messages=[
    {
       "role": "user",
       "content": [
         {
            "type": "image_url",
            "image_url": "https://upload.wikimedia.org/wikipedia/commons/thumb/d/dd/Gfp-wisconsin-
madison-the-nature-boardwalk.jpg/2560px-Gfp-wisconsin-madison-the-nature-boardwalk.jpg",
         },
         {
            "type": "text",
            "text": "What's in these images? Is there any difference between them?",
         },
         {
            "type": "image_url",
            "image_url": "https://upload.wikimedia.org/wikipedia/commons/thumb/d/dd/Gfp-wisconsin-
```

madison-the-nature-boardwalk.jpg/2560px-Gfp-wisconsin-madison-the-nature-boardwalk.jpg",

```
},
    ],
    ],
    ],
    max_tokens=300,
```

print(response.choices[0])

Here the model is shown two copies of the same image and can answer questions about both or each of the images independently.

Low or high fidelity image understanding

By controlling the detail parameter, which has two options, low or high, you have control over how the model processes the image and generates its textual understanding.

- low will disable the "high res" model. The model will receive a low-res 512 x 512 version of the image, and represent the image with a budget of 65 tokens. This allows the API to return faster responses and consume fewer input tokens for use cases that do not require high detail.
- high will enable "high res" mode, which first allows the model to see the low res image and then creates detailed crops of input images as 512px squares based on the input image size. Each of the detailed crops uses twice the token budget (65 tokens) for a total of 129 tokens.

Managing images

The Chat Completions API, unlike the Assistants API, is not stateful. That means you have to manage the messages (including images) you pass to the model yourself. If you want to pass the same image to the model multiple times, you will have to pass the image each time you make a request to the API.

For long running conversations, we suggest passing images via URL's instead of base64. The latency of the model can also be improved by downsizing your images ahead of time to be less than the maximum size they are expected them to be. For low res mode, we expect a 512px x 512px image. For high rest mode, the short side of the image should be less than 768px and the long side should be less than 2,000px.

After an image has been processed by the model, it is deleted from OpenAl servers and not retained. We do not use data uploaded via the OpenAl API to train our models.

Limitations

While GPT-4 with vision is powerful and can be used in many situations, it is important to understand the limitations of the model. Here are some of the limitations we are aware of:

- Medical images: The model is not suitable for interpreting specialized medical images like CT scans and shouldn't be used for medical advice.
- Non-English: The model may not perform optimally when handling images with text of non-Latin alphabets, such as Japanese or Korean.
- Big text: Enlarge text within the image to improve readability, but avoid cropping important details.

- Rotation: The model may misinterpret rotated / upside-down text or images.
- Visual elements: The model may struggle to understand graphs or text where colors or styles like solid, dashed, or dotted lines vary.
- Spatial reasoning: The model struggles with tasks requiring precise spatial localization, such as identifying chess positions.
- Accuracy: The model may generate incorrect descriptions or captions in certain scenarios.
- Image shape: The model struggles with panoramic and fisheye images.
- Metadata and resizing: The model doesn't process original file names or metadata, and images are resized before analysis, affecting their original dimensions.
- Counting: May give approximate counts for objects in images.
- CAPTCHAS: For safety reasons, we have implemented a system to block the submission of CAPTCHAS.

Calculating costs

Image inputs are metered and charged in tokens, just as text inputs are. The token cost of a given image is determined by two factors: its size, and the detail option on each image_url block. All images with detail: low cost 85 tokens each. detail: high images are first scaled to fit within a 2048 x 2048 square, maintaining their aspect ratio. Then, they are scaled such that the shortest side of the image is 768px long. Finally, we count how many 512px squares the image consists of. Each of those squares costs **170 tokens**. Another **85 tokens** are always added to the final total.

Here are some examples demonstrating the above.

- A 1024 x 1024 square image in detail: high mode costs 765 tokens
 - 1024 is less than 2048, so there is no initial resize.
 - The shortest side is 1024, so we scale the image down to 768 x 768.
 - 4 512px square tiles are needed to represent the image, so the final token cost is 170 * 4 + 85 = 765.
- A 2048 x 4096 image in detail: high mode costs 1105 tokens
 - We scale down the image to 1024 x 2048 to fit within the 2048 square.
 - The shortest side is 1024, so we further scale down to 768 x 1536.
 - \circ 6 512px tiles are needed, so the final token cost is 170 * 6 + 85 = 1105.
- A 4096 x 8192 image in detail: low most costs 85 tokens
 - Regardless of input size, low detail images are a fixed cost.

FAQ

Can I fine-tune the image capabilities in gpt-4?

No, we do not support fine-tuning the image capabilities of gpt-4 at this time.

Can I use gpt-4 to generate images?

No, you can use dall-e-3 to generate images and gpt-4-vision-preview to understand images.

What type of files can I upload?

We currently support PNG (.png), JPEG (.jpeg and .jpg), WEBP (.webp), and non-animated GIF (.gif).

Is there a limit to the size of the image I can upload?

Yes, we restrict image uploads to 20MB per image.

Can I delete an image I uploaded?

No, we will delete the image for you automatically after it has been processed by the model.

Where can I learn more about the considerations of GPT-4 with Vision?

You can find details about our evaluations, preparation, and mitigation work in the <u>GPT-4 with Vision system card</u>.

We have further implemented a system to block the submission of CAPTCHAs.

How do rate limits for GPT-4 with Vision work?

We process images at the token level, so each image we process counts towards your tokens per minute (TPM) limit. See the calculating costs section for details on the formula used to determine token count per image.

Can GPT-4 with Vision understand image metadata?

No, the model does not receive image metadata.

What happens if my image is unclear?

If an image is ambiguous or unclear, the model will do its best to interpret it. However, the results may be less accurate. A good rule of thumb is that if an average human cannot see the info in an image at the resolutions used in low/high res mode, then the model cannot either.

From https://platform.openai.com/docs/guides/vision

Text to speech

Learn how to turn text into lifelike spoken audio

Introduction

The Audio API provides a text to speech endpoint, speech, based on our TTS (text-to-speech) model. It comes with 6 build in voices and can be used to:

- Narrate a written blog post
- Produce spoken audio in multiple languages
- Give real time audio output using streaming

Here is an example of the alloy voice:

Please note that our <u>Usage Policies</u> require you to provide a clear disclosure to end users that the TTS voice they are hearing is AI-generated and not a human voice.

Quick start

The speech endpoint takes in three key inputs: the model name, the text that should be turned into audio, and the voice to be used for the audio generation. A simple request would look like the following:

```
Speak this text
python
Copy
2
3
4
5
6
7
8
9
10
11
from pathlib import Path
from openai import OpenAI
client = OpenAI()
speech_file_path = Path(__file__).parent / "speech.mp3"
response = client.audio.speech.create(
 model="tts-1",
 voice="alloy",
 input="Today is a wonderful day to build something people love!"
)
response.stream to file(speech file path)
```

By default, the endpoint will output a MP3 file of the spoken audio but it can also be configured to output any of our <u>supported formats</u>.

Audio quality

For real-time applications, the standard tts-1 model provides the lowest latency but at a lower quality than the tts-1-hd model. Due to the way the audio is generated, tts-1 is likely to generate content that has more static in certain situations than tts-1-hd. In some cases, the audio may not have noticeable differences depending on your listening device and the individual person.

Voice options

Experiment with different voices (alloy, echo, fable, onyx, nova, and shimmer) to find one that matches your desired tone and audience.

Alloy

Echo

<u>Fable</u>

Onyx

<u>Nova</u>

Shimmer

Supported output formats

The default response format is "mp3", but other formats like "opus", "aac", or "flac" are available.

- Opus: For internet streaming and communication, low latency.
- AAC: For digital audio compression, preferred by YouTube, Android, iOS.
- FLAC: For lossless audio compression, favored by audio enthusiasts for archiving.

Streaming real time audio

The Speech API provides support for real time audio streaming using <u>chunk transfer</u> <u>encoding</u>. This means that the audio is able to be played before the full file has been generated and made accessible.

```
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12
13
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21
22
23
24
25
26
27
import io
from openai import OpenAI
from pydub import AudioSegment
from pydub.playback import play
client = OpenAI()
def stream_and_play(text):
 response = client.audio.speech.create(
  model="tts-1",
  voice="alloy",
  input=text,
 )
```

```
# Convert the binary response content to a byte stream
byte_stream = io.BytesIO(response.content)
# Read the audio data from the byte stream
audio = AudioSegment.from_file(byte_stream, format="mp3")
# Play the audio
play(audio)

if __name__ == "__main__":
text = input("Enter text: ")
stream_and_play(text)
```

FAQ

How can I control the emotional range of the generated audio?

There is no direct mechanism to control the emotional output of the audio generated. Certain factors may influence the output audio like capitalization or grammar but our internal tests with these have yielded mixed results.

Can I create a custom copy of my own voice?

No, this is not something we support.

Do I own the outputted audio files?

Yes, like with all outputs from our API, the person who created them owns the output. You are still required to inform end users that they are hearing audio generated by AI and not a real person talking to them.

From < https://platform.openai.com/docs/guides/text-to-speech>

Speech to text

Learn how to turn audio into text

Introduction

The Audio API provides two speech to text endpoints, transcriptions and translations, based on our state-of-the-art open source large-v2 Whisper model. They can be used to:

- Transcribe audio into whatever language the audio is in.
- Translate and transcribe the audio into english.

File uploads are currently limited to 25 MB and the following input file types are supported: mp3, mp4, mpeg, mpga, m4a, wav, and webm.

Quickstart

Transcriptions

The transcriptions API takes as input the audio file you want to transcribe and the desired output file format for the transcription of the audio. We currently support multiple input and output file formats.

```
Transcribe audio

python

Copy
1
2
3
4
5
6
7
8
from openai import OpenAl
client = OpenAl()
audio_file= open("/path/to/file/audio.mp3", "rb")
transcript = client.audio.transcriptions.create(
    model="whisper-1",
    file=audio_file
)
```

By default, the response type will be json with the raw text included.

{ "text": "Imagine the wildest idea that you've ever had, and you're curious about how it might scale to something that's a 100, a 1,000 times bigger. }

The Audio API also allows you to set additional parameters in a request. For example, if you want to set the response_format as text, your request would look like the following:

```
Additional options
python
Copy
2
3
4
5
6
7
8
from openai import OpenAI
client = OpenAI()
audio_file = open("speech.mp3", "rb")
transcript = client.audio.transcriptions.create(
 model="whisper-1",
 file=audio_file,
 response_format="text"
```

The API Reference includes the full list of available parameters.

Translations

The translations API takes as input the audio file in any of the supported languages and transcribes, if necessary, the audio into English. This differs from our /Transcriptions endpoint since the output is not in the original input language and is instead translated to English text.

Translate audio

```
python
Copy
1
2
3
4
5
6
7
8
from openai import OpenAl
client = OpenAl()
audio_file= open("/path/to/file/german.mp3", "rb")
transcript = client.audio.translations.create(
    model="whisper-1",
    file=audio_file
)
```

In this case, the inputted audio was german and the outputted text looks like:

Hello, my name is Wolfgang and I come from Germany. Where are you heading today? We only support translation into english at this time.

Supported languages

We currently <u>support the following languages</u> through both the transcriptions and translations endpoint:

Afrikaans, Arabic, Armenian, Azerbaijani, Belarusian, Bosnian, Bulgarian, Catalan, Chinese, Croatian, Czech, Danish, Dutch, English, Estonian, Finnish, French, Galician, German, Greek, Hebrew, Hindi, Hungarian, Icelandic, Indonesian, Italian, Japanese, Kannada, Kazakh, Korean, Latvian, Lithuanian, Macedonian, Malay, Marathi, Maori, Nepali, Norwegian, Persian, Polish, Portuguese, Romanian, Russian, Serbian, Slovak, Slovenian, Spanish, Swahili, Swedish, Tagalog, Tamil, Thai, Turkish, Ukrainian, Urdu, Vietnamese, and Welsh.

While the underlying model was trained on 98 languages, we only list the languages that exceeded <50% word error rate (WER) which is an industry standard benchmark for speech to text model accuracy. The model will return results for languages not listed above but the quality will be low.

Longer inputs

By default, the Whisper API only supports files that are less than 25 MB. If you have an audio file that is longer than that, you will need to break it up into chunks of 25 MB's or less or used a compressed audio format. To get the best performance, we suggest that you avoid breaking the audio up mid-sentence as this may cause some context to be lost.

One way to handle this is to use the <u>PyDub open source Python package</u> to split the audio:

- 1
- 3
- 4

```
5
6
7
8
9
10
from pydub import AudioSegment
song = AudioSegment.from_mp3("good_morning.mp3")
# PyDub handles time in milliseconds
ten_minutes = 10 * 60 * 1000
first_10_minutes = song[:ten_minutes]
first_10_minutes.export("good_morning_10.mp3", format="mp3")
OpenAI makes no guarantees about the usability or security of 3rd party software like
PyDub.
```

Prompting

You can use a <u>prompt</u> to improve the quality of the transcripts generated by the Whisper API. The model will try to match the style of the prompt, so it will be more likely to use capitalization and punctuation if the prompt does too. However, the current prompting system is much more limited than our other language models and only provides limited control over the generated audio. Here are some examples of how prompting can help in different scenarios:

- 1. Prompts can be very helpful for correcting specific words or acronyms that the model often misrecognizes in the audio. For example, the following prompt improves the transcription of the words DALL-E and GPT-3, which were previously written as "GDP 3" and "DALI": "The transcript is about OpenAI which makes technology like DALL-E, GPT-3, and ChatGPT with the hope of one day building an AGI system that benefits all of humanity"
- 2. To preserve the context of a file that was split into segments, you can prompt the model with the transcript of the preceding segment. This will make the transcript more accurate, as the model will use the relevant information from the previous audio. The model will only consider the final 224 tokens of the prompt and ignore anything earlier. For multilingual inputs, Whisper uses a custom tokenizer. For English only inputs, it uses the standard GPT-2 tokenizer which are both accessible through the open source Whisper Python package.
- Sometimes the model might skip punctuation in the transcript. You can avoid this by using a simple prompt that includes punctuation: "Hello, welcome to my lecture."
- 4. The model may also leave out common filler words in the audio. If you want to keep the filler words in your transcript, you can use a prompt that contains them: "Umm, let me think like, hmm... Okay, here's what I'm, like, thinking."
- 5. Some languages can be written in different ways, such as simplified or traditional Chinese. The model might not always use the writing style that you want for your transcript by default. You can improve this by using a prompt in your preferred writing style.

Improving reliability

As we explored in the prompting section, one of the most common challenges faced when using Whisper is the model often does not recognize uncommon words or

acronyms. To address this, we have highlighted different techniques which improve the reliability of Whisper in these cases:

Using the prompt parameter

The first method involves using the optional prompt parameter to pass a dictionary of the correct spellings.

Since it wasn't trained using instruction-following techniques, Whisper operates more like a base GPT model. It's important to keep in mind that Whisper only considers the first 244 tokens of the prompt.

transcribe(filepath, prompt="ZyntriQix, Digique Plus, CynapseFive, VortiQore V8, EchoNix Array, OrbitalLink Seven, DigiFractal Matrix, PULSE, RAPT, B.R.I.C.K., Q.U.A.R.T.Z., F.L.I.N.T.")

While it will increase reliability, this technique is limited to only 244 characters so your list of SKUs would need to be relatively small in order for this to be a scalable solution.

Collapse

Post-processing with GPT-4

The second method involves a post-processing step using GPT-4 or GPT-3.5-Turbo.

We start by providing instructions for GPT-4 through the system_prompt variable. Similar to what we did with the prompt parameter earlier, we can define our company and product names.

system_prompt = "You are a helpful assistant for the company ZyntriQix. Your task is to correct any spelling discrepancies in the transcribed text. Make sure that the names of the following products are spelled correctly: ZyntriQix, Digique Plus, CynapseFive, VortiQore V8, EchoNix Array, OrbitalLink Seven, DigiFractal Matrix, PULSE, RAPT, B.R.I.C.K., Q.U.A.R.T.Z., F.L.I.N.T. Only add necessary punctuation such as periods, commas, and capitalization, and use only the context provided."

def generate_corrected_transcript(temperature, system_prompt, audio_file):

response = client.chat.completions.create(

return response['choices'][0]['message']['content']

corrected_text = generate_corrected_transcript(0, system_prompt, fake_company_filepath) If you try this on your own audio file, you can see that GPT-4 manages to correct many misspellings in the transcript. Due to its larger context window, this method might be more scalable than using Whisper's prompt parameter and is more reliable since GPT-4 can be instructed and guided in ways that aren't possible with Whisper given the lack of instruction following.

From < https://platform.openai.com/docs/guides/speech-to-text>

Moderation

Overview

The <u>moderations</u> endpoint is a tool you can use to check whether content complies with OpenAl's <u>usage policies</u>. Developers can thus identify content that our usage policies prohibits and take action, for instance by filtering it.

The models classifies the following categories:

| CATEGORY | DESCRIPTION |
|----------------------------|---|
| hate | Content that expresses, incites, or promotes hate based on race, gender, ethnicity, religion, nationality, sexual orientation, disability status, or caste. Hateful content aimed at non-protected groups (e.g., chess players) is harrassment. |
| hate/threat ening | Hateful content that also includes violence or serious harm towards the targeted group based on race, gender, ethnicity, religion, nationality, sexual orientation, disability status, or caste. |
| harassment | Content that expresses, incites, or promotes harassing language towards any target. |
| harassment /threatening | Harassment content that also includes violence or serious harm towards any target. |
| self-harm | Content that promotes, encourages, or depicts acts of self-harm, such as |

suicide, cutting, and eating disorders.

self- Content where the speaker expresses that they are engaging or intend to

harm/intent engage in acts of self-harm, such as suicide, cutting, and eating

disorders.

self- Content that encourages performing acts of self-harm, such as suicide, harm/instru cutting, and eating disorders, or that gives instructions or advice on how

ctions to commit such acts.

sexual Content meant to arouse sexual excitement, such as the description of

sexual activity, or that promotes sexual services (excluding sex education

and wellness).

sexual/min Sexual content that includes an individual who is under 18 years old.

ors

violence Content that depicts death, violence, or physical injury.

violence/gr Content that depicts death, violence, or physical injury in graphic detail.

aphic

The moderation endpoint is free to use when monitoring the inputs and outputs of OpenAI APIs. We currently disallow other use cases. Accuracy may be lower on longer pieces of text. For higher accuracy, try splitting long pieces of text into smaller chunks each less than 2,000 characters.

We are continuously working to improve the accuracy of our classifier. Our support for non-English languages is currently limited.

Quickstart

To obtain a classification for a piece of text, make a request to the <u>moderation</u> endpoint as demonstrated in the following code snippets:

Example: Getting moderations

curl
Copy
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curl https://api.openai.com/v1/moderations \
-X POST \
-H "Content-Type: application/json" \
-H "Authorization: Bearer \$OPENAI_API_KEY" \
-d '{"input": "Sample text goes here"}'

Below is an example output of the endpoint. It returns the following fields:

- flagged: Set to true if the model classifies the content as violating OpenAl's usage policies, false otherwise.
- categories: Contains a dictionary of per-category binary usage policies violation flags. For each category, the value is true if the model flags the corresponding category as violated, false otherwise.
- category_scores: Contains a dictionary of per-category raw scores output by the model, denoting the model's confidence that the input violates the OpenAl's policy for the category. The value is between 0 and 1, where higher values denote higher confidence. The scores should not be interpreted as probabilities.

```
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24
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26
27
28
29
30
31
32
33
34
35
 "id": "modr-XXXXX",
 "model": "text-moderation-005",
 "results": [
   "flagged": true,
   "categories": {
     "sexual": false,
     "hate": false,
     "harassment": false,
     "self-harm": false,
     "sexual/minors": false,
     "hate/threatening": false,
     "violence/graphic": false,
     "self-harm/intent": false,
     "self-harm/instructions": false,
     "harassment/threatening": true,
     "violence": true,
    "category_scores": {
     "sexual": 1.2282071e-06,
     "hate": 0.010696256,
```

```
"harassment": 0.29842457,
"self-harm": 1.5236925e-08,
"sexual/minors": 5.7246268e-08,
"hate/threatening": 0.0060676364,
"violence/graphic": 4.435014e-06,
"self-harm/intent": 8.098441e-10,
"self-harm/instructions": 2.8498655e-11,
"harassment/threatening": 0.63055265,
"violence": 0.99011886,
}
}
```

OpenAI will continuously upgrade the moderation endpoint's underlying model. Therefore, custom policies that rely on category_scores may need recalibration over time.

From < https://platform.openai.com/docs/guides/moderation/quickstart>

Assistants API

Beta

The Assistants API allows you to build AI assistants within your own applications. An Assistant has instructions and can leverage models, tools, and knowledge to respond to user queries. The Assistants API currently supports three types of tools: Code Interpreter, Retrieval, and Function calling. In the future, we plan to release more OpenAI-built tools, and allow you to provide your own tools on our platform.

At a high level, a typical integration of the Assistants API has the following flow:

- 1. Create an <u>Assistant</u> in the API by defining it custom instructions and picking a model. If helpful, enable tools like Code Interpreter, Retrieval, and Function calling.
- 2. Create a Thread when a user starts a conversation.
- 3. Add Messages to the Thread as the user ask questions.
- 4. Run the Assistant on the Thread to trigger responses. This automatically calls the relevant tools.

The Assistants API is in **beta** and we are actively working on adding more functionality. Share your feedback in our <u>Developer Forum!</u>

This starter guide walks through the key steps to create and run an Assistant that uses Code Interpreter.

Step 1: Create an Assistant

An Assistant represents an entity that can be configured to respond to users' Messages using several parameters like:

- Instructions: how the Assistant and model should behave or respond
- Model: you can specify any GPT-3.5 or GPT-4 models, including fine-tuned models. The Retrieval tool requires gpt-3.5-turbo-1106 and gpt-4-1106-preview models.
- Tools: the API supports Code Interpreter and REtrieval that are built and hosted by OpenAI.
- Functions: the API allows you to define custom function signatures, with similar behavior as our <u>function calling</u> feature.

In this example, we're creating an Assistant that is a personal math tutor, with the Code Interpreter tool enabled:

Calls to the Assistants API require that you pass a Beta header. This is handled automatically if you're using OpenAI's official Python and Node.js SDKs.

```
OpenAI-Beta: assistants=v1

python
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assistant = client.beta.assistants.create(
    name="Math Tutor",
    instructions="You are a personal math tutor. Write and run code to answer math questions.",
    tools=[{"type": "code_interpreter"}],
    model="gpt-4-1106-preview"
)
```

Step 2: Create a Thread

A Thread represents a conversation. We recommend creating one Thread per user as soon as the user initiates the conversation. Pass any user-specific context and files in this thread by creating Messages.

```
python
Copy
thread = client.beta.threads.create()
```

Threads don't have a size limit. You can pass as many Messages as you want to a Thread. The API will ensure that requests to the model fit within the maximum context window, using relevant optimization techniques such as truncation.

Step 3: Add a Message to a Thread

A Message contains the user's text, and optionally, any <u>files</u> that the user uploads. Image files aren't supported today, but we plan to add support for them in the coming months.

```
python
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message = client.beta.threads.messages.create(
    thread_id=thread.id,
    role="user",
    content="I need to solve the equation `3x + 11 = 14`. Can you help me?"
)
```

Now if you <u>list Messages in Thread</u>, you will see that this message is added to the thread on creation:

```
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```

```
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6
7
8
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10
11
12
13
14
15
16
17
 "object": "list",
 "data": [
   "created_at": 1696995451,
   "id": "msg_4rb1Skx3XgQZEe4PHVRFQhr0",
   "object": "thread.message",
   "thread_id": "thread_34p0sfdas0823smfv",
   "role": "user",
   "content": [{
     "type": "text",
     "text": {
      "value": "I need to solve the equation 3x + 11 = 14. Can you help me?",
      "annotations": []
     }
      }],
```

Step 4: Run the Assistant

For the Assistant to respond to the user message, you need to create a Run. This makes the Assistant read the Thread and decide whether to call tools or simply use the model to best answer the user query. As the run progresses, the assistant appends Messages to the thread with the role="assistant".

You can optionally pass additional instructions to the Assistant while creating the Run:

```
python
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run = client.beta.threads.runs.create(
    thread_id=thread.id,
    assistant_id=assistant.id,
    instructions="Please address the user as Jane Doe. The user has a premium account."
)
```

Step 5: Display the Assistant's Response

This creates a Run in a queued status. You can periodically retrieve the Run to check on its <u>status</u> to see if it has moved to completed.

```
python
Copy
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```

```
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run = client.beta.threads.runs.retrieve(
    thread_id=thread.id,
    run_id=run.id
)
Once the Run completes, you can retrieve the Messages added by the Assistant to the Thread.
```

```
python
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3
messages = client.beta.threads.messages.list(
    thread_id=thread.id
)
```

And finally, display them to the user! During this Run, the Assistant added two new Messages to the Thread.

```
user I need to solve the equation 3x + 11 = 14. Can you help me?

assista Certainly, Jane Doe. To solve the equation (3x + 11 = 14) for (x), you'll want to isolate (x) on one side of the equation. Here's how you can do that:

1. Subtract 11 from both sides of the equation to get (3x = 3).

2. Then, divide both sides by 3 to solve for (x).

Let me calculate the value of (x) for you.

assista The solution to the equation (3x + 11 = 14) is (x = 1).
```

You can also retrieve the <u>Run Steps</u> of this Run if you'd like to explore or display the inner workings of the Assistant and its tools.

Next

- 1. Dive deeper into How Assistants work
- 2. Learn more about <u>Tools</u>

From < https://platform.openai.com/docs/assistants/overview>

How Assistants work

<u>Beta</u>

The Assistants API is designed to help developers build powerful AI assistants capable of performing a variety of tasks.

The Assistants API is in **beta** and we are actively working on adding more functionality. Share your feedback in our <u>Developer Forum!</u>

- Assistants can call OpenAl's <u>models</u> with specific instructions to tune their personality and capabilities.
- 2. Assistants can access multiple tools in parallel. These can be both OpenAl-

- hosted tools like <u>Code interpreter</u> and <u>Knowledge retrieval</u> or tools you build / host (via Function calling).
- 3. Assistants can access **persistent Threads**. Threads simplify AI application development by storing message history and truncating it when the conversation gets too long for the model's context length. You create a Thread once, and simply append Messages to it as your users reply.
- 4. Assistants can access <u>Files</u> in several formats either as part of their creation or as part of Threads between Assistants and users. When using tools, Assistants can also create files (e.g., images, spreadsheets, etc) and cite files they reference in the Messages they create.

Objects

OBJECT WHAT IT REPRESENTS

Assista Purpose-built AI that uses OpenAI's models and calls tools

nt

Thread A conversation session between an Assistant and a user. Threads store Messages and automatically handle truncation to fit content into a model's context.

Messa A message created by an Assistant or a user. Messages can include text, ge images, and other files. Messages stored as a list on the Thread.

Run An invocation of an Assistant on a Thread. The Assistant uses it's configuration and the Thread's Messages to perform tasks by calling models and tools. As part of a Run, the Assistant appends Messages to the Thread.

Run A detailed list of steps the Assistant took as part of a Run. An Assistant can call tools or create Messages during it's run. Examining Run Steps allows you to introspect how the Assistant is getting to it's final results.

Creating Assistants

We recommend using OpenAl's <u>latest models</u> with the Assistants API for best results and maximum compatibility with tools.

To get started, creating an Assistant only requires specifying the model to use. But you can further customize the behavior of the Assistant:

- Use the instructions parameter to guide the personality of the Assistant and define it's goals. Instructions are similar to system messages in the Chat Completions API.
- 2. Use the tools parameter to give the Assistant access to up to 128 tools. You can give it access to OpenAl-hosted tools like code_interpreter and retrieval, or call a third-party tools via a function calling.
- 3. Use the file_ids parameter to give the tools like code_interpreter and retrieval access to files. Files are uploaded using the File <u>upload endpoint</u> and must have the purpose set to assistants to be used with this API.

For example, to create an Assistant that can create data visualization based on a .csv file, first upload a file.

python Copy 1

2

3

```
file = client.files.create(
 file=open("speech.py", "rb"),
 purpose='assistants'
And then create the Assistant with the uploaded file.
python
Copy
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6
assistant = client.beta.assistants.create(
 name="Data visualizer",
 description="You are great at creating beautiful data visualizations. You analyze data present in .csv files,
understand trends, and come up with data visualizations relevant to those trends. You also share a brief text
summary of the trends observed.",
 model="gpt-4-1106-preview",
 tools=[{"type": "code interpreter"}],
 file_ids=[file.id]
```

You can attach a maximum of 20 files per Assistant, and they can be at most 512 MB each. In addition, the size of all the files uploaded by your organization should not exceed 100GB. You can request an increase in this storage limit using our <u>help center</u>.

You can also use the <u>AssistantFile</u> object to create, delete, or view associations between Assistant and File objects. Note that deleting an AssistantFile doesn't delete the original File object, it simply deletes the association between that File and the Assistant. To delete a File, use the File delete endpoint instead.

Managing Threads and Messages

Threads and Messages represent a conversation session between an Assistant and a user. There is no limit to the number of Messages you can store in a Thread. Once the size of the Messages exceeds the context window of the model, the Thread smartly truncates them to fit. You can create a Thread with an initial list of Messages like this:

```
python
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8
thread = client.beta.threads.create(
 messages=[
    "role": "user",
    "content": "Create 3 data visualizations based on the trends in this file.",
    "file_ids": [file.id]
  }
]
```

Messages can contain text, images, or files. At the moment, user-created Messages cannot contain image files but we plan to add support for this in the future.

Message annotations

Messages created by Assistants may contain <u>annotations</u> within the content array of the object. Annotations provide information around how you should annotate the text in the Message.

There are two types of Annotations:

- 1. file_citation: File citations are created by the <u>retrieval</u> tool and define references to a specific quote in a specific file that was uploaded and used by the Assistant to generate the response.
- 2. file_path: File path annotations are created by the <u>code_interpreter</u> tool and contain references to the files generated by the tool.

When annotations are present in the Message object, you'll see illegible model-generated substrings in the text that you should replace with the annotations. These strings may look something like 【13+source】 or sandbox:/mnt/data/file.csv. Here's an example python code snippet that replaces these strings with information present in the annotations.

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# Retrieve the message object
message = client.beta.threads.messages.retrieve(
thread_id="...",
 message_id="..."
```

```
# Extract the message content
message_content = message.content[0].text
annotations = message_content.annotations
citations = []
# Iterate over the annotations and add footnotes
for index, annotation in enumerate(annotations):
  # Replace the text with a footnote
  message_content.value = message_content.value.replace(annotation.text, f' [{index}]')
  # Gather citations based on annotation attributes
  if (file citation := getattr(annotation, 'file citation', None)):
     cited_file = client.files.retrieve(file_citation.file_id)
     citations.append(f'[{index}] {file_citation.quote} from {cited_file.filename}')
  elif (file path := getattr(annotation, 'file path', None)):
     cited file = client.files.retrieve(file path.file id)
     citations.append(f'[{index}] Click <here> to download {cited_file.filename}')
     # Note: File download functionality not implemented above for brevity
# Add footnotes to the end of the message before displaying to user
message content.value += '\n' + '\n'.join(citations)
```

Runs and Run Steps

When you have all the context you need from your user in the Thread, you can run the Thread with an Assistant of your choice.

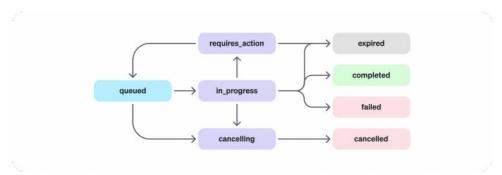
```
python
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run = client.beta.threads.runs.create(
    thread_id=thread.id,
    assistant_id=assistant.id
)
By default, a Run will use the model and tools configuration specified in Assistant object,
but you can override most of these when creating the Run for added flexibility:
```

```
python
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7
run = client.beta.threads.runs.create(
    thread_id=thread.id,
    assistant_id=assistant.id,
    model="gpt-4-1106-preview",
    instructions="additional instructions",
    tools=[{"type": "code_interpreter"}, {"type": "retrieval"}]
)
```

Note: file_ids associated with the Assistant cannot be overridden during Run creation. You must use the modify Assistant endpoint to do this.

Run lifecycle

Run objects can have multiple statuses.



STATUS DEFINITION

queued When Runs are first created or when you complete the required_action, they are moved to a queued status. They should almost immediately move to in_progress.

in_progr While in_progress, the Assistant uses the model and tools to perform steps. ess You can view progress being made by the Run by examining the Run Steps.

complet The Run successfully completed! You can now view all Messages the Assistant added to the Thread, and all the steps the Run took. You can also continue the conversation by adding more user Messages to the Thread and creating another Run.

requires When using the <u>Function calling</u> tool, the Run will move to a required_action state once the model determines the names and arguments of the functions to be called. You must then run those functions and <u>submit the outputs</u> before the run proceeds. If the outputs are not provided before the expires_at timestamp passes (roughly 10 mins past creation), the run will move to an expired status.

expired This happens when the function calling outputs were not submitted before expires_at and the run expires. Additionally, if the runs take too long to execute and go beyond the time stated in expires_at, our systems will expire the run.

cancelli You can attempt to cancel an in_progress run using the <u>Cancel</u>
ng <u>Run</u> endpoint. Once the attempt to cancel succeeds, status of the Run moves
to cancelled. Cancellation is attempted but not guaranteed.

cancelle Run was successfully cancelled.

failed You can view the reason for the failure by looking at the last_error object in the Run. The timestamp for the failure will be recorded under failed_at.

Polling for updates

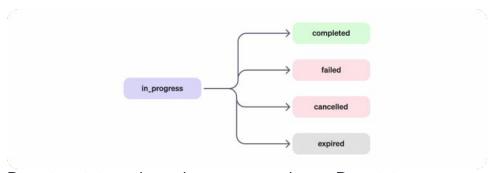
In order to keep the status of your run up to date, you will have to periodically <u>retrieve</u> the Run object. You can check the status of the run each time you retrieve the object to determine what your application should do next. We plan to add support for streaming to make this simpler in the near future.

Thread locks

When a Run is in_progress and not in a terminal state, the Thread is locked. This means that:

- New Messages cannot be added to the Thread.
- New Runs cannot be created on the Thread.

Run steps



Run step statuses have the same meaning as Run statuses.

Most of the interesting detail in the Run Step object lives in the step_details field. There can be two types of step details:

- message_creation: This Run Step is created when the Assistant creates a Message on the Thread.
- 2. tool_calls: This Run Step is created when the Assistant calls a tool. Details around this are covered in the relevant sections of the Tools guide.

Limitations

During this beta, there are several known limitations we are looking to address in the coming weeks and months. We will publish a changelog on this page when we add support for additional functionality.

- Support for streaming output (including Messages and Run Steps).
- Support for notifications to share object status updates without the need for polling.
- Support for DALL-E as a tool.
- Support for user message creation with images.

Next

Learn more about Tools

From https://platform.openai.com/docs/assistants/how-it-works/limitations

Tools

Beta

Give Assistants access to OpenAI-hosted tools like Code Interpreter and Knowledge Retrieval, or build your own tools using Function calling.

The Assistants API is in **beta** and we are actively working on adding more functionality. Share your feedback in our <u>Developer Forum!</u>

Code Interpreter

Code Interpreter allows the Assistants API to write and run Python code in a sandboxed execution

environment. This tool can process files with diverse data and formatting, and generate files with data and images of graphs. Code Interpreter allows your Assistant to run code iteratively to solve challenging code and math problems. When your Assistant writes code that fails to run, it can iterate on this code by attempting to run different code until the code execution succeeds.

Enabling Code Interpreter

Pass the code_interpreterin the tools parameter of the Assistant object to enable Code Interpreter:

```
python
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1
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5
assistant = client.beta.assistants.create(
    instructions="You are a personal math tutor. When asked a math question, write and run code to answer the question.",
    model="gpt-4-1106-preview",
    tools=[{"type": "code_interpreter"}]
)
```

The model then decides when to invoke Code Interpreter in a Run based on the nature of the user request. This behavior can be promoted by prompting in the Assistant's instructions (e.g., "write code to solve this problem").

Passing files to Code Interpreter

Code Interpreter can parse data from files. This is useful when you want to provide a large volume of data to the Assistant or allow your users to upload their own files for analysis.

Files that are passed at the Assistant level are accessible by all Runs with this Assistant:

```
python
Copy
2
3
4
5
6
7
8
9
10
11
12
13
# Upload a file with an "assistants" purpose
file = client.files.create(
 file=open("speech.py", "rb"),
 purpose='assistants'
# Create an assistant using the file ID
assistant = client.beta.assistants.create(
 instructions="You are a personal math tutor. When asked a math question, write and run code to answer
the question.",
 model="gpt-4-1106-preview",
 tools=[{"type": "code_interpreter"}],
 file_ids=[file.id]
```

Files can also be passed at the Thread level. These files are only accessible in the specific Thread. Upload the File using the File upload endpoint and then pass the File ID as part of the Message creation request:

Files have a maximum size of 512 MB. Code Interpreter supports a variety of file formats including .csv, .pdf, .json and many more. More details on the file extensions (and their corresponding MIME-types) supported can be found in the <u>Supported files</u> section below.

Reading images and files generated by Code Interpreter

Code Interpreter in the API also outputs files, such as generating image diagrams, CSVs, and PDFs. There are two types of files that are generated:

- 1. Images
- 2. Data files (e.g. a csv file with data generated by the Assistant)

When Code Interpreter generates an image, you can look up and download this file in the file_id field of the Assistant Message response:

```
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16
  "id": "msg_OHGpsFRGFYmz69MM1u8KYCwf",
  "object": "thread.message",
```

```
"created_at": 1698964262,
  "thread_id": "thread_uqorHcTs46BZhYMyPn6Mg5gW",
  "role": "assistant",
  "content": [
   "type": "image_file",
   "image_file": {
     "file_id": "file-WsgZPYWAauPuW4uvcgNUGcb"
  }
 ]
 # ...
The file content can then be downloaded by passing the file ID to the Files API:
python
Copy
content = client.files.retrieve_content(file.id)
When Code Interpreter references a file path (e.g., "Download this csv file"), file paths are listed as
annotations. You can convert these annotations into links to download the file:
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19
20
21
22
 "id": "msg_3jylh3DgunZSNMCOORflDyih",
 "object": "thread.message",
 "created at": 1699073585.
 "thread_id": "thread_ZRvNTPOoYVGssUZr3G8cRRzE",
 "role": "assistant",
 "content": [
    "type": "text",
     "value": "The rows of the CSV file have been shuffled and saved to a new CSV file. You
can download the shuffled CSV file from the following link:\n\n[Download Shuffled CSV File]
(sandbox:/mnt/data/shuffled_file.csv)",
```

Input and output logs of Code Interpreter

By listing the steps of a Run that called Code Interpreter, you can inspect the code input and outputs logs of Code Interpreter:

```
python
Copy
2
3
run_steps = client.beta.threads.runs.steps.list(
thread_id=thread.id,
 run_id=run.id
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
 "object": "list",
 "data": [
   "id": "step_DQfPq3JPu8hRKW0ctAraWC9s",
    "object": "assistant.run.step",
   "type": "tool_calls",
```

Knowledge Retrieval

Retrieval augments the Assistant with knowledge from outside its model, such as proprietary product information or documents provided by your users. Once a file is uploaded and passed to the Assistant, OpenAI will automatically chunk your documents, index and store the embeddings, and implement vector search to retrieve relevant content to answer user queries.

Enabling Retrieval

Pass the retrieval in the tools parameter of the Assistant to enable Retrieval:

```
python
Copy
1
2
3
4
5
assistant = client.beta.assistants.create(
  instructions="You are a customer support chatbot. Use your knowledge base to best respond to customer queries.",
  model="gpt-4-1106-preview",
  tools=[{"type": "retrieval"}]
)
```

How it works

The model then decides when to retrieve content based on the user Messages. The Assistants API automatically chooses between two retrieval techniques:

- 1. it either passes the file content in the prompt for short documents, or
- 2. performs a vector search for longer documents

Retrieval currently optimizes for quality by adding all relevant content to the context of model calls. We plan to introduce other retrieval strategies to enable developers to choose a different tradeoff between retrieval quality and model usage cost.

Uploading files for retrieval

Similar to Code Interpreter, files can be passed at the Assistant-level or at the Thread-level

python

```
2
3
4
5
6
7
8
9
10
11
12
13
# Upload a file with an "assistants" purpose
file = client.files.create(
 file=open("knowledge.pdf", "rb"),
 purpose='assistants'
# Add the file to the assistant
assistant = client.beta.assistants.create(
 instructions="You are a customer support chatbot. Use your knowledge base to best respond to customer
queries.",
 model="gpt-4-1106-preview",
 tools=[{"type": "retrieval"}],
 file_ids=[file.id]
Files can also be added to a Message in a Thread. These files are only accessible within this specific
thread. After having uploaded a file, you can pass the ID of this File when creating the Message:
python
Copy
2
3
4
5
message = client.beta.threads.messages.create(
thread_id=thread.id,
 role="user",
 content="I can't find in the PDF manual how to turn off this device.",
 file_ids=[file.id]
Maximum file size is 512MB. Retrieval supports a variety of file formats including .pdf, .md, .docx and
many more. More details on the file extensions (and their corresponding MIME-types) supported can be
found in the <u>Supported files</u> section below.
Deleting files
To remove a file from the assistant, you can detach the file from the assistant:
python
Copy
1
2
3
file_deletion_status = client.beta.assistants.files.delete(
```

Copy

assistant_id=assistant.id,

```
file_id=file.id
```

Detaching the file from the assistant removes the file from the retrieval index as well.

File citations

When Code Interpreter outputs file paths in a Message, you can convert them to corresponding file downloads using the annotations field. See the <u>Annotations section</u> for an example of how to do this.

```
1
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11
12
13
14
15
16
17
18
19
20
21
22
23
24
25
26
27
28
29
30
  "id": "msg_3jylh3DgunZSNMCOORflDyih",
  "object": "thread.message",
  "created_at": 1699073585,
  "thread id": "thread ZRvNTPOoYVGssUZr3G8cRRzE",
  "role" "assistant".
  "content": [
    "type": "text",
      "value": "The rows of the CSV file have been shuffled and saved to a new CSV file. You
can download the shuffled CSV file from the following link:\n\n[Download Shuffled CSV File]
(sandbox:/mnt/data/shuffled_file.csv)",
      "annotations": [
        "type": "file_path",
        "text": "sandbox:/mnt/data/shuffled_file.csv",
```

```
"start_index": 167,
    "end_index": 202,
    "file_path": {
        "file_id": "file-oSgJAzAnnQkVB3u7yCoE9CBe"
        }
     }
     ]
     }
     ]
     file_ids": [
        "file_ids": [
        "file-oSgJAzAnnQkVB3u7yCoE9CBe"
     ],
     ...
},
```

Function calling

Similar to the Chat Completions API, the Assistants API supports function calling. Function calling allows you to describe functions to the Assistants and have it intelligently return the functions that need to be called along with their arguments. The Assistants API will pause execution during a Run when it invokes functions, and you can supply the results of the function call back to continue the Run execution.

Defining functions

First, define your functions when creating an Assistant:

```
assistant = client.beta.assistants.create(
  instructions="You are a weather bot. Use the provided functions to answer questions.",
  model="gpt-4-1106-preview",
  tools=[{
     "type": "function",
    "function": {
     "name": "getCurrentWeather",
     "description": "Get the weather in location",
     "parameters": {
      "type": "object",
      "properties": {
       "location": {"type": "string", "description": "The city and state e.g. San Francisco, CA"},
       "unit": {"type": "string", "enum": ["c", "f"]}
      "required": ["location"]
    }
    "type": "function",
   "function": {
     "name": "getNickname",
     "description": "Get the nickname of a city",
     "parameters": {
      "type": "object",
      "properties": {
       "location": {"type": "string", "description": "The city and state e.g. San Francisco, CA"},
       "required": ["location"]
    }
) }]
```

Reading the functions called by the Assistant

When you initiate a Run with a user Message that triggers the function, the Run will enter a requires_action status. The model can provide multiple functions to call at once via the <u>parallel</u> function calling feature:

```
23
24
25
26
27
28
29
30
 "id": "run_3HV7rrQsagiqZmYynKwEdcxS",
 "object": "thread.run",
 "assistant_id": "asst_rEEOF3OGMan2ChvEALwTQakP",
 "thread_id": "thread_dXgWKGf8Cb7md8p0wKiMDGKc",
 "status": "requires_action",
 "required_action": {
  "type": "submit_tool_outputs",
  "submit_tool_outputs": {
   "tool_calls": [
      "tool_call_id": "call_Vt5AqcWr8QsRTNGv4cDlpsmA",
      "type": "function",
      "function": {
       "name": "getCurrentWeather",
       "arguments": "{\"location\":\"San Francisco\"}"
    },
      "tool_call_id": "call_45y0df8230430n34f8saa",
      "type": "function",
      "function": {
       "name": "getNickname",
       "arguments": "{\"location\":\"Los Angeles\"}"
 },
```

Submitting functions outputs

You can then complete the Run by submitting the output from the function(s) you call. Pass the tool call id referenced in the required action object above to match output to each function call.

```
python
Copy
1
2
3
4
5
6
7
8
9
10
11
12
13
14
```

After submitting outputs, the run will enter the queued state before it continues it's execution.

Supported files

For text/ MIME types, the encoding must be one of utf-8, utf-16, or ascii.

| FILE FORMAT | MIME TYPE | CODE INTERPRETER | RETRIEVA L |
|----------------|---|---------------------|---------------|
| .c | text/x-c | | |
| .срр | text/x-c++ | | |
| .csv | application/csv | | |
| .docx | application/vnd.openxmlformats-officedocument.wordprocessingml.document | | |
| .html | text/html | | |
| .java | text/x-java | | |
| .json | application/json | | |
| .md | text/markdown | | |
| .pdf | application/pdf | | |
| .php | text/x-php | | |
| .pptx | application/vnd.openxmlformats-officedocument.presentationml.presentation | | |
| .py | text/x-python | | |
| .py | text/x-script.python | | |
| .rb | text/x-ruby | | |
| .tex | text/x-tex | | |
| .txt | text/plain | | |
| .css | text/css | | |
| .jpeg | image/jpeg | | |
| .jpg | image/jpeg | | |
| .js | text/javascript | | |
| .gif | image/gif | | |
| .png | image/png | | |
| .tar | application/x-tar | | |

.ts application/typescript

.xlsx application/vnd.openxmlformats-

officedocument.spreadsheetml.sheet

.xml application/xml or "text/xml"

.zip application/zip

From < https://platform.openai.com/docs/assistants/tools/supported-files>

Prompt engineering

This guide shares strategies and tactics for getting better results from large language models (sometimes referred to as GPT models) like GPT-4. The methods described here can sometimes be deployed in combination for greater effect. We encourage experimentation to find the methods that work best for you.

Some of the examples demonstrated here currently work only with our most capable model, gpt-4. In general, if you find that a model fails at a task and a more capable model is available, it's often worth trying again with the more capable model.

Six strategies for getting better results

Write clear instructions

These models can't read your mind. If outputs are too long, ask for brief replies. If outputs are too simple, ask for expert-level writing. If you dislike the format, demonstrate the format you'd like to see. The less the model has to guess at what you want, the more likely you'll get it.

Tactics:

- Include details in your query to get more relevant answers
- Ask the model to adopt a persona
- Use delimiters to clearly indicate distinct parts of the input
- Specify the steps required to complete a task
- Provide examples
- Specify the desired length of the output

Provide reference text

Language models can confidently invent fake answers, especially when asked about esoteric topics or for citations and URLs. In the same way that a sheet of notes can help a student do better on a test, providing reference text to these models can help in answering with fewer fabrications.

Tactics:

- Instruct the model to answer using a reference text
- Instruct the model to answer with citations from a reference text

Split complex tasks into simpler subtasks

Just as it is good practice in software engineering to decompose a complex system into a set of modular components, the same is true of tasks submitted to a language model.

Complex tasks tend to have higher error rates than simpler tasks. Furthermore, complex tasks can often be re-defined as a workflow of simpler tasks in which the outputs of earlier tasks are used to construct the inputs to later tasks.

Tactics:

- Use intent classification to identify the most relevant instructions for a user query
- For dialogue applications that require very long conversations, summarize or filter previous dialogue
- Summarize long documents piecewise and construct a full summary recursively

Give the model time to "think"

If asked to multiply 17 by 28, you might not know it instantly, but can still work it out with time. Similarly, models make more reasoning errors when trying to answer right away, rather than taking time to work out an answer. Asking for a "chain of thought" before an answer can help the model reason its way toward correct answers more reliably.

Tactics:

- Instruct the model to work out its own solution before rushing to a conclusion
- Use inner monologue or a sequence of queries to hide the model's reasoning process
- Ask the model if it missed anything on previous passes

Use external tools

Compensate for the weaknesses of the model by feeding it the outputs of other tools. For example, a text retrieval system (sometimes called RAG or retrieval augmented generation) can tell the model about relevant documents. A code execution engine like OpenAl's Code Interpreter can help the model do math and run code. If a task can be done more reliably or efficiently by a tool rather than by a language model, offload it to get the best of both.

Tactics:

- Use embeddings-based search to implement efficient knowledge retrieval
- Use code execution to perform more accurate calculations or call external APIs
- Give the model access to specific functions

Test changes systematically

Improving performance is easier if you can measure it. In some cases a modification to a prompt will achieve better performance on a few isolated examples but lead to worse overall performance on a more representative set of examples. Therefore to be sure that a change is net positive to performance it may be necessary to define a comprehensive test suite (also known an as an "eval").

Tactic:

• Evaluate model outputs with reference to gold-standard answers

Tactics

Each of the strategies listed above can be instantiated with specific tactics. These tactics are meant to provide ideas for things to try. They are by no means fully comprehensive, and you should feel free to try creative ideas not represented here.

Strategy: Write clear instructions

Tactic: Include details in your query to get more relevant answers

In order to get a highly relevant response, make sure that requests provide any important details or context. Otherwise you are leaving it up to the model to guess what you mean.

| Worse | Better |
|---|---|
| How do I add numbers in Excel? | How do I add up a row of dollar amounts in Excel? I want to do this automatically for a whole sheet of rows with all the totals ending up on the right in a column called "Total". |
| Who's president? | Who was the president of Mexico in 2021, and how frequently are elections held? |
| Write code to calculate the Fibonacci sequence. | Write a TypeScript function to efficiently calculate the Fibonacci sequence. Comment the code liberally to explain what each piece does and why it's written that way. |
| Summarize the meeting notes. | Summarize the meeting notes in a single paragraph. Then write a markdown list of the speakers and each of their key points. Finally, list the next steps or action items suggested by the speakers, if any. |

Tactic: Ask the model to adopt a persona

The system message can be used to specify the persona used by the model in its replies.

SYSTEM

When I ask for help to write something, you will reply with a document that contains at least one joke or playful comment in every paragraph.

USER

Write a thank you note to my steel bolt vendor for getting the delivery in on time and in short notice. This made it possible for us to deliver an important order.

Open in Playground

Tactic: Use delimiters to clearly indicate distinct parts of the input

Delimiters like triple quotation marks, XML tags, section titles, etc. can help demarcate sections of text to be treated differently.

USER

Summarize the text delimited by triple quotes with a haiku. """insert text here"""

Open in Playground

SYSTEM

You will be provided with a pair of articles (delimited with XML tags) about the same topic. First summarize the arguments of each article. Then indicate which of them makes a better argument and explain why.

USER

<article> insert first article here </article> <article> insert second article here </article>

Open in Playground

SYSTEM

You will be provided with a thesis abstract and a suggested title for it. The thesis title should give the reader a good idea of the topic of the thesis but should also be eyecatching. If the title does not meet these criteria, suggest 5 alternatives.

Abstract: insert abstract here Title: insert title here

Open in Playground

For straightforward tasks such as these, using delimiters might not make a difference in the output quality. However, the more complex a task is the more important it is to disambiguate task details. Don't make the model work to understand exactly what you are asking of them.

Tactic: Specify the steps required to complete a task

Some tasks are best specified as a sequence of steps. Writing the steps out explicitly can make it easier for the model to follow them.

SYSTEM

Use the following step-by-step instructions to respond to user inputs. Step 1 - The user will provide you with text in triple quotes. Summarize this text in one sentence with a prefix that says "Summary: ". Step 2 - Translate the summary from Step 1 into Spanish, with a prefix that says "Translation: ".

USER

"""insert text here"""

Open in Playground

Tactic: Provide examples

Providing general instructions that apply to all examples is generally more efficient than demonstrating all permutations of a task by example, but in some cases providing examples may be easier. For example, if you intend for the model to copy a particular style of responding to user queries which is difficult to describe explicitly. This is known as "few-shot" prompting.

SYSTEM

Answer in a consistent style.

USFR

Teach me about patience.

ASSISTANT

The river that carves the deepest valley flows from a modest spring; the grandest symphony originates from a single note; the most intricate tapestry begins with a solitary thread.

USER

Teach me about the ocean.

Open in Playground

Tactic: Specify the desired length of the output

You can ask the model to produce outputs that are of a given target length. The

targeted output length can be specified in terms of the count of words, sentences, paragraphs, bullet points, etc. Note however that instructing the model to generate a specific number of words does not work with high precision. The model can more reliably generate outputs with a specific number of paragraphs or bullet points.

USER

Summarize the text delimited by triple quotes in about 50 words. """insert text here"""

Open in Playground

USER

Summarize the text delimited by triple quotes in 2 paragraphs. """insert text here"""

Open in Playground

USER

Summarize the text delimited by triple quotes in 3 bullet points. """insert text here"""

Open in Playground

Strategy: Provide reference text

Tactic: Instruct the model to answer using a reference text

If we can provide a model with trusted information that is relevant to the current query, then we can instruct the model to use the provided information to compose its answer.

SYSTEM

Use the provided articles delimited by triple quotes to answer questions. If the answer cannot be found in the articles, write "I could not find an answer."

USER

<insert articles, each delimited by triple quotes> Question: <insert question here>

Open in Playground

Given that all models have limited context windows, we need some way to dynamically lookup information that is relevant to the question being asked. Embeddings can be used to implement efficient knowledge retrieval. See the tactic "Use embeddings-based search to implement efficient knowledge retrieval" for more details on how to implement this.

Tactic: Instruct the model to answer with citations from a reference text

If the input has been supplemented with relevant knowledge, it's straightforward to request that the model add citations to its answers by referencing passages from provided documents. Note that citations in the output can then be verified programmatically by string matching within the provided documents.

SYSTEM

You will be provided with a document delimited by triple quotes and a question. Your task is to answer the question using only the provided document and to cite the passage(s) of the document used to answer the question. If the document does not contain the information needed to answer this question then simply write: "Insufficient information." If an answer to the question is provided, it must be annotated with a

citation. Use the following format for to cite relevant passages ({"citation": ...}). **USER**

"""<insert document here>""" Question: <insert question here>

Open in Playground

Strategy: Split complex tasks into simpler subtasks Tactic: Use intent classification to identify the most relevant instructions for a user query

For tasks in which lots of independent sets of instructions are needed to handle different cases, it can be beneficial to first classify the type of query and to use that classification to determine which instructions are needed. This can be achieved by defining fixed categories and hardcoding instructions that are relevant for handling tasks in a given category. This process can also be applied recursively to decompose a task into a sequence of stages. The advantage of this approach is that each query will contain only those instructions that are required to perform the next stage of a task which can result in lower error rates compared to using a single query to perform the whole task. This can also result in lower costs since larger prompts cost more to run (see pricing information).

Suppose for example that for a customer service application, queries could be usefully classified as follows:

SYSTEM

You will be provided with customer service queries. Classify each query into a primary category and a secondary category. Provide your output in json format with the keys: primary and secondary. Primary categories: Billing, Technical Support, Account Management, or General Inquiry. Billing secondary categories: - Unsubscribe or upgrade - Add a payment method - Explanation for charge - Dispute a charge Technical Support secondary categories: - Troubleshooting - Device compatibility - Software updates Account Management secondary categories: - Password reset - Update personal information - Close account - Account security General Inquiry secondary categories: - Product information - Pricing - Feedback - Speak to a human USER

I need to get my internet working again.

Open in Playground

Based on the classification of the customer query, a set of more specific instructions can be provided to a model for it to handle next steps. For example, suppose the customer requires help with "troubleshooting".

SYSTEM

You will be provided with customer service inquiries that require troubleshooting in a technical support context. Help the user by: - Ask them to check that all cables to/from the router are connected. Note that it is common for cables to come loose over time. - If all cables are connected and the issue persists, ask them which router model they are using - Now you will advise them how to restart their device: -- If the model number is MTD-327J, advise them to push the red button and hold it for 5 seconds, then wait 5 minutes before testing the connection. -- If the model number is MTD-327S, advise them to unplug and replug it, then wait 5 minutes before testing the connection. - If the customer's issue persists after restarting the device and waiting 5 minutes, connect

them to IT support by outputting {"IT support requested"}. - If the user starts asking questions that are unrelated to this topic then confirm if they would like to end the current chat about troubleshooting and classify their request according to the following scheme: <insert primary/secondary classification scheme from above here>

I need to get my internet working again.

Open in Playground

Notice that the model has been instructed to emit special strings to indicate when the state of the conversation changes. This enables us to turn our system into a state machine where the state determines which instructions are injected. By keeping track of state, what instructions are relevant at that state, and also optionally what state transitions are allowed from that state, we can put guardrails around the user experience that would be hard to achieve with a less structured approach.

<u>Tactic:</u> For dialogue applications that require very long conversations, summarize or filter previous dialogue

Since models have a fixed context length, dialogue between a user and an assistant in which the entire conversation is included in the context window cannot continue indefinitely.

There are various workarounds to this problem, one of which is to summarize previous turns in the conversation. Once the size of the input reaches a predetermined threshold length, this could trigger a query that summarizes part of the conversation and the summary of the prior conversation could be included as part of the system message. Alternatively, prior conversation could be summarized asynchronously in the background throughout the entire conversation.

An alternative solution is to dynamically select previous parts of the conversation that are most relevant to the current query. See the tactic "Use embeddings-based search to implement efficient knowledge retrieval".

Tactic: Summarize long documents piecewise and construct a full summary recursively

Since models have a fixed context length, they cannot be used to summarize a text longer than the context length minus the length of the generated summary in a single query.

To summarize a very long document such as a book we can use a sequence of queries to summarize each section of the document. Section summaries can be concatenated and summarized producing summaries of summaries. This process can proceed recursively until an entire document is summarized. If it's necessary to use information about earlier sections in order to make sense of later sections, then a further trick that can be useful is to include a running summary of the text that precedes any given point in the book while summarizing content at that point. The effectiveness of this procedure for summarizing books has been studied in previous research by OpenAI using variants of GPT-3.

Strategy: Give models time to "think"

Tactic: Instruct the model to work out its own solution before rushing to

a conclusion

Sometimes we get better results when we explicitly instruct the model to reason from first principles before coming to a conclusion. Suppose for example we want a model to evaluate a student's solution to a math problem. The most obvious way to approach this is to simply ask the model if the student's solution is correct or not.

SYSTEM

Determine if the student's solution is correct or not.

USER

Problem Statement: I'm building a solar power installation and I need help working out the financials. - Land costs \$100 / square foot - I can buy solar panels for \$250 / square foot - I negotiated a contract for maintenance that will cost me a flat \$100k per year, and an additional \$10 / square foot What is the total cost for the first year of operations as a function of the number of square feet. Student's Solution: Let x be the size of the installation in square feet. 1. Land cost: 100x 2. Solar panel cost: 250x 3. Maintenance cost: 100,000 + 100x Total cost: 100x + 250x + 100,000 + 100x = 450x + 100,000

The student's solution is correct.

Open in Playground

But the student's solution is actually not correct! We can get the model to successfully notice this by prompting the model to generate its own solution first.

SYSTEM

First work out your own solution to the problem. Then compare your solution to the student's solution and evaluate if the student's solution is correct or not. Don't decide if the student's solution is correct until you have done the problem yourself.

USER

Problem Statement: I'm building a solar power installation and I need help working out the financials. - Land costs \$100 / square foot - I can buy solar panels for \$250 / square foot - I negotiated a contract for maintenance that will cost me a flat \$100k per year, and an additional \$10 / square foot What is the total cost for the first year of operations as a function of the number of square feet. Student's Solution: Let x be the size of the installation in square feet. 1. Land cost: 100x 2. Solar panel cost: 250x 3. Maintenance cost: 100,000 + 100x Total cost: 100x + 250x + 100,000 + 100x = 450x + 100,000

Let x be the size of the installation in square feet. 1. Land cost: 100x 2. Solar panel cost: 250x 3. Maintenance cost: 100,000 + 10x Total cost: 100x + 250x + 100,000 + 10x = 360x + 100,000 The student's solution is incorrect. They made an error in the maintenance cost calculation, using 100x instead of 10x. The correct total cost for the first year of operations is 360x + 100,000.

Open in Playground

Tactic: Use inner monologue or a sequence of queries to hide the model's reasoning process

The previous tactic demonstrates that it is sometimes important for the model to reason in detail about a problem before answering a specific question. For some applications, the reasoning process that a model uses to arrive at a final answer would be inappropriate to share with the user. For example, in tutoring applications we may want to encourage students to work out their own answers, but a model's reasoning process

about the student's solution could reveal the answer to the student.

Inner monologue is a tactic that can be used to mitigate this. The idea of inner monologue is to instruct the model to put parts of the output that are meant to be hidden from the user into a structured format that makes parsing them easy. Then before presenting the output to the user, the output is parsed and only part of the output is made visible.

SYSTEM

Follow these steps to answer the user queries. Step 1 - First work out your own solution to the problem. Don't rely on the student's solution since it may be incorrect. Enclose all your work for this step within triple quotes ("""). Step 2 - Compare your solution to the student's solution and evaluate if the student's solution is correct or not. Enclose all your work for this step within triple quotes ("""). Step 3 - If the student made a mistake, determine what hint you could give the student without giving away the answer. Enclose all your work for this step within triple quotes ("""). Step 4 - If the student made a mistake, provide the hint from the previous step to the student (outside of triple quotes). Instead of writing "Step 4 - ..." write "Hint:".

USER

Problem Statement: <insert problem statement> Student Solution: <insert student solution>

Open in Playground

Alternatively, this can be achieved with a sequence of queries in which all except the last have their output hidden from the end user.

First, we can ask the model to solve the problem on its own. Since this initial query doesn't require the student's solution, it can be omitted. This provides the additional advantage that there is no chance that the model's solution will be biased by the student's attempted solution.

USER

<insert problem statement>

Open in Playground

Next, we can have the model use all available information to assess the correctness of the student's solution.

SYSTEM

Compare your solution to the student's solution and evaluate if the student's solution is correct or not.

USER

Problem statement: """<insert problem statement>""" Your solution: """<insert model generated solution>""" Student's solution: """"<insert student's solution>"""

Open in Playground

Finally, we can let the model use its own analysis to construct a reply in the persona of a helpful tutor.

SYSTEM

You are a math tutor. If the student made an error, offer a hint to the student in a way that does not reveal the answer. If the student did not make an error, simply offer them an encouraging comment.

USER

Problem statement: """<insert problem statement>""" Your solution: """<insert model generated solution>""" Student's solution: """<insert student's solution>""" Analysis: """<insert model generated analysis from previous step>"""

Open in Playground

Tactic: Ask the model if it missed anything on previous passes

Suppose that we are using a model to list excerpts from a source which are relevant to a particular question. After listing each excerpt the model needs to determine if it should start writing another or if it should stop. If the source document is large, it is common for a model to stop too early and fail to list all relevant excerpts. In that case, better performance can often be obtained by prompting the model with followup queries to find any excerpts it missed on previous passes.

SYSTEM

You will be provided with a document delimited by triple quotes. Your task is to select excerpts which pertain to the following question: "What significant paradigm shifts have occurred in the history of artificial intelligence." Ensure that excerpts contain all relevant context needed to interpret them - in other words don't extract small snippets that are missing important context. Provide output in JSON format as follows: [{"excerpt":

"..."}, ... {"excerpt": "..."}]

USER

"""<insert document here>"""

ASSISTANT

[{"excerpt": "the model writes an excerpt here"}, ... {"excerpt": "the model writes another excerpt here"}]

USER

Are there more relevant excerpts? Take care not to repeat excerpts. Also ensure that excerpts contain all relevant context needed to interpret them - in other words don't extract small snippets that are missing important context.

Open in Playground

Strategy: Use external tools

<u>Tactic: Use embeddings-based search to implement efficient knowledge</u> <u>retrieval</u>

A model can leverage external sources of information if provided as part of its input. This can help the model to generate more informed and up-to-date responses. For example, if a user asks a question about a specific movie, it may be useful to add high quality information about the movie (e.g. actors, director, etc...) to the model's input. Embeddings can be used to implement efficient knowledge retrieval, so that relevant information can be added to the model input dynamically at run-time.

A text embedding is a vector that can measure the relatedness between text strings. Similar or relevant strings will be closer together than unrelated strings. This fact, along with the existence of fast vector search algorithms means that embeddings can be used to implement efficient knowledge retrieval. In particular, a text corpus can be split up

into chunks, and each chunk can be embedded and stored. Then a given query can be embedded and vector search can be performed to find the embedded chunks of text from the corpus that are most related to the query (i.e. closest together in the embedding space).

Example implementations can be found in the <u>OpenAl Cookbook</u>. See the tactic <u>"Instruct the model to use retrieved knowledge to answer queries"</u> for an example of how to use knowledge retrieval to minimize the likelihood that a model will make up incorrect facts.

<u>Tactic: Use code execution to perform more accurate calculations or call</u> external APIs

Language models cannot be relied upon to perform arithmetic or long calculations accurately on their own. In cases where this is needed, a model can be instructed to write and run code instead of making its own calculations. In particular, a model can be instructed to put code that is meant to be run into a designated format such as triple backtick. After an output is produced, the code can be extracted and run. Finally, if necessary, the output from the code execution engine (i.e. Python interpreter) can be provided as an input to the model for the next query.

SYSTEM

You can write and execute Python code by enclosing it in triple backticks, e.g. ```code goes here```. Use this to perform calculations.

Find all real-valued roots of the following polynomial: 3*x**5 - 5*x**4 - 3*x**3 - 7*x - 10.

Open in Playground

Another good use case for code execution is calling external APIs. If a model is instructed in the proper use of an API, it can write code that makes use of it. A model can be instructed in how to use an API by providing it with documentation and/or code samples showing how to use the API.

SYSTEM

You can write and execute Python code by enclosing it in triple backticks. Also note that you have access to the following module to help users send messages to their friends: ```python import message message.write(to="John", message="Hey, want to meetup after work?")```

Open in Playground

WARNING: Executing code produced by a model is not inherently safe and precautions should be taken in any application that seeks to do this. In particular, a sandboxed code execution environment is needed to limit the harm that untrusted code could cause.

Tactic: Give the model access to specific functions

The Chat Completions API allows passing a list of function descriptions in requests. This enables models to generate function arguments according to the provided schemas. Generated function arguments are returned by the API in JSON format and can be used to execute function calls. Output provided by function calls can then be fed back into a model in the following request to close the loop. This is the recommended

way of using OpenAI models to call external functions. To learn more see the <u>function</u> <u>calling section</u> in our introductory text generation guide and more <u>function calling</u> <u>examples</u> in the OpenAI Cookbook.

Strategy: Test changes systematically

Sometimes it can be hard to tell whether a change — e.g., a new instruction or a new design — makes your system better or worse. Looking at a few examples may hint at which is better, but with small sample sizes it can be hard to distinguish between a true improvement or random luck. Maybe the change helps performance on some inputs, but hurts performance on others.

Evaluation procedures (or "evals") are useful for optimizing system designs. Good evals are:

- Representative of real-world usage (or at least diverse)
- Contain many test cases for greater statistical power (see table below for guidelines)
- Easy to automate or repeat

DIFFERENCE TO DETECT SAMPLE SIZE NEEDED FOR 95% CONFIDENCE

| 30% | ~10 |
|-----|---------|
| 10% | ~100 |
| 3% | ~1,000 |
| 1% | ~10,000 |

Evaluation of outputs can be done by computers, humans, or a mix. Computers can automate evals with objective criteria (e.g., questions with single correct answers) as well as some subjective or fuzzy criteria, in which model outputs are evaluated by other model queries. OpenAl Evals is an open-source software framework that provides tools for creating automated evals.

Model-based evals can be useful when there exists a range of possible outputs that would be considered equally high in quality (e.g. for questions with long answers). The boundary between what can be realistically evaluated with a model-based eval and what requires a human to evaluate is fuzzy and is constantly shifting as models become more capable. We encourage experimentation to figure out how well model-based evals can work for your use case.

Tactic: Evaluate model outputs with reference to gold-standard answers

Suppose it is known that the correct answer to a question should make reference to a specific set of known facts. Then we can use a model query to count how many of the required facts are included in the answer.

For example, using the following system message:

SYSTEM

You will be provided with text delimited by triple quotes that is supposed to be the answer to a question. Check if the following pieces of information are directly contained in the answer: - Neil Armstrong was the first person to walk on the moon. - The date Neil Armstrong first walked on the moon was July 21, 1969. For each of these points perform the following steps: 1 - Restate the point. 2 - Provide a citation from the answer which is closest to this point. 3 - Consider if someone reading the citation who doesn't

know the topic could directly infer the point. Explain why or why not before making up your mind. 4 - Write "yes" if the answer to 3 was yes, otherwise write "no". Finally, provide a count of how many "yes" answers there are. Provide this count as {"count": <insert count here>}.

Here's an example input where both points are satisfied:

SYSTEM

<insert system message above>

"""Neil Armstrong is famous for being the first human to set foot on the Moon. This historic event took place on July 21, 1969, during the Apollo 11 mission."""

Open in Playground

Here's an example input where only one point is satisfied:

SYSTEM

<insert system message above>

USER

"""Neil Armstrong made history when he stepped off the lunar module, becoming the first person to walk on the moon."""

Open in Playground

Here's an example input where none are satisfied:

SYSTEM

<insert system message above>

USER

"""In the summer of '69, a voyage grand, Apollo 11, bold as legend's hand. Armstrong took a step, history unfurled, "One small step," he said, for a new world."""

Open in Playground

There are many possible variants on this type of model-based eval. Consider the following variation which tracks the kind of overlap between the candidate answer and the gold-standard answer, and also tracks whether the candidate answer contradicts any part of the gold-standard answer.

SYSTEM

Use the following steps to respond to user inputs. Fully restate each step before proceeding. i.e. "Step 1: Reason...". Step 1: Reason step-by-step about whether the information in the submitted answer compared to the expert answer is either: disjoint, equal, a subset, a superset, or overlapping (i.e. some intersection but not subset/superset). Step 2: Reason step-by-step about whether the submitted answer contradicts any aspect of the expert answer. Step 3: Output a JSON object structured like: {"type_of_overlap": "disjoint" or "equal" or "subset" or "superset" or "overlapping", "contradiction": true or false}

Here's an example input with a substandard answer which nonetheless does not contradict the expert answer:

SYSTEM

<insert system message above>

USER

Question: """What event is Neil Armstrong most famous for and on what date did it occur? Assume UTC time.""" Submitted Answer: """Didn't he walk on the moon or something?""" Expert Answer: """Neil Armstrong is most famous for being the first person to walk on the moon. This historic event occurred on July 21, 1969."""

Open in Playground

Here's an example input with answer that directly contradicts the expert answer:

SYSTEM

<insert system message above>

IISER

Question: """What event is Neil Armstrong most famous for and on what date did it occur? Assume UTC time.""" Submitted Answer: """On the 21st of July 1969, Neil Armstrong became the second person to walk on the moon, following after Buzz Aldrin.""" Expert Answer: """Neil Armstrong is most famous for being the first person to walk on the moon. This historic event occurred on July 21, 1969."""

Open in Playground

Here's an example input with a correct answer that also provides a bit more detail than is necessary:

SYSTEM

<insert system message above>

USER

Question: """What event is Neil Armstrong most famous for and on what date did it occur? Assume UTC time.""" Submitted Answer: """At approximately 02:56 UTC on July 21st 1969, Neil Armstrong became the first human to set foot on the lunar surface, marking a monumental achievement in human history.""" Expert Answer: """Neil Armstrong is most famous for being the first person to walk on the moon. This historic event occurred on July 21, 1969."""

Open in Playground

Other resources

For more inspiration, visit the OpenAl Cookbook, which contains example code and also links to third-party resources such as:

- Prompting libraries & tools
- Prompting guides
- Video courses
- Papers on advanced prompting to improve reasoning

From < https://platform.openai.com/docs/guides/prompt-engineering/other-resources>

Production best practices

This guide provides a comprehensive set of best practices to help you transition from

prototype to production. Whether you are a seasoned machine learning engineer or a recent enthusiast, this guide should provide you with the tools you need to successfully put the platform to work in a production setting: from securing access to our API to designing a robust architecture that can handle high traffic volumes. Use this guide to help develop a plan for deploying your application as smoothly and effectively as possible.

Setting up your organization

Once you <u>log in</u> to your OpenAl account, you can find your organization name and ID in your <u>organization settings</u>. The organization name is the label for your organization, shown in user interfaces. The organization ID is the unique identifier for your organization which can be used in API requests.

Users who belong to multiple organizations can <u>pass a header</u> to specify which organization is used for an API request. Usage from these API requests will count against the specified organization's quota. If no header is provided, the <u>default organization</u> will be billed. You can change your default organization in your <u>user settings</u>.

You can invite new members to your organization from the <u>Team page</u>. Members can be **readers** or **owners**. Readers can make API requests and view basic organization information, while owners can modify billing information and manage members within an organization.

Managing billing limits

New free trial users receive an initial credit of \$5 that expires after three months. Once the credit has been used or expires, you can choose to enter <u>billing information</u> to continue your use of the API. If no billing information is entered, you will still have login access but will be unable to make any further API requests.

Once you've entered your billing information, you will have an approved usage limit of \$100 per month, which is set by OpenAI. Your quota limit will automatically increase as your usage on your platform increases and you move from one <u>usage tier</u> to another. You can review your current usage limit in the <u>limits</u> page in your account settings.

If you'd like to be notified when your usage exceeds a certain dollar amount, you can set a notification threshold through the <u>usage limits</u> page. When the notification threshold is reached, the owners of the organization will receive an email notification. You can also set a monthly budget so that, once the monthly budget is reached, any subsequent API requests will be rejected. Note that these limits are best effort, and there may be 5 to 10 minutes of delay between the usage and the limits being enforced.

API keys

The OpenAl API uses API keys for authentication. Visit your <u>API keys</u> page to retrieve the API key you'll use in your requests.

This is a relatively straightforward way to control access, but you must be vigilant about securing these keys. Avoid exposing the API keys in your code or in public repositories; instead, store them in a secure location. You should expose your keys to your application using environment variables or secret management service, so that you don't need to hard-code them in your codebase. Read more in our Best practices for API key safety.

Staging accounts

As you scale, you may want to create separate organizations for your staging and production environments. Please note that you can sign up using two separate email addresses like bob+prod@widgetcorp.com and bob+dev@widgetcorp.com to create two organizations. This will allow you to isolate your development and testing work so you don't accidentally disrupt your live application. You can also limit access to your production organization this way.

Scaling your solution architecture

When designing your application or service for production that uses our API, it's important to consider how you will scale to meet traffic demands. There are a few key areas you will need to consider regardless of the cloud service provider of your choice:

- Horizontal scaling: You may want to scale your application out horizontally to
 accommodate requests to your application that come from multiple sources. This
 could involve deploying additional servers or containers to distribute the load. If
 you opt for this type of scaling, make sure that your architecture is designed to
 handle multiple nodes and that you have mechanisms in place to balance the load
 between them.
- Vertical scaling: Another option is to scale your application up vertically, meaning
 you can beef up the resources available to a single node. This would involve
 upgrading your server's capabilities to handle the additional load. If you opt for this
 type of scaling, make sure your application is designed to take advantage of these
 additional resources.
- Caching: By storing frequently accessed data, you can improve response times
 without needing to make repeated calls to our API. Your application will need to be
 designed to use cached data whenever possible and invalidate the cache when
 new information is added. There are a few different ways you could do this. For
 example, you could store data in a database, filesystem, or in-memory cache,
 depending on what makes the most sense for your application.
- Load balancing: Finally, consider load-balancing techniques to ensure requests are distributed evenly across your available servers. This could involve using a load balancer in front of your servers or using DNS round-robin. Balancing the load will help improve performance and reduce bottlenecks.

Managing rate limits

When using our API, it's important to understand and plan for rate limits.

Improving latencies

Latency is the time it takes for a request to be processed and a response to be returned. In this section, we will discuss some factors that influence the latency of our text generation models and provide suggestions on how to reduce it.

The latency of a completion request is mostly influenced by two factors: the model and the number of tokens generated. The life cycle of a completion request looks like this:

Network

End user to API latency

Server

Time to process prompt tokens

Network

API to end user latency

The bulk of the latency typically arises from the token generation step.

Intuition: Prompt tokens add very little latency to completion calls. Time to generate completion tokens is much longer, as tokens are generated one at a time. Longer generation lengths will accumulate latency due to generation required for each token.

Common factors affecting latency and possible mitigation techniques

Now that we have looked at the basics of latency, let's take a look at various factors that can affect latency, broadly ordered from most impactful to least impactful.

Model

Our API offers different models with varying levels of complexity and generality. The most capable models, such as gpt-4, can generate more complex and diverse completions, but they also take longer to process your query. Models such as gpt-3.5-turbo, can generate faster and cheaper chat completions, but they may generate results that are less accurate or relevant for your query. You can choose the model that best suits your use case and the trade-off between speed and quality.

Number of completion tokens

Requesting a large amount of generated tokens completions can lead to increased latencies:

- Lower max tokens: for requests with a similar token generation count, those that have a lower max_tokens parameter incur less latency.
- Include stop sequences: to prevent generating unneeded tokens, add a stop sequence. For example, you can use stop sequences to generate a list with a specific number of items. In this case, by using 11. as a stop sequence, you can generate a list with only 10 items, since the completion will stop when 11. is reached. Read our help article on stop sequences for more context on how you can do this.
- **Generate fewer completions**: lower the values of n and best_of when possible where n refers to how many completions to generate for each prompt and best_of is used to represent the result with the highest log probability per token.

If n and best_of both equal 1 (which is the default), the number of generated tokens will be at most, equal to max_tokens.

If n (the number of completions returned) or best_of (the number of completions generated for consideration) are set to > 1, each request will create multiple outputs. Here, you can consider the number of generated tokens as [max_tokens * max (n, best_of)]

Streaming

Setting stream: true in a request makes the model start returning tokens as soon as they are available, instead of waiting for the full sequence of tokens to be generated. It does not change the time to get all the tokens, but it reduces the time for first token for an application where we want to show partial progress or are going to stop generations. This can be a better user experience and a UX improvement so it's worth experimenting

with streaming.

Infrastructure

Our servers are currently located in the US. While we hope to have global redundancy in the future, in the meantime you could consider locating the relevant parts of your infrastructure in the US to minimize the roundtrip time between your servers and the OpenAI servers.

Batching

Depending on your use case, batching *may help*. If you are sending multiple requests to the same endpoint, you can <u>batch the prompts</u> to be sent in the same request. This will reduce the number of requests you need to make. The prompt parameter can hold up to 20 unique prompts. We advise you to test out this method and see if it helps. In some cases, you may end up increasing the number of generated tokens which will slow the response time.

Managing costs

To monitor your costs, you can set a <u>notification threshold</u> in your account to receive an email alert once you pass a certain usage threshold. You can also set a <u>monthly budget</u>. Please be mindful of the potential for a monthly budget to cause disruptions to your application/users. Use the <u>usage tracking dashboard</u> to monitor your token usage during the current and past billing cycles.

Text generation

One of the challenges of moving your prototype into production is budgeting for the costs associated with running your application. OpenAl offers a <u>pay-as-you-go pricing</u> <u>model</u>, with prices per 1,000 tokens (roughly equal to 750 words). To estimate your costs, you will need to project the token utilization. Consider factors such as traffic levels, the frequency with which users will interact with your application, and the amount of data you will be processing.

One useful framework for thinking about reducing costs is to consider costs as a function of the number of tokens and the cost per token. There are two potential avenues for reducing costs using this framework. First, you could work to reduce the cost per token by switching to smaller models for some tasks in order to reduce costs. Alternatively, you could try to reduce the number of tokens required. There are a few ways you could do this, such as by using shorter prompts, fine-tuning models, or caching common user queries so that they don't need to be processed repeatedly.

You can experiment with our interactive <u>tokenizer tool</u> to help you estimate costs. The API and playground also returns token counts as part of the response. Once you've got things working with our most capable model, you can see if the other models can produce the same results with lower latency and costs. Learn more in our <u>token usage help article</u>.

MLOps strategy

As you move your prototype into production, you may want to consider developing an MLOps strategy. MLOps (machine learning operations) refers to the process of managing the end-to-end life cycle of your machine learning models, including any models you may be fine-tuning using our API. There are a number of areas to consider when designing your MLOps strategy. These include

- Data and model management: managing the data used to train or fine-tune your model and tracking versions and changes.
- Model monitoring: tracking your model's performance over time and detecting any potential issues or degradation.
- Model retraining: ensuring your model stays up to date with changes in data or evolving requirements and retraining or fine-tuning it as needed.
- Model deployment: automating the process of deploying your model and related artifacts into production.

Thinking through these aspects of your application will help ensure your model stays relevant and performs well over time.

Security and compliance

As you move your prototype into production, you will need to assess and address any security and compliance requirements that may apply to your application. This will involve examining the data you are handling, understanding how our API processes data, and determining what regulations you must adhere to. Our <u>security practices</u> and <u>trust and compliance portal</u> provide our most comprehensive and up-to-date documentation. For reference, here is our <u>Privacy Policy</u> and <u>Terms of Use</u>.

Some common areas you'll need to consider include data storage, data transmission, and data retention. You might also need to implement data privacy protections, such as encryption or anonymization where possible. In addition, you should follow best practices for secure coding, such as input sanitization and proper error handling.

Safety best practices

When creating your application with our API, consider our <u>safety best practices</u> to ensure your application is safe and successful. These recommendations highlight the importance of testing the product extensively, being proactive about addressing potential issues, and limiting opportunities for misuse.

Safety best practices

Use our free Moderation API

OpenAI's <u>Moderation API</u> is free-to-use and can help reduce the frequency of unsafe content in your completions. Alternatively, you may wish to develop your own content filtration system tailored to your use case.

Adversarial testing

We recommend "red-teaming" your application to ensure it's robust to adversarial input. Test your product over a wide range of inputs and user behaviors, both a representative set and those reflective of someone trying to 'break' your application. Does it wander off topic? Can someone easily redirect the feature via prompt injections, e.g. "ignore the previous instructions and do this instead"?

Human in the loop (HITL)

Wherever possible, we recommend having a human review outputs before they are used in practice. This is especially critical in high-stakes domains, and for code generation. Humans should be aware of the limitations of the system, and have access to any information needed to verify the outputs (for example, if the application summarizes notes, a human should have easy access to the original notes to refer back).

Prompt engineering

"Prompt engineering" can help constrain the topic and tone of output text. This reduces the chance of producing undesired content, even if a user tries to produce it. Providing additional context to the model (such as by giving a few high-quality examples of desired behavior prior to the new input) can make it easier to steer model outputs in desired directions.

"Know your customer" (KYC)

Users should generally need to register and log-in to access your service. Linking this service to an existing account, such as a Gmail, LinkedIn, or Facebook log-in, may help, though may not be appropriate for all use-cases. Requiring a credit card or ID card reduces risk further.

Constrain user input and limit output tokens

Limiting the amount of text a user can input into the prompt helps avoid prompt injection. Limiting the number of output tokens helps reduce the chance of misuse.

Narrowing the ranges of inputs or outputs, especially drawn from trusted sources, reduces the extent of misuse possible within an application.

Allowing user inputs through validated dropdown fields (e.g., a list of movies on Wikipedia) can be more secure than allowing open-ended text inputs.

Returning outputs from a validated set of materials on the backend, where possible, can be safer than returning novel generated content (for instance, routing a customer query to the best-matching existing customer support article, rather than attempting to answer the query from-scratch).

Allow users to report issues

Users should generally have an easily-available method for reporting improper functionality or other concerns about application behavior (listed email address, ticket submission method, etc). This method should be monitored by a human and responded to as appropriate.

Understand and communicate limitations

From hallucinating inaccurate information, to offensive outputs, to bias, and much more, language models may not be suitable for every use case without significant modifications. Consider whether the model is fit for your purpose, and evaluate the performance of the API on a wide range of potential inputs in order to identify cases where the API's performance might drop. Consider your customer base and the range of inputs that they will be using, and ensure their expectations are calibrated appropriately.

Safety and security are very important to us at OpenAl.

If in the course of your development you do notice any safety or security issues with the API or anything else related to OpenAI, please submit these through our <u>Coordinated Vulnerability Disclosure Program</u>.

End-user IDs

Sending end-user IDs in your requests can be a useful tool to help OpenAI monitor and detect abuse. This allows OpenAI to provide your team with more actionable feedback in the event that we detect any policy violations in your application.

The IDs should be a string that uniquely identifies each user. We recommend hashing their username or email address, in order to avoid sending us any identifying information. If you offer a preview of your product to non-logged in users, you can send a session ID instead.

You can include end-user IDs in your API requests via the user parameter as follows:

```
Example: Providing a user identifer
python
Copy
2
3
4
5
6
7
8
from openai import OpenAI
client = OpenAI()
response = client.completions.create(
 model="gpt-3.5-turbo-instruct",
 prompt="This is a test",
 max_tokens=5,
 user="user_123456"
```

From < https://platform.openai.com/docs/guides/safety-best-practices>

Rate limits

Rate limits are restrictions that our API imposes on the number of times a user or client can access our services within a specified period of time.

Why do we have rate limits?

Rate limits are a common practice for APIs, and they're put in place for a few different reasons:

- They help protect against abuse or misuse of the API. For example, a malicious actor could flood the API with requests in an attempt to overload it or cause disruptions in service. By setting rate limits, OpenAI can prevent this kind of activity.
- Rate limits help ensure that everyone has fair access to the API. If one person or organization makes an excessive number of requests, it could bog down the API for everyone else. By throttling the number of requests that a single user can make, OpenAI ensures that the most number of people have an opportunity to use the API without experiencing slowdowns.
- Rate limits can help OpenAI manage the aggregate load on its infrastructure. If requests to the

API increase dramatically, it could tax the servers and cause performance issues. By setting rate limits, OpenAI can help maintain a smooth and consistent experience for all users.

Please work through this document in its entirety to better understand how OpenAl's rate limit system works. We include code examples and possible solutions to handle common issues. We also include details around how your rate limits are automatically increased in the usage tiers section below.

How do these rate limits work?

Rate limits are measured in four ways: **RPM** (requests per minute), **RPD** (requests per day), **TPM** (tokens per minute), and **IPM** (images per minute). Rate limits can be hit across any of the options depending on what occurs first. For example, you might send 20 requests with only 100 tokens to the ChatCompletions endpoint and that would fill your limit (if your RPM was 20), even if you did not send 150k tokens (if your TPM limit was 150k) within those 20 requests.

Other important things worth noting:

- Rate limits are imposed at the <u>organization level</u>, not user level.
- Rate limits vary by the model being used.
- Limits are also placed on the total amount an organization can spend on the API each month. These are also known as "usage limits".

Usage tiers

You can view the rate and usage limits for your organization under the <u>limits</u> section of your account settings. As your usage of the OpenAI API and your spend on our API goes up, we automatically graduate you to the next usage tier. This usually results in an increase in rate limits across most models. Organizations in higher tiers also get access to lower latency models.

| TIER | QUALIFICATION | USAGE LIMITS | | | |
|--|--|------------------|--|--|--|
| Free | User must be in an allowed geography | \$100 / month | | | |
| Tier 1 | \$5 paid | \$100 / month | | | |
| Tier 2 | \$50 paid and 7+ days since first successful payment | \$500 / month | | | |
| Tier 3 | \$100 paid and 7+ days since first successful payment | \$1,000 / month | | | |
| Tier 4 | \$250 paid and 14+ days since first successful payment | \$5,000 / month | | | |
| Tier 5 | \$1,000 paid and 30+ days since first successful payment | \$10,000 / month | | | |
| Select a tier below to view a high-level summary of rate limits per model. | | | | | |

FreeTier 1Tier 2Tier 3Tier 4Tier 5

Free tier rate limits

This is a high level summary and there are per-model exceptions to these limits (e.g. some legacy models or models with larger context windows have different rate limits). To view the exact rate limits per model for your account, visit the <u>limits</u> section of your account settings.

| MODEL | RPM | RPD | TPM |
|------------------------|-----------|-----|---------|
| gpt-4 | 3 | 200 | 10,000 |
| gpt-3.5-turbo | 3 | 200 | 20,000 |
| text-embedding-ada-002 | 3 | 200 | 150,000 |
| whisper-1 | 3 | 200 | |
| tts-1 | 3 | 200 | |
| dall-e-2 | 5 img/min | | |

Rate limits in headers

In addition to seeing your rate limit on your <u>account page</u>, you can also view important information about your rate limits such as the remaining requests, tokens, and other metadata in the headers of the HTTP response.

You can expect to see the following header fields:

| FIELD | SAMPLE VALUE | DESCRIPTION |
|------------------------------------|-----------------|---|
| x-ratelimit-limit- requests | 60 | The maximum number of requests that are permitted before exhausting the rate limit. |
| x-ratelimit-limit-tokens | 150000 | The maximum number of tokens that are permitted before exhausting the rate limit. |
| x-ratelimit-remaining- requests | 59 | The remaining number of requests that are permitted before exhausting the rate limit. |
| x-ratelimit-remaining- tokens | 149984 | The remaining number of tokens that are permitted before exhausting the rate limit. |
| x-ratelimit-reset- requests | 1s | The time until the rate limit (based on requests) resets to its initial state. |
| x-ratelimit-reset-tokens | 6m0s | The time until the rate limit (based on tokens) resets to its initial state. |

Error Mitigation

What are some steps I can take to mitigate this?

The OpenAI Cookbook has a <u>Python notebook</u> that explains how to avoid rate limit errors, as well an example <u>Python script</u> for staying under rate limits while batch processing API requests.

You should also exercise caution when providing programmatic access, bulk processing features, and automated social media posting - consider only enabling these for trusted customers.

To protect against automated and high-volume misuse, set a usage limit for individual users within a specified time frame (daily, weekly, or monthly). Consider implementing a hard cap or a manual review process for users who exceed the limit.

Retrying with exponential backoff

One easy way to avoid rate limit errors is to automatically retry requests with a random exponential backoff. Retrying with exponential backoff means performing a short sleep when a rate limit error is hit, then retrying the unsuccessful request. If the request is still unsuccessful, the sleep length is increased and the process is repeated. This continues until the request is successful or until a maximum number of retries is reached. This approach has many benefits:

- Automatic retries means you can recover from rate limit errors without crashes or missing data
- Exponential backoff means that your first retries can be tried quickly, while still benefiting from longer delays if your first few retries fail
- Adding random jitter to the delay helps retries from all hitting at the same time.

Note that unsuccessful requests contribute to your per-minute limit, so continuously resending a request won't work.

Below are a few example solutions for Python that use exponential backoff.

Example 1: Using the Tenacity library

Tenacity is an Apache 2.0 licensed general-purpose retrying library, written in Python, to simplify the task of adding retry behavior to just about anything. To add exponential backoff to your requests, you can use the tenacity.retry decorator. The below example uses

the tenacity.wait_random_exponential function to add random exponential backoff to a request.

```
Using the Tenacity library
python
Copy
1
2
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13
14
from openai import OpenAI
client = OpenAI()
from tenacity import (
  retry,
  stop_after_attempt,
  wait_random_exponential,
) # for exponential backoff
@retry(wait=wait_random_exponential(min=1, max=60), stop=stop_after_attempt(6))
def completion_with_backoff(**kwargs):
  return client.completions.create(**kwargs)
completion_with_backoff(model="gpt-3.5-turbo-instruct", prompt="Once upon a time,")
Note that the Tenacity library is a third-party tool, and OpenAI makes no guarantees about its reliability
or security.
Collapse
```

Example 2: Using the backoff library

Another python library that provides function decorators for backoff and retry is backoff:

```
Using the Tenacity library python
Copy
1
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6
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8
9
```

```
10
import backoff
import openai
from openai import OpenAI
client = OpenAI()
@backoff.on_exception(backoff.expo, openai.RateLimitError)
def completions_with_backoff(**kwargs):
  return client.completions.create(**kwargs)
```

completions_with_backoff(model="gpt-3.5-turbo-instruct", prompt="Once upon a time,") Like Tenacity, the backoff library is a third-party tool, and OpenAI makes no guarantees about its reliability or security.

Collapse

Example 3: Manual backoff implementation

If you don't want to use third-party libraries, you can implement your own backoff logic following this example:

Using manual backoff implementation

python

Copy

```
42
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# imports
import random
import time
import openai
from openai import OpenAI
client = OpenAI()
# define a retry decorator
def retry_with_exponential_backoff(
  func,
  initial_delay: float = 1,
  exponential_base: float = 2,
  jitter: bool = True,
  max_retries: int = 10,
  errors: tuple = (openai.RateLimitError,),
):
  """Retry a function with exponential backoff."""
  def wrapper(*args, **kwargs):
     # Initialize variables
     num_retries = 0
     delay = initial_delay
     # Loop until a successful response or max_retries is hit or an exception is raised
     while True:
       try:
          return func(*args, **kwargs)
       # Retry on specific errors
       except errors as e:
          # Increment retries
          num_retries += 1
          # Check if max retries has been reached
          if num_retries > max_retries:
            raise Exception(
               f"Maximum number of retries ({max_retries}) exceeded."
          # Increment the delay
          delay *= exponential_base * (1 + jitter * random.random())
          # Sleep for the delay
          time.sleep(delay)
       # Raise exceptions for any errors not specified
       except Exception as e:
          raise e
  return wrapper
```

```
@retry_with_exponential_backoff
def completions_with_backoff(**kwargs):
    return client.completions.create(**kwargs)
```

Again, OpenAI makes no guarantees on the security or efficiency of this solution but it can be a good starting place for your own solution.

Collapse

Reduce the max_tokens to match the size of your completions

Your rate limit is calculated as the maximum of max_tokens and the estimated number of tokens based on the character count of your request. Try to set the max_tokens value as close to your expected response size as possible.

Batching requests

The OpenAI API has separate limits for requests per minute and tokens per minute.

If you're hitting the limit on requests per minute, but have available capacity on tokens per minute, you can increase your throughput by batching multiple tasks into each request. This will allow you to process more tokens per minute, especially with our smaller models.

Sending in a batch of prompts works exactly the same as a normal API call, except you pass in a list of strings to the prompt parameter instead of a single string.

Example without batching

```
No batching
python
Copy
2
3
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12
13
14
from openai import OpenAI
client = OpenAI()
num_stories = 10
prompt = "Once upon a time,"
# serial example, with one story completion per request
for _ in range(num_stories):
  response = client.completions.create(
    model="curie",
    prompt=prompt,
    max_tokens=20,
  # print story
  print(prompt + response.choices[0].text)
Collapse
```

Example with batching

Batching python

```
Copy
2
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14
15
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17
18
19
20
21
from openai import OpenAI
client = OpenAI()
num_stories = 10
prompts = ["Once upon a time,"] * num_stories
# batched example, with 10 story completions per request
response = client.completions.create(
  model="curie",
  prompt=prompts,
  max_tokens=20,
# match completions to prompts by index
stories = [""] * len(prompts)
for choice in response.choices:
  stories[choice.index] = prompts[choice.index] + choice.text
# print stories
for story in stories:
  print(story)
Collapse
```

From < https://platform.openai.com/docs/guides/rate-limits?context=tier-free >

Error codes

This guide includes an overview on error codes you might see from both the <u>API</u> and our <u>official Python library</u>. Each error code mentioned in the overview has a dedicated section with further guidance.

API errors

CODE OVERVIEW

401 - Invalid Authentication Cause: Invalid Authentication

Solution: Ensure the correct API key and

requesting organization are being used.

401 - Incorrect API key provided **Cause:** The requesting API key is not correct.

> **Solution:** Ensure the API key used is correct, clear your browser cache, or generate a new one.

401 - You must be a member of an Cause: Your account is not part of an

organization to use the API organization.

> **Solution:** Contact us to get added to a new organization or ask your organization manager

to invite you to an organization.

429 - Rate limit reached for **Cause:** You are sending requests too quickly.

Solution: Pace your requests. Read the Rate limit

guide.

429 - You exceeded your current Cause: You have run out of credits or hit your

quota, please check your plan and maximum monthly spend. billing details

Solution: Buy more credits or learn how

to increase your limits.

500 - The server had an error while **Cause:** Issue on our servers.

processing your request **Solution:** Retry your request after a brief wait and

contact us if the issue persists. Check the status

page.

503 - The engine is currently **Cause:** Our servers are experiencing high traffic. overloaded, please try again later **Solution:** Please retry your requests after a brief

wait.

401 - Invalid Authentication

This error message indicates that your authentication credentials are invalid. This could happen for several reasons, such as:

- You are using a revoked API key.
- You are using a different API key than the one assigned to the requesting organization.
- You are using an API key that does not have the required permissions for the endpoint you are calling.

To resolve this error, please follow these steps:

- Check that you are using the correct API key and organization ID in your request header. You can find your API key and organization ID in your account settings.
- If you are unsure whether your API key is valid, you can generate a new one. Make sure to replace your old API key with the new one in your requests and follow our best practices guide.

Collapse

requests

401 - Incorrect API key provided

This error message indicates that the API key you are using in your request is not correct. This could happen for several reasons, such as:

- There is a typo or an extra space in your API key.
- You are using an API key that belongs to a different organization.
- You are using an API key that has been deleted or deactivated.
- An old, revoked API key might be cached locally.

To resolve this error, please follow these steps:

- Try clearing your browser's cache and cookies, then try again.
- Check that you are using the correct API key in your request header.
- If you are unsure whether your API key is correct, you can <u>generate a new one</u>.
 Make sure to replace your old API key in your codebase and follow our <u>best</u> practices guide.

Collapse

401 - You must be a member of an organization to use the API

This error message indicates that your account is not part of an organization. This could happen for several reasons, such as:

- You have left or been removed from your previous organization.
- Your organization has been deleted.

To resolve this error, please follow these steps:

- If you have left or been removed from your previous organization, you can either request a new organization or get invited to an existing one.
- To request a new organization, reach out to us via help.openai.com
- Existing organization owners can invite you to join their organization via the <u>Team page</u>.

Collapse

429 - Rate limit reached for requests

This error message indicates that you have hit your assigned rate limit for the API. This means that you have submitted too many tokens or requests in a short period of time and have exceeded the number of requests allowed. This could happen for several reasons, such as:

- You are using a loop or a script that makes frequent or concurrent requests.
- You are sharing your API key with other users or applications.
- You are using a free plan that has a low rate limit.

To resolve this error, please follow these steps:

- Pace your requests and avoid making unnecessary or redundant calls.
- If you are using a loop or a script, make sure to implement a backoff mechanism or a retry logic that respects the rate limit and the response headers. You can read more about our rate limiting policy and best practices in our <u>rate limit guide</u>.
- If you are sharing your organization with other users, note that limits are applied
 per organization and not per user. It is worth checking on the usage of the rest of
 your team as this will contribute to the limit.
- If you are using a free or low-tier plan, consider upgrading to a pay-as-you-go plan that offers a higher rate limit. You can compare the restrictions of each plan in our <u>rate limit guide</u>.

Collapse

429 - You exceeded your current quota, please check your plan and billing details

This error message indicates that you hit your monthly [usage limit](/account/limits) for the API, or for prepaid credits customers that you've consumed all your credits. You can view your maximum usage limit on the [limits page](/account/limits). This could happen for several reasons, such as:

 You are using a high-volume or complex service that consumes a lot of credits or tokens.

- Your monthly budget is set too low for your organization's usage. To resolve this error, please follow these steps:
 - Check your current usage of your account, and compare that to your account's limits.
- If you are on a free plan, consider upgrading to a paid plan to get higher limits. Collapse

503 - The engine is currently overloaded, please try again later

This error message indicates that our servers are experiencing high traffic and are unable to process your request at the moment. This could happen for several reasons, such as:

- There is a sudden spike or surge in demand for our services.
- There is scheduled or unscheduled maintenance or update on our servers.
- There is an unexpected or unavoidable outage or incident on our servers.

To resolve this error, please follow these steps:

- Retry your request after a brief wait. We recommend using an exponential backoff strategy or a retry logic that respects the response headers and the rate limit. You can read more about our rate limit best practices.
- Check our status page for any updates or announcements regarding our services and servers.
- If you are still getting this error after a reasonable amount of time, please contact us for further assistance. We apologize for any inconvenience and appreciate your patience and understanding.

Collapse

Python library error types

TYPE **OVERVIEW**

APIError Cause: Issue on our side.

Solution: Retry your request after a brief wait and contact us if the issue

persists.

Timeout Cause: Request timed out.

Solution: Retry your request after a brief wait and contact us if the issue

persists.

RateLimitE Cause: You have hit your assigned rate limit.

rror **Solution:** Pace your requests. Read more in our Rate limit guide.

APIConne Cause: Issue connecting to our services.

ctionError **Solution**: Check your network settings, proxy configuration, SSL

certificates, or firewall rules.

InvalidReg Cause: Your request was malformed or missing some required uestError

parameters, such as a token or an input.

Solution: The error message should advise you on the specific error made. Check the documentation for the specific API method you are calling and make sure you are sending valid and complete parameters. You may also need to check the encoding, format, or size of your request

data.

Authentica Cause: Your API key or token was invalid, expired, or revoked.

tionError Solution: Check your API key or token and make sure it is correct and

active. You may need to generate a new one from your account

dashboard.

ServiceUn Cause: Issue on our servers.

availableEr Solution: Retry your request after a brief wait and contact us if the issue

ror persists. Check the status page.

APIError

An `APIError` indicates that something went wrong on our side when processing your request. This could be due to a temporary error, a bug, or a system outage. We apologize for any inconvenience and we are working hard to resolve any issues as soon as possible. You can check our system status page for more information.

If you encounter an APIError, please try the following steps:

- Wait a few seconds and retry your request. Sometimes, the issue may be resolved quickly and your request may succeed on the second attempt.
- Check our status page for any ongoing incidents or maintenance that may affect our services. If there is an active incident, please follow the updates and wait until it is resolved before retrying your request.
- If the issue persists, check out our Persistent errors next steps section.

Our support team will investigate the issue and get back to you as soon as possible. Note that our support queue times may be long due to high demand. You can also <u>post</u> in our Community Forum but be sure to omit any sensitive information.

Collapse

Timeout

A `Timeout` error indicates that your request took too long to complete and our server closed the connection. This could be due to a network issue, a heavy load on our services, or a complex request that requires more processing time.

If you encounter a Timeout error, please try the following steps:

- Wait a few seconds and retry your request. Sometimes, the network congestion or the load on our services may be reduced and your request may succeed on the second attempt.
- Check your network settings and make sure you have a stable and fast internet connection. You may need to switch to a different network, use a wired connection, or reduce the number of devices or applications using your bandwidth.
- If the issue persists, check out our persistent errors next steps section.

Collapse

RateLimitError

A `RateLimitError` indicates that you have hit your assigned rate limit. This means that you have sent too many tokens or requests in a given period of time, and our services have temporarily blocked you from sending more.

We impose rate limits to ensure fair and efficient use of our resources and to prevent abuse or overload of our services.

If you encounter a RateLimitError, please try the following steps:

- Send fewer tokens or requests or slow down. You may need to reduce the frequency or volume of your requests, batch your tokens, or implement exponential backoff. You can read our <u>Rate limit guide</u> for more details.
- Wait until your rate limit resets (one minute) and retry your request. The error message should give you a sense of your usage rate and permitted usage.

You can also check your API usage statistics from your account dashboard.
 Collapse

APIConnectionError

An `APIConnectionError` indicates that your request could not reach our servers or establish a secure connection. This could be due to a network issue, a proxy configuration, an SSL certificate, or a firewall rule.

If you encounter an APIConnectionError, please try the following steps:

- Check your network settings and make sure you have a stable and fast internet connection. You may need to switch to a different network, use a wired connection, or reduce the number of devices or applications using your bandwidth.
- Check your proxy configuration and make sure it is compatible with our services.
 You may need to update your proxy settings, use a different proxy, or bypass the proxy altogether.
- Check your SSL certificates and make sure they are valid and up-to-date. You may need to install or renew your certificates, use a different certificate authority, or disable SSL verification.
- Check your firewall rules and make sure they are not blocking or filtering our services. You may need to modify your firewall settings.
- If appropriate, check that your container has the correct permissions to send and receive traffic.
- If the issue persists, check out our persistent errors next steps section.

Collapse

InvalidRequestError

An InvalidRequestError indicates that your request was malformed or missing some required parameters, such as a token or an input. This could be due to a typo, a formatting error, or a logic error in your code.

If you encounter an InvalidRequestError, please try the following steps:

- Read the error message carefully and identify the specific error made. The error message should advise you on what parameter was invalid or missing, and what value or format was expected.
- Check the <u>API Reference</u> for the specific API method you were calling and make sure you are sending valid and complete parameters. You may need to review the parameter names, types, values, and formats, and ensure they match the documentation.
- Check the encoding, format, or size of your request data and make sure they are compatible with our services. You may need to encode your data in UTF-8, format your data in JSON, or compress your data if it is too large.
- Test your request using a tool like Postman or curl and make sure it works as expected. You may need to debug your code and fix any errors or inconsistencies in your request logic.
- If the issue persists, check out our persistent errors next steps section.
 Collapse

AuthenticationError

An `AuthenticationError` indicates that your API key or token was invalid, expired, or revoked. This could be due to a typo, a formatting error, or a security breach. If you encounter an AuthenticationError, please try the following steps:

- Check your API key or token and make sure it is correct and active. You may need
 to generate a new key from the API Key dashboard, ensure there are no extra
 spaces or characters, or use a different key or token if you have multiple ones.
- Ensure that you have followed the correct formatting.

Collapse

ServiceUnavailableError

A `ServiceUnavailableError` indicates that our servers are temporarily unable to handle your request. This could be due to a planned or unplanned maintenance, a system upgrade, or a server failure. These errors can also be returned during periods of high traffic.

We apologize for any inconvenience and we are working hard to restore our services as soon as possible.

If you encounter a ServiceUnavailableError, please try the following steps:

- Wait a few minutes and retry your request. Sometimes, the issue may be resolved quickly and your request may succeed on the next attempt.
- Check our <u>status page</u> for any ongoing incidents or maintenance that may affect our services. If there is an active incident, please follow the updates and wait until it is resolved before retrying your request.
- If the issue persists, check out our persistent errors next steps section.

Collapse

Persistent errors

If the issue persists, <u>contact our support team via chat</u> and provide them with the following information:

- The model you were using
- The error message and code you received
- The request data and headers you sent
- The timestamp and timezone of your request
- Any other relevant details that may help us diagnose the issue

Our support team will investigate the issue and get back to you as soon as possible. Note that our support queue times may be long due to high demand. You can also <u>post in our Community Forum</u> but be sure to omit any sensitive information.

Handling errors

We advise you to programmatically handle errors returned by the API. To do so, you may want to use a code snippet like below:

```
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import openai
from openai import OpenAI
client = OpenAI()
try:
 #Make your OpenAl API request here
 response = client.completions.create(
  prompt="Hello world",
  model="gpt-3.5-turbo-instruct"
 )
except openai.APIError as e:
 #Handle API error here, e.g. retry or log
 print(f"OpenAl API returned an API Error: {e}")
 pass
except openai.APIConnectionError as e:
 #Handle connection error here
 print(f"Failed to connect to OpenAl API: {e}")
 pass
except openai.RateLimitError as e:
 #Handle rate limit error (we recommend using exponential backoff)
 print(f"OpenAl API request exceeded rate limit: {e}")
 pass
```

From < https://platform.openai.com/docs/guides/error-codes

Libraries

Python library

We provide a Python library, which you can install as follows:

```
$ pip install openai
```

Once installed, you can use the bindings and your secret key to run the following:

1

```
3
4
5
6
7
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9
10
from openai import OpenAI
client = OpenAI(
  # Defaults to os.environ.get("OPENAI_API_KEY")
  # Otherwise use: api_key="Your_API_Key",
chat_completion = client.chat.completions.create(
  model="gpt-3.5-turbo",
  messages=[{"role": "user", "content": "Hello world"}]
```

The bindings also will install a command-line utility you can use as follows:

\$ openai api chat_completions.create -m gpt-3.5-turbo -g user "Hello world"

Node.js library

We also have a Node.js library, which you can install by running the following command in your Node.js project directory:

\$ npm install openai

Once installed, you can use the library and your secret key to run the following:

```
1
2
3
4
5
6
7
8
9
10
import OpenAI from "openai";
const openai = new OpenAI({
  apiKey: process.env.OPENAI_API_KEY,
});
const chatCompletion = await openai.chat.completions.create({
  messages: [{ role: "user", content: "Say this is a test" }],
  model: "gpt-3.5-turbo",
});
```

Azure OpenAl libraries

Microsoft's Azure team maintains libraries that are compatible with both the OpenAl API and Azure OpenAl services. Read the library documentation below to learn how you can use them with the OpenAl API.

- Azure OpenAl client library for .NET
- Azure OpenAl client library for JavaScript
- Azure OpenAl client library for Java
- Azure OpenAl client library for Go

Community libraries

The libraries below are built and maintained by the broader developer community. If you'd like to add a new library here, please follow the instructions in our help-center-article on adding community libraries. You can also watch-our-OpenAPI-specification repository on GitHub to get timely updates on when we make changes to our API.

Please note that OpenAl does not verify the correctness or security of these projects. **Use them at your own risk!**

C# / .NET

- Betalgo.OpenAl by Betalgo
- OpenAI-API-dotnet by OkGoDolt
- OpenAI-DotNet by RageAgainstThePixel

C++

• <u>liboai</u> by <u>D7EAD</u>

<u>Clojure</u>

openai-clojure by wkok

Crystal

• openai-crystal by sferik

Dart/Flutter

openai by anasfik

<u>Delphi</u>

• <u>DelphiOpenAl</u> by <u>HemulGM</u>

<u>Elixir</u>

openai.ex by mgallo

<u>Go</u>

• go-gpt3 by sashabaranov

Java

openai-java by Theo Kanning

Julia

• OpenAl.jl by rory-linehan

<u>Kotlin</u>

openai-kotlin by Mouaad Aallam

Node.js

- openai-api by Njerschow
- openai-api-node by erlapso
- gpt-x by ceifa
- gpt3 by poteat
- gpts by thence
- @dalenguyen/openai by dalenguyen
- tectalic/openai by tectalic

PHP

- <u>orhanerday/open-ai</u> by <u>orhanerday</u>
- tectalic/openai by tectalic
- openai-php clinet by openai-php

Python

• chronology by OthersideAl

R

• rgpt3 by ben-aaron188

Ruby

- · openai by nileshtrivedi
- <u>ruby-openai</u> by <u>alexrudall</u>

Rust

- async-openai by 64bit
- <u>fieri</u> by <u>lbkolev</u>

Scala

• openai-scala-client by cequence-io

Swift

- OpenAlKit by dylanshine
- OpenAI by MacPaw

<u>Unity</u>

- OpenAi-Api-Unity by hexthedev
- com.openai.unity by RageAgainstThePixel

Unreal Engine

OpenAl-Api-Unreal by KellanM

From < https://platform.openai.com/docs/libraries/community-libraries>

Deprecations

<u>Overview</u>

As we launch safer and more capable models, we regularly retire older models. Software relying on OpenAI models may need occasional updates to keep working. Impacted customers will always be notified by email and in our documentation along

with <u>blog posts</u> for larger changes.

This page lists all API deprecations, along with recommended replacements.

Incremental model updates

As <u>announced</u> in March 2023, we regularly release new versions of gpt-4 and gpt-3.5-turbo.

Each model version is dated with an -MMDD suffix; e.g., gpt-4-0613. The undated model name, e.g., gpt-4, will typically point to the latest version (e.g. gpt-4 points to gpt-4-0613). Users of undated model names will be notified by email typically 2 weeks before any change takes place.

After a new version is launched, older versions will typically be deprecated 3 months later.

Migrating to replacements

Once a model is deprecated, be sure to migrate all usage to a suitable replacement before the shutdown date. Requests to models past the shutdown date will fail.

To help measure the performance of replacement models on your tasks, we have open-sourced <u>Evals</u>, a Python framework for evaluating language models.

If new models perform worse on your tasks, let us know by by submitting a <u>pull</u> request to our Evals repo with examples of the task.

Deprecation history

All deprecations are listed below, with the most recent announcements at the top.

2023-11-06: Chat model updates

On November 6th, 2023, we <u>announced</u> the release of an updated GPT-3.5-Turbo model (which now comes by default with 16k context) along with deprecation of gpt-3.5-turbo-0613 and gpt-3.5-turbo-16k-0613.

| SHUTDOWN DATE | LEGACY MODEL | LEGACY MODEL PRICE | RECOMMENDED REPLACEMENT |
|------------------|----------------------------|--|----------------------------|
| 2024-06-13 | gpt-3.5- turbo-0613 | \$0.0015 / 1K input tokens + \$0.0020 / 1K output tokens | gpt-3.5-turbo-1106 |
| 2024-06-13 | gpt-3.5- turbo-16k-0613 | \$0.0030 / 1K input tokens + \$0.0040 / 1K output tokens | gpt-3.5-turbo-1106 |

Fine-tuned models created from these base models are not effected by this deprecation, but you will no longer be able to create new fine-tuned versions with these models.

2023-08-22: Fine-tunes endpoint

On August 22nd, 2023, we <u>announced</u> the new fine-tuning API (/v1/fine_tuning/jobs) and that the original /v1/fine-tunes API along with legacy models (including those fine-tuned with the /v1/fine-tunes API) will be shut down on January 04, 2024.

Fine-tunes endpoint

2024-01-04 /v1/fine-tunes /v1/fine_tuning/jobs

2023-07-06: GPT and embeddings

On July 06, 2023, we <u>announced</u> the upcoming retirements of older GPT-3 and GPT-3.5 models served via the completions endpoint. We also announced the upcoming retirement of our first-generation text embedding models. They will be shut down on January 04, 2024.

InstructGPT models

| SHUTDOWN DATE | LEGACY MODEL | LEGACY MODEL PRICE | RECOMMENDED REPLACEMENT |
|---------------|------------------|----------------------|-------------------------|
| 2024-01-04 | text-ada-001 | \$0.0004 / 1K tokens | gpt-3.5-turbo-instruct |
| 2024-01-04 | text-babbage-001 | \$0.0005 / 1K tokens | gpt-3.5-turbo-instruct |
| 2024-01-04 | text-curie-001 | \$0.0020 / 1K tokens | gpt-3.5-turbo-instruct |
| 2024-01-04 | text-davinci-001 | \$0.0200 / 1K tokens | gpt-3.5-turbo-instruct |
| 2024-01-04 | text-davinci-002 | \$0.0200 / 1K tokens | gpt-3.5-turbo-instruct |
| 2024-01-04 | text-davinci-003 | \$0.0200 / 1K tokens | gpt-3.5-turbo-instruct |

Pricing for the replacement gpt-3.5-turbo-instruct model can be found on the <u>pricing page</u>.

Base GPT models

| SHUTDOWN DATE | LEGACY MODEL | LEGACY MODEL PRICE | RECOMMENDED REPLACEMENT |
|---------------|------------------|----------------------|-------------------------|
| 2024-01-04 | ada | \$0.0004 / 1K tokens | babbage-002 |
| 2024-01-04 | babbage | \$0.0005 / 1K tokens | babbage-002 |
| 2024-01-04 | curie | \$0.0020 / 1K tokens | davinci-002 |
| 2024-01-04 | davinci | \$0.0200 / 1K tokens | davinci-002 |
| 2024-01-04 | code-davinci-002 | free to researchers | gpt-3.5-turbo-base |

Pricing for the replacement babbage-002 and davinci-002 models can be found on the <u>pricing page</u>.

Edit models & endpoint

| SHUTDOWN DATE | MODEL / SYSTEM | RECOMMENDED REPLACEMENT |
|---------------|-----------------------|-------------------------|
| 2024-01-04 | text-davinci-edit-001 | gpt-4 |
| 2024-01-04 | code-davinci-edit-001 | gpt-4 |
| 2024-01-04 | /v1/edits | /v1/chat/completions |

Fine-tuning GPT models

| SHUTDOWN DATE | LEGACY MODEL | TRAINING PRICE | USAGE PRICE | RECOMMENDED REPLACEMENT |
|------------------|-----------------|-------------------------|-------------------------|----------------------------|
| 2024-01-04 | ada | \$0.0004 / 1K tokens | \$0.0016 / 1K tokens | babbage-002 |
| 2024-01-04 | babbage | \$0.0006 / 1K tokens | \$0.0024 / 1K tokens | babbage-002 |
| 2024-01-04 | curie | \$0.003 / 1K tokens | \$0.012 / 1K tokens | davinci-002 |
| 2024-01-04 | davinci | \$0.03 / 1K | \$0.12 / 1K | davinci-002, gpt-3.5- |

turbo, gpt-4 tokens tokens

First-generation text embedding models

| SHUTDOWN DATE | LEGACY MODEL | LEGACY MODEL PRICE | RECOMMENDED REPLACEMENT |
|------------------|-----------------------------------|------------------------|----------------------------|
| 2024-01-04 | text-similarity-ada-001 | \$0.004 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | text-search-ada-doc-001 | \$0.004 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | text-search-ada-query-001 | \$0.004 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | code-search-ada-code-001 | \$0.004 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | code-search-ada-text-001 | \$0.004 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | text-similarity-babbage-001 | \$0.005 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | text-search-babbage- doc-001 | \$0.005 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | text-search-babbage- query-001 | \$0.005 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | code-search-babbage- code-001 | \$0.005 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | code-search-babbage- text-001 | \$0.005 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | text-similarity-curie-001 | \$0.020 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | text-search-curie-doc-001 | \$0.020 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | text-search-curie-query-001 | \$0.020 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | text-similarity-davinci-001 | \$0.200 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | text-search-davinci-doc-001 | \$0.200 / 1K tokens | text-embedding- ada-002 |
| 2024-01-04 | text-search-davinci- query-001 | \$0.200 / 1K tokens | text-embedding- ada-002 |

2023-06-13: Updated chat modelsOn June 13, 2023, we announced new chat model versions in the <u>Function calling and</u> other API updates blog post. The three original versions will be retired in June 2024 at the earliest.

| SHUTDOWN DATE | LEGACY MODEL | LEGACY MODEL PRICE | RECOMMENDED REPLACEMENT |
|---------------------------|------------------------|--|-------------------------|
| at earliest 2024-06-13 | gpt-3.5- turbo-0301 | \$0.0015 / 1K input tokens + \$0.0020 / 1K output tokens | gpt-3.5-turbo-0613 |

| at earliest 2024-06-13 | gpt-4-0314 | \$0.03 / 1K input tokens + \$0.06 / 1K output tokens | gpt-4-0613 |
|---------------------------|--------------------|--|----------------|
| at earliest 2024-06-13 | gpt-4-32k-03 14 | \$0.06 / 1K input tokens + \$0.12 / 1K output tokens | gpt-4-32k-0613 |

2023-03-20: Codex models

| SHUTDOWN DATE | LEGACY MODEL | RECOMMENDED REPLACEMENT |
|---------------|------------------|------------------------------------|
| 2023-03-23 | code-davinci-002 | gpt-4 or researcher access program |
| 2023-03-23 | code-davinci-001 | gpt-4 or researcher access program |
| 2023-03-23 | code-cushman-002 | gpt-4 or researcher access program |
| 2023-03-23 | code-cushman-001 | gpt-4 or researcher access program |

2022-06-03: Legacy endpoints

| SHUTDOWN DATE | SYSTEM | RECOMMENDED REPLACEMENT |
|---------------|---------------------|-------------------------|
| 2022-12-03 | /v1/engines | <u>/v1/models</u> |
| 2022-12-03 | /v1/search | View transition guide |
| 2022-12-03 | /v1/classifications | View transition guide |
| 2022-12-03 | /v1/answers | View transition guide |

From < https://platform.openai.com/docs/deprecations>

Chat Plugins

Beta

Learn how to build a plugin that allows ChatGPT to intelligently call your API.

GPTs and custom Actions are here!

We're rolling out custom versions of ChatGPT that you can create for a specific purpose—called GPTs. GPTs are a new way for anyone to create a tailored version of ChatGPT to be more helpful in their daily life, at specific tasks, at work, or at home—and then share that creation with others. We are excited to announce Actions, which build on plugins. Actions take many of the core ideas of plugins while also introducing many new features builders have been asking for.

Introduction

OpenAI plugins connect ChatGPT to third-party applications. These plugins enable ChatGPT to interact with APIs defined by developers, enhancing ChatGPT's capabilities and allowing it to perform a wide range of actions. Plugins enable ChatGPT to do things like:

- Retrieve real-time information; e.g., sports scores, stock prices, the latest news, etc.
- Retrieve knowledge-base information; e.g., company docs, personal notes, etc.
- Assist users with actions; e.g., booking a flight, ordering food, etc.

If you want to have an example running as you read through the documentation and learn more about plugins, you can begin with our <u>plugin quickstart repo</u>.

Plugin developers expose one or more API endpoints, accompanied by a standardized manifest file and an OpenAPI specification. These define the plugin's functionality, allowing ChatGPT to consume the files and make calls to the developer-defined APIs.

The AI model acts as an intelligent API caller. Given an API spec and a natural-language description of when to use the API, the model proactively calls the API to perform actions. For instance, if a user asks, "Where should I stay in Paris for a couple nights?", the model may choose to call a hotel reservation plugin API, receive the API response, and generate a user-facing answer combining the API data and its natural language capabilities.

Over time, we anticipate the system will evolve to accommodate more advanced use cases.

Plugin flow

To build a plugin, it is important to understand the end-to-end flow.

- 1. Create a manifest file and host it at yourdomain.com/.well-known/ai-plugin.json
 - The file includes metadata about your plugin (name, logo, etc.), details about authentication required (type of auth, OAuth URLs, etc.), and an OpenAPI spec for the endpoints you want to expose.
 - The model will see the OpenAPI description fields, which can be used to provide a natural language description for the different fields.
 - We suggest exposing only 1-2 endpoints in the beginning with a minimum number of parameters to minimize the length of the text. The plugin description, API requests, and API responses are all inserted into the conversation with ChatGPT. This counts against the context limit of the model.
- 2. Register your plugin in the ChatGPT UI
 - Select the plugin model from the top drop down, then select "Plugins", "Plugin Store", and finally "Develop your own plugin".
 - If authentication is required, provide an OAuth 2 client_id and client_secret or an API key.
- 3. Users activate your plugin
 - Users must manually activate your plugin in the ChatGPT UI. (ChatGPT will not use your plugin by default.)
 - You will be able to share your plugin with 100 additional users (only other developers can install unverified plugins).
 - If OAuth is required, users will be redirected via OAuth to your plugin to sign in
- 4. Users begin a conversation
 - OpenAI will inject a compact description of your plugin in a message to ChatGPT, invisible to end users. This will include the plugin description, endpoints, and examples.
 - When a user asks a relevant question, the model may choose to invoke an API call from your plugin if it seems relevant; for POST requests, we require that developers build a user confirmation flow to avoid destruction actions.
 - The model will incorporate the API call results into its response to the user.
 - The model might include links returned from the API calls in its response.

These will be displayed as rich previews (using the OpenGraph protocol, where we pull the site_name, title, description, image, and url fields).

 The model can also format data from your API in markdown and the ChatGPT UI will render the markdown automatically.

Currently, we will be sending the user's country and state in the Plugin conversation header (if you are in California for example, it would look like {"openai-subdivision-1-isocode": "US-CA"}. This is useful for shopping, restaurants, weather, and more. You can read more in our developer terms of use.

Next steps

Now that you know the basics of plugins, you might want to:

- Get started building a plugin
- Explore example plugins
- Find the right plugin authentication schema
- Read about important steps for productionizing your plugin
- Learn about the plugin review process
- Familiarize yourself with the plugin policies

From < https://platform.openai.com/docs/plugins/introduction>

Actions in GPTs

GPTs and custom Actions are here!

We're rolling out custom versions of ChatGPT that you can create for a specific purpose—called GPTs. GPTs are a new way for anyone to create a tailored version of ChatGPT to be more helpful in their daily life, at specific tasks, at work, or at home—and then share that creation with others. We are excited to announce Actions, which build on plugins. Actions take many of the core ideas of plugins while also introducing many new features builders have been asking for.

What is a GPT?

GPTs provide the ability to deeply customize ChatGPT with all new capabilities. GPTs also lower the barrier for builders. You can read more in the GPT launch blog post.

What is an action?

In addition to using our built-in capabilities, you can also define custom actions by making one or more APIs available to the GPT. Like plugins, actions allow GPTs to integrate external data or interact with the real-world. Connect GPTs to databases, plug them into emails, or make them your shopping assistant. For example, you could integrate a travel listings database, connect a user's email inbox, or facilitate ecommerce orders.

The design of actions builds upon insights from our plugins beta, granting developers greater control over the model and how their APIs are called. Migrating from the plugins beta is easy with the ability to use your existing plugin manifest to define actions for your GPT.

Create an Action

To create an Action, you can define an OpenAPI specification similarly to that of a plugin with a few changes listed below. If you have a plugin today, creating a GPT with an action should only take a few minutes.

You can start by creating a GPT in the ChatGPT UI and then connect it to your existing plugin OpenAPI reference.

From the GPT editor:

- Select "Configure"
- "Add Action"
- Fill in your OpenAPI spec or paste in a URL where it is hosted (you can use an existing plugin URL)

Actions vs Plugins

Like ChatGPT plugins, Actions allow you to connect a GPT to a custom API. There are a few noticeable differences between Actions and plugins which you can see mentioned below.

Functions

Endpoints defined in the OpenAPI specification are now called "functions". There is no difference in how these are defined.

Hosted OpenAPI specification

With Actions, OpenAI now hosts the OpenAPI specification for your API. This means you no longer need to host your own OpenAPI specification. You can import and existing OpenAPI specification or create a new one from scratch using the UI in the GPT creator.

Consequential flag

In the OpenAPI specification, you can now set certain endpoints as "consequential" as shown below:

1 2 3 4 5 6 get: operationId: blah x-openai-isConsequential: false post: operationId: blah2

x-openai-isConsequential: true

- If the x-openai-isConsequential field is true, we treat the operation as "must always prompt the user for confirmation before running" and don't show an "always allow" button (both are new features of GPTs designed to give users more control).
- If the x-openai-isConsequential field is false, we show the "always allow button".

 If the field isn't present, we default all GET operations to false and all other operations to true

Multiple authentication schemas

Actions now support multiple authentication schemas which can be set on a perendpoint basis. This means you can have some endpoints that require authentication and some that don't.

This can be set as a components -> securityschemes -> object in the OpenAPI spec, and on each operation in the spec there will be a security object. If no security object is specified in the operation, we consider it unauthed or noauth.

Updated store process

The GPT marketplace will supersede the plugin store. As the GPT marketplace rolls out, we will have more to share.

From https://platform.openai.com/docs/actions

Introduction

You can interact with the API through HTTP requests from any language, via our official Python bindings, our official Node.js library, or a **community-maintained library**.

To install the official Python bindings, run the following command:

pip install openai

To install the official Node.js library, run the following command in your Node.js project directory:

npm install openai@^4.0.0

Authentication

The OpenAl API uses API keys for authentication. Visit your API Keys page to retrieve the API key you'll use in your requests.

Remember that your API key is a secret! Do not share it with others or expose it in any client-side code (browsers, apps). Production requests must be routed through your own backend server where your API key can be securely loaded from an environment variable or key management service.

All API requests should include your API key in an Authorization HTTP header as follows:

Authorization: Bearer OPENAI_API_KEY

Organization (optional)

For users who belong to multiple organizations, you can pass a header to specify which organization is used for an API request. Usage from these API requests will count as usage for the specified organization.

Example curl command:

```
1
2
3
curl <a href="https://api.openai.com/v1/models">https://api.openai.com/v1/models</a> \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -H "OpenAI-Organization: org-Q13fm7txVz9t6LI45P8FoXBp"
Example with the openai Python package:
1
2
3
4
5
from openai import OpenAl
client = OpenAI(
 organization='org-Q13fm7txVz9t6LI45P8FoXBp',
client.models.list()
Example with the openai Node.js package:
1
2
3
4
5
6
import { Configuration, OpenAlApi } from "openai";
const configuration = new Configuration({
  organization: "org-Q13fm7txVz9t6LI45P8FoXBp",
  apiKey: process.env.OPENAI_API_KEY,
});
const openai = new OpenAlApi(configuration);
const response = await openai.listEngines();
Organization IDs can be found on your Organization settings page.
Making requests
You can paste the command below into your terminal to run your first API request.
Make sure to replace $OPENAI_API_KEY with your secret API key.
1
2
3
4
5
6
```

7 8

curl https://api.openai.com/v1/chat/completions \

```
-H "Content-Type: application/json" \
-H "Authorization: Bearer $OPENAI_API_KEY" \
-d '{
    "model": "gpt-3.5-turbo",
    "messages": [{"role": "user", "content": "Say this is a test!"}],
    "temperature": 0.7
}'
```

This request queries the gpt-3.5-turbo model (which under the hood points to the <u>latest gpt-3.5-turbo model variant</u>) to complete the text starting with a prompt of "Say this is a test". You should get a response back that resembles the following:

```
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
  "id": "chatcmpl-abc123",
  "object": "chat.completion",
  "created": 1677858242,
  "model": "gpt-3.5-turbo-1006",
  "usage": {
     "prompt_tokens": 13,
     "completion_tokens": 7,
     "total_tokens": 20
  },
  "choices": [
       "message": {
          "role" "assistant".
          "content": "\n\nThis is a test!"
       "finish_reason": "stop",
       "index": 0
    }
  ]
```

Now that you've generated your first chat completion, let's break down the <u>response</u> <u>object</u>. We can see the finish_reason is stop which means the API returned the full chat

completion generated by the model without running into any limits. In the choices list, we only generated a single message but you can set the n parameter to generate multiple messages choices.

Audio

Learn how to turn audio into text or text into audio.

Related guide: Speech to text

Create speech

POST https://api.openai.com/v1/audio/speech Generates audio from the input text.

Request body

model

string

Required

One of the available **TTS models**: tts-1 or tts-1-hd

input

string

Required

The text to generate audio for. The maximum length is 4096 characters.

voice

string

Required

The voice to use when generating the audio. Supported voices are alloy, echo, fable, onyx, nova, and shimmer.

response format

string

Optional

Defaults to mp3

The format to audio in. Supported formats are mp3, opus, aac, and flac.

speed

number

Optional

Defaults to 1

The speed of the generated audio. Select a value from 0.25 to 4.0. 1.0 is the default.

Returns

The audio file content.

Example request

```
curl
Copy
2
3
4
5
6
7
8
curl https://api.openai.com/v1/audio/speech \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -H "Content-Type: application/json" \
  "model": "tts-1",
  "input": "The quick brown fox jumped over the lazy dog.",
  "voice": "alloy"
 --output speech.mp3
```

Create transcription

POST https://api.openai.com/v1/audio/transcriptions

Transcribes audio into the input language.

Request body

file

file

Required

The audio file object (not file name) to transcribe, in one of these formats: flac, mp3, mp4, mpeg, mpga, m4a, ogg, wav, or webm.

model

string

Required

ID of the model to use. Only whisper-1 is currently available.

language

string

Optional

The language of the input audio. Supplying the input language in ISO-639-1 format will improve accuracy and latency.

prompt

string

Optional

An optional text to guide the model's style or continue a previous audio segment.

The **prompt** should match the audio language.

response_format

string

Optional

Defaults to json

The format of the transcript output, in one of these options: json, text, srt, verbose_json, or vtt.

temperature

number

Optional

Defaults to 0

The sampling temperature, between 0 and 1. Higher values like 0.8 will make the output more random, while lower values like 0.2 will make it more focused and deterministic. If set to 0, the model will use **log probability** to automatically increase the temperature until certain thresholds are hit.

Returns

The transcribed text.

Example request

```
curl
Copy
1
2
3
4
5
curl <a href="https://api.openai.com/v1/audio/transcriptions">https://api.openai.com/v1/audio/transcriptions</a> \
-H "Authorization: Bearer $OPENAI_API_KEY" \
-H "Content-Type: multipart/form-data" \
-F file="@/path/to/file/audio.mp3" \
-F model="whisper-1"
```

Response

```
Copy
1
2
3
{
```

"text": "Imagine the wildest idea that you've ever had, and you're curious about how it might scale to something that's a 100, a 1,000 times bigger. This is a place where you can get to do that."

Create translation

POST https://api.openai.com/v1/audio/translations

Translates audio into English.

Request body

file

file

Required

The audio file object (not file name) translate, in one of these formats: flac, mp3, mp4, mpeg, mpga, m4a, ogg, wav, or webm.

model

string

Required

ID of the model to use. Only whisper-1 is currently available.

prompt

string

Optional

An optional text to guide the model's style or continue a previous audio segment.

The **prompt** should be in English.

response_format

string

Optional

Defaults to ison

The format of the transcript output, in one of these options: json, text, srt, verbose_json, or vtt.

temperature

number

Optional

Defaults to 0

The sampling temperature, between 0 and 1. Higher values like 0.8 will make the output more random, while lower values like 0.2 will make it more focused and deterministic. If set to 0, the model will use **log probability** to automatically increase the temperature until certain thresholds are hit.

Returns

The translated text.

```
Example request
```

```
curl
Copy
1
2
3
4
curl https://api.openai.com/v1/audio/translations \
 -H "Authorization: Bearer $OPENAL API KEY" \
 -H "Content-Type: multipart/form-data" \
 -F file="@/path/to/file/german.m4a" \
-F model="whisper-1"
Response
Copy
1
2
3
 "text": "Hello, my name is Wolfgang and I come from Germany. Where are you heading today?"
```

Chat

Given a list of messages comprising a conversation, the model will return a response.

Related guide: Chat Completions

The chat completion object

Represents a chat completion response returned by model, based on the provided input.

id

string

A unique identifier for the chat completion.

choices

array

A list of chat completion choices. Can be more than one if n is greater than 1. Show properties

created

integer

The Unix timestamp (in seconds) of when the chat completion was created.

model

string

The model used for the chat completion.

system_fingerprint

string

This fingerprint represents the backend configuration that the model runs with.

Can be used in conjunction with the seed request parameter to understand when backend changes have been made that might impact determinism.

object

string

The object type, which is always chat.completion.

usage

object

Usage statistics for the completion request.

Show properties

The chat completion object

```
Copy
1
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3
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7
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10
11
12
13
14
15
16
17
18
19
20
 "id": "chatcmpl-123",
 "object": "chat.completion",
 "created": 1677652288,
 "model": "gpt-3.5-turbo-0613",
 "system_fingerprint": "fp_44709d6fcb",
 "choices": [{
  "index": 0,
  "message": {
   "role": "assistant",
   "content": "\n\nHello there, how may I assist you today?",
```

```
},
  "finish_reason": "stop"
}],
"usage": {
  "prompt_tokens": 9,
  "completion_tokens": 12,
  "total_tokens": 21
}
```

The chat completion chunk object

Represents a streamed chunk of a chat completion response returned by model, based on the provided input.

id

string

A unique identifier for the chat completion. Each chunk has the same ID.

choices

array

A list of chat completion choices. Can be more than one if n is greater than 1.

Show properties

created

integer

The Unix timestamp (in seconds) of when the chat completion was created. Each chunk has the same timestamp.

model

string

The model to generate the completion.

object

string

```
The object type, which is always chat.completion.chunk.
The chat completion chunk object
Copy
1
2
3
4
5
6
7
8
9
10
11
12
{"id":"chatcmpl-123","object":"chat.completion.chunk", "created":1694268190, "model": "gpt-3.5-turbo-0613",
"system fingerprint": "fp 44709d6fcb",
"choices":[{"index":0,"delta":{"role":"assistant","content":""},"finish_reason":null}]}
{"id":"chatcmpl-123","object":"chat.completion.chunk", "created":1694268190, "model": "gpt-3.5-turbo-0613",
"system_fingerprint": "fp_44709d6fcb", "choices":[{"index":0,"delta":{"content":"Hello"},"finish_reason":null}]}
{"id":"chatcmpl-123","object":"chat.completion.chunk","created":1694268190,"model":"gpt-3.5-turbo-0613",
"system fingerprint": "fp 44709d6fcb", "choices":[{"index":0,"delta":{"content":"!"},"finish reason":null}]}
{"id":"chatcmpl-123","object":"chat.completion.chunk","created":1694268190,"model":"gpt-3.5-turbo-0613",
"system fingerprint": "fp 44709d6fcb", "choices":[{"index":0,"delta":{"content":" today"}, "finish reason":null}]}
{"id":"chatcmpl-123","object":"chat.completion.chunk","created":1694268190,"model":"gpt-3.5-turbo-0613",
"system_fingerprint": "fp_44709d6fcb", "choices":[{"index":0,"delta":{"content":"?"},"finish_reason":null}]}
{"id":"chatcmpl-123","object":"chat.completion.chunk","created":1694268190,"model":"gpt-3.5-turbo-0613",
"system_fingerprint": "fp_44709d6fcb", "choices":[{"index":0,"delta":{},"finish_reason":"stop"}]}
```

Create chat completion

POST https://api.openai.com/v1/chat/completions

Creates a model response for the given chat conversation.

Request body

messages

array

Required

A list of messages comprising the conversation so far. **Example Python code**.

Show possible types

model

string

Required

ID of the model to use. See the **model endpoint compatibility** table for details on which models work with the Chat API.

frequency_penalty

number or null

Optional

Defaults to 0

Number between -2.0 and 2.0. Positive values penalize new tokens based on their existing frequency in the text so far, decreasing the model's likelihood to repeat the same line verbatim.

See more information about frequency and presence penalties.

logit_bias

map

Optional

Defaults to null

Modify the likelihood of specified tokens appearing in the completion.

Accepts a JSON object that maps tokens (specified by their token ID in the tokenizer) to an associated bias value from -100 to 100. Mathematically, the bias is added to the logits generated by the model prior to sampling. The exact effect will vary per model, but values between -1 and 1 should decrease or increase likelihood of selection; values like -100 or 100 should result in a ban or exclusive selection of the relevant token.

max_tokens

integer or null

Optional

Defaults to inf

The maximum number of **tokens** to generate in the chat completion.

The total length of input tokens and generated tokens is limited by the model's context length. **Example Python code** for counting tokens.

n

integer or null

Optional

Defaults to 1

How many chat completion choices to generate for each input message.

presence penalty

number or null

Optional

Defaults to 0

Number between -2.0 and 2.0. Positive values penalize new tokens based on whether they appear in the text so far, increasing the model's likelihood to talk about new topics.

See more information about frequency and presence penalties.

response_format

object

Optional

An object specifying the format that the model must output.

Setting to { type: "json_object" } enables JSON mode, which guarantees the message the model generates is valid JSON.

Important: when using JSON mode you **must** still instruct the model to produce JSON yourself via some conversation message, for example via your system message. If you don't do this, the model

may generate an unending stream of whitespace until the generation reaches the token limit, which may take a lot of time and give the appearance of a "stuck" request. Also note that the message content may be partial (i.e. cut off) if finish_reason="length", which indicates the generation exceeded max_tokens or the conversation exceeded the max context length. Show properties

seed

integer or null

Optional

This feature is in Beta. If specified, our system will make a best effort to sample deterministically, such that repeated requests with the same seed and parameters should return the same result. Determinism is not guaranteed, and you should refer to the system_fingerprint response parameter to monitor changes in the backend.

stop

string / array / null

Optional

Defaults to null

Up to 4 sequences where the API will stop generating further tokens.

stream

boolean or null

Optional

Defaults to false

If set, partial message deltas will be sent, like in ChatGPT. Tokens will be sent as data-only **server-sent events** as they become available, with the stream terminated by a data:

[DONE] message. Example Python code.

temperature

number or null

Optional

Defaults to 1

What sampling temperature to use, between 0 and 2. Higher values like 0.8 will make the output more random, while lower values like 0.2 will make it more focused and deterministic.

We generally recommend altering this or top_p but not both.

top_p

number or null

Optional

Defaults to 1

An alternative to sampling with temperature, called nucleus sampling, where the model considers the results of the tokens with top_p probability mass. So 0.1 means only the tokens comprising the top 10% probability mass are considered.

We generally recommend altering this or temperature but not both.

tools

array

Optional

A list of tools the model may call. Currently, only functions are supported as a tool. Use this to provide a list of functions the model may generate JSON inputs for.

Show properties

tool_choice

string or object

Optional

Controls which (if any) function is called by the model. none means the model will not call a function and instead generates a message. auto means the model can pick between generating a message or calling a function. Specifying a particular function via {"type: "function", "function": {"name": "my_function"}} forces the model to call that function.

none is the default when no functions are present. auto is the default if functions are present. Show possible types

user

string

Optional

A unique identifier representing your end-user, which can help OpenAI to monitor and detect abuse. **Learn more**.

function call

Deprecated

string or object Optional

Deprecated in favor of tool_choice.

Controls which (if any) function is called by the model. none means the model will not call a function and instead generates a message. auto means the model can pick between generating a message or calling a function. Specifying a particular function via {"name": "my_function"} forces the model to call that function.

none is the default when no functions are present. `auto`` is the default if functions are present. Show possible types

functions

Deprecated

array

2

Optional

Deprecated in favor of tools.

A list of functions the model may generate JSON inputs for. Show properties

Returns

Returns a **<u>chat completion</u>** object, or a streamed sequence of **<u>chat completion chunk</u>** objects if the request is streamed.

DEFAULTIMAGE INPUTSTREAMINGFUNCTION CALLING

```
Example request
gpt-3.5-turbo
curl
Copy
1
2
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4
5
6
7
8
9
10
11
12
13
14
15
16
curl https://api.openai.com/v1/chat/completions \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAL APL KEY" \
  "model": "gpt-3.5-turbo",
  "messages": [
     "role": "system",
     "content": "You are a helpful assistant."
     "role": "user",
     "content": "Hello!"
Response
Copy
1
```

```
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
 "id": "chatcmpl-123",
 "object": "chat.completion",
 "created": 1677652288,
 "model": "gpt-3.5-turbo-0613",
 "system_fingerprint": "fp_44709d6fcb",
 "choices": [{
  "index": 0,
  "message": {
   "role": "assistant",
   "content": "\n\nHello there, how may I assist you today?",
  "finish_reason": "stop"
 "usage": {
  "prompt_tokens": 9,
  "completion_tokens": 12,
  "total_tokens": 21
```

Completions

Legacy

3

Given a prompt, the model will return one or more predicted completions, and can also return the probabilities of alternative tokens at each position. We recommend most users use our Chat Completions API. <u>Learn more</u>

Related guide: **Legacy Completions**

The completion object

Legacy

Represents a completion response from the API. Note: both the streamed and non-streamed response objects share the same shape (unlike the chat endpoint).

id

string

A unique identifier for the completion.

choices

array

The list of completion choices the model generated for the input prompt. Show properties

created

integer

The Unix timestamp (in seconds) of when the completion was created.

model

string

The model used for completion.

system_fingerprint

string

This fingerprint represents the backend configuration that the model runs with.

Can be used in conjunction with the seed request parameter to understand when backend changes have been made that might impact determinism.

object

string

The object type, which is always "text_completion"

usage

object

Usage statistics for the completion request.

Show properties

The completion object

```
Copy
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8
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10
11
12
13
14
15
16
17
18
19
 "id": "cmpl-uqkvlQyYK7bGYrRHQ0eXlWi7",
 "object": "text_completion",
 "created": 1589478378,
 "model": "gpt-3.5-turbo",
 "choices": [
   "text": "\n\nThis is indeed a test",
   "index": 0,
   "logprobs": null,
   "finish_reason": "length"
  }
 "usage": {
  "prompt_tokens": 5,
  "completion_tokens": 7,
  "total_tokens": 12
```

Create completion

Legacy

POST https://api.openai.com/v1/completions

Creates a completion for the provided prompt and parameters.

Request body

model

string

Required

ID of the model to use. You can use the List models API to see all of your available models, or see

our Model overview for descriptions of them.

prompt

string or array

Required

The prompt(s) to generate completions for, encoded as a string, array of strings, array of tokens, or array of token arrays.

Note that <|endoftext|> is the document separator that the model sees during training, so if a prompt is not specified the model will generate as if from the beginning of a new document.

best of

integer or null

Optional

Defaults to 1

Generates best_of completions server-side and returns the "best" (the one with the highest log probability per token). Results cannot be streamed.

When used with n, best_of controls the number of candidate completions and n specifies how many to return – best_of must be greater than n.

Note: Because this parameter generates many completions, it can quickly consume your token quota. Use carefully and ensure that you have reasonable settings for max_tokens and stop.

echo

boolean or null

Optional

Defaults to false

Echo back the prompt in addition to the completion

frequency_penalty

number or null

Optional

Defaults to 0

Number between -2.0 and 2.0. Positive values penalize new tokens based on their existing frequency in the text so far, decreasing the model's likelihood to repeat the same line verbatim.

See more information about frequency and presence penalties.

logit bias

map

Optional

Defaults to null

Modify the likelihood of specified tokens appearing in the completion.

Accepts a JSON object that maps tokens (specified by their token ID in the GPT tokenizer) to an associated bias value from -100 to 100. You can use this **tokenizer tool** (which works for both GPT-2 and GPT-3) to convert text to token IDs. Mathematically, the bias is added to the logits generated by the model prior to sampling. The exact effect will vary per model, but values between -1 and 1 should decrease or increase likelihood of selection; values like -100 or 100 should result in a ban or exclusive selection of the relevant token.

As an example, you can pass {"50256": -100} to prevent the <|endoftext|> token from being generated.

logprobs

integer or null

Optional

Defaults to null

Include the log probabilities on the logprobs most likely tokens, as well the chosen tokens. For example, if logprobs is 5, the API will return a list of the 5 most likely tokens. The API will always return the logprob of the sampled token, so there may be up to logprobs+1 elements in the response.

The maximum value for logprobs is 5.

max_tokens

integer or null

Optional

Defaults to 16

The maximum number of **tokens** to generate in the completion.

The token count of your prompt plus max_tokens cannot exceed the model's context length. **Example Python code** for counting tokens.

n

integer or null

Optional

Defaults to 1

How many completions to generate for each prompt.

Note: Because this parameter generates many completions, it can quickly consume your token quota. Use carefully and ensure that you have reasonable settings for max_tokens and stop.

presence_penalty

number or null

Optional

Defaults to 0

Number between -2.0 and 2.0. Positive values penalize new tokens based on whether they appear in the text so far, increasing the model's likelihood to talk about new topics.

See more information about frequency and presence penalties.

seed

integer or null

Optional

If specified, our system will make a best effort to sample deterministically, such that repeated requests with the same seed and parameters should return the same result.

Determinism is not guaranteed, and you should refer to the system_fingerprint response parameter to monitor changes in the backend.

stop

string / array / null

Optional

Defaults to null

Up to 4 sequences where the API will stop generating further tokens. The returned text will not contain the stop sequence.

stream

boolean or null

Optional

Defaults to false

Whether to stream back partial progress. If set, tokens will be sent as data-only <u>server-sent</u> <u>events</u> as they become available, with the stream terminated by a data: [DONE] message. <u>Example</u> <u>Python code</u>.

suffix

string or null

Optional

Defaults to null

The suffix that comes after a completion of inserted text.

temperature

number or null

Optional

Defaults to 1

What sampling temperature to use, between 0 and 2. Higher values like 0.8 will make the output more random, while lower values like 0.2 will make it more focused and deterministic.

We generally recommend altering this or top p but not both.

top_p

number or null

Optional

Defaults to 1

An alternative to sampling with temperature, called nucleus sampling, where the model considers the results of the tokens with top_p probability mass. So 0.1 means only the tokens comprising the top 10% probability mass are considered.

We generally recommend altering this or temperature but not both.

user

string

Optional

A unique identifier representing your end-user, which can help OpenAl to monitor and detect abuse. **Learn more**.

Returns

Returns a **completion** object, or a sequence of completion objects if the request is streamed.

NO STREAMINGSTREAMING

```
Example request
gpt-3.5-turbo-instruct
curl
Copy
1
2
3
4
5
6
7
8
9
curl https://api.openai.com/v1/completions \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -d '{
  "model": "gpt-3.5-turbo-instruct",
  "prompt": "Say this is a test",
  "max_tokens": 7,
  "temperature": 0
 }'
Response
gpt-3.5-turbo-instruct
Copy
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
 "id": "cmpl-uqkvlQyYK7bGYrRHQ0eXIWi7",
 "object": "text_completion",
"created": 1589478378,
"model": "gpt-3.5-turbo-instruct",
 "system_fingerprint": "fp_44709d6fcb",
 "choices": [
    "text": "\n\nThis is indeed a test",
    "index": 0,
    "logprobs": null,
    "finish_reason": "length"
  }
 ],
```

```
"usage": {
    "prompt_tokens": 5,
    "completion_tokens": 7,
    "total_tokens": 12
  }
}
```

Embeddings

Get a vector representation of a given input that can be easily consumed by machine learning models and algorithms.

Related guide: **Embeddings**

The embedding object

Represents an embedding vector returned by embedding endpoint.

index

integer

The index of the embedding in the list of embeddings.

embedding

array

The embedding vector, which is a list of floats. The length of vector depends on the model as listed in the **embedding guide**.

object

string

The object type, which is always "embedding".

```
The embedding object
```

```
Copy
1
2
3
4
5
6
7
8
9
10
 "object": "embedding",
 "embedding": [
  0.0023064255,
  -0.009327292,
  .... (1536 floats total for ada-002)
  -0.0028842222,
 "index": 0
```

Create embeddings

POST https://api.openai.com/v1/embeddings

Creates an embedding vector representing the input text.

Request body

input

string or array Required

Input text to embed, encoded as a string or array of tokens. To embed multiple inputs in a single request, pass an array of strings or array of token arrays. The input must not exceed the max input tokens for the model (8192 tokens for text-embedding-ada-002) and cannot be an empty string. **Example Python code** for counting tokens.

model

```
string
Required
```

ID of the model to use. You can use the <u>List models</u> API to see all of your available models, or see our **Model overview** for descriptions of them.

encoding_format

string Optional

Defaults to float

The format to return the embeddings in. Can be either float or base64.

user

string

Optional

A unique identifier representing your end-user, which can help OpenAI to monitor and detect abuse. **Learn more**.

Returns

A list of **embedding** objects.

Example request

```
curl
Copy
2
3
4
5
6
7
curl https://api.openai.com/v1/embeddings \
 -H "Authorization: Bearer $OPENAL APL KEY" \
 -H "Content-Type: application/json" \
 -d '{
  "input": "The food was delicious and the waiter...",
  "model": "text-embedding-ada-002",
  "encoding_format": "float"
 }'
Response
Copy
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
 "object": "list",
 "data": [
    "object": "embedding",
```

Fine-tuning

Manage fine-tuning jobs to tailor a model to your specific training data.

Related guide: Fine-tune models

The fine-tuning job object

The fine_tuning.job object represents a fine-tuning job that has been created through the API.

id

string

The object identifier, which can be referenced in the API endpoints.

created_at

integer

The Unix timestamp (in seconds) for when the fine-tuning job was created.

error

object or null

For fine-tuning jobs that have failed, this will contain more information on the cause of the failure. Show properties

fine_tuned_model

string or null

The name of the fine-tuned model that is being created. The value will be null if the fine-tuning job is still running.

finished_at

integer or null

The Unix timestamp (in seconds) for when the fine-tuning job was finished. The value will be null if the fine-tuning job is still running.

hyperparameters

object

The hyperparameters used for the fine-tuning job. See the **fine-tuning guide** for more details. Show properties

model

string

The base model that is being fine-tuned.

object

string

The object type, which is always "fine_tuning.job".

organization_id

string

The organization that owns the fine-tuning job.

result_files

arrav

The compiled results file ID(s) for the fine-tuning job. You can retrieve the results with the Files API.

status string

The current status of the fine-tuning job, which can be

either validating files, queued, running, succeeded, failed, or cancelled.

trained tokens

integer or null

The total number of billable tokens processed by this fine-tuning job. The value will be null if the fine-tuning job is still running.

training_file

string

The file ID used for training. You can retrieve the training data with the Files API.

validation file

string or null

The file ID used for validation. You can retrieve the validation results with the Files API.

```
The fine-tuning job object
```

```
Copy
1
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10
11
12
13
14
15
16
17
18
19
 "object": "fine_tuning.job",
 "id": "ftjob-abc123",
 "model": "davinci-002",
 "created_at": 1692661014,
 "finished_at": 1692661190,
 "fine_tuned_model": "ft:davinci-002:my-org:custom_suffix:7q8mpxmy",
 "organization_id": "org-123",
 "result_files": [
   "file-abc123"
 "status": "succeeded",
 "validation_file": null,
 "training_file": "file-abc123",
 "hyperparameters": {
   "n_epochs": 4,
 "trained_tokens": 5768
```

Create fine-tuning job

POST https://api.openai.com/v1/fine_tuning/jobs

Creates a job that fine-tunes a specified model from a given dataset.

Response includes details of the enqueued job including job status and the name of the fine-tuned models once complete.

Learn more about fine-tuning

Request body

model

string

Required

The name of the model to fine-tune. You can select one of the supported models.

training_file

string

Required

The ID of an uploaded file that contains training data.

See upload file for how to upload a file.

Your dataset must be formatted as a JSONL file. Additionally, you must upload your file with the purpose fine-tune.

See the **fine-tuning guide** for more details.

hyperparameters

object

Optional

The hyperparameters used for the fine-tuning job.

Show properties

suffix

string or null

Optional

Defaults to null

A string of up to 18 characters that will be added to your fine-tuned model name.

For example, a suffix of "custom-model-name" would produce a model name like ft:gpt-3.5-turbo:openai:custom-model-name:7p4lURel.

validation_file

string or null

Optional

The ID of an uploaded file that contains validation data.

If you provide this file, the data is used to generate validation metrics periodically during fine-tuning. These metrics can be viewed in the fine-tuning results file. The same data should not be present in both train and validation files.

Your dataset must be formatted as a JSONL file. You must upload your file with the purpose fine-

See the **fine-tuning guide** for more details.

Returns

A fine-tuning.job object.

NO HYPERPARAMETERSHYPERPARAMETERSVALIDATION FILE

Example request

```
curl
Copy
1
2
3
4
5
6
curl https://api.openai.com/v1/fine_tuning/jobs \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
  "training_file": "file-BK7bzQj3FfZFXr7DbL6xJwfo",
  "model": "gpt-3.5-turbo"
}'
Response
Copy
1
2
3
4
5
```

```
6
7
8
9
10
11
12
 "object": "fine_tuning.job",
 "id": "ftjob-abc123",
 "model": "gpt-3.5-turbo-0613",
 "created_at": 1614807352,
 "fine_tuned_model": null,
 "organization_id": "org-123",
 "result_files": [],
 "status": "queued",
 "validation_file": null,
 "training_file": "file-abc123",
```

List fine-tuning jobs

GET https://api.openai.com/v1/fine_tuning/jobs List your organization's fine-tuning jobs

Query parameters

after

string

Optional

Identifier for the last job from the previous pagination request.

limit

integer

Optional

Defaults to 20

Number of fine-tuning jobs to retrieve.

Returns

A list of paginated **fine-tuning job** objects.

Example request

```
curl
Copy
1
2
```

curl https://api.openai.com/v1/fine_tuning/jobs?limit=2 \

-H "Authorization: Bearer \$OPENAI_API_KEY"

Response

```
Copy
1
2
3
4
5
6
7
8
9
10
11
```

Retrieve fine-tuning job

GET https://api.openai.com/v1/fine_tuning/jobs/{fine_tuning_job_id} Get info about a fine-tuning job.

Learn more about fine-tuning

Path parameters

```
fine_tuning_job_id
string
Required
The ID of the fine-tuning job.
```

"finished_at": 1692661190,

Returns

The **fine-tuning** object with the given ID.

```
Example request
curl
Copy
2
curl https://api.openai.com/v1/fine_tuning/jobs/ft-AF1WoRqd3aJAHsqc9NY7iL8F \
 -H "Authorization: Bearer $OPENAI_API_KEY"
Response
Copy
1
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3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
 "object": "fine_tuning.job",
 "id": "ftjob-abc123",
 "model": "davinci-002",
 "created_at": 1692661014,
```

Cancel fine-tuning

POST https://api.openai.com/v1/fine_tuning/jobs/{fine_tuning_job_id}/cancel Immediately cancel a fine-tune job.

Path parameters

```
fine_tuning_job_id
string
Required
The ID of the fine-tuning job to cancel.
```

Returns

The cancelled **fine-tuning** object.

```
Example request
curl
Copy
1
2
curl -X POST https://api.openai.com/v1/fine_tuning/jobs/ftjob-abc123/cancel \
 -H "Authorization: Bearer $OPENAL API KEY"
Response
Copy
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
 "object": "fine_tuning.job",
 "id": "ftjob-abc123",
 "model": "gpt-3.5-turbo-0613", 
"created_at": 1689376978,
 "fine_tuned_model": null,
 "organization_id": "org-123",
 "result_files": [],
 "hyperparameters": {
  "n_epochs": "auto"
 },
 "status": "cancelled",
 "validation_file": "file-abc123",
```

```
"training_file": "file-abc123" }
```

The fine-tuning job event object

Fine-tuning job event object

```
id
string
created_at
integer
level
string
message
string
object
The fine-tuning job event object
2
3
4
5
6
7
 "object": "fine_tuning.job.event",
 "id": "ftevent-abc123"
 "created_at": 1677610602,
 "level": "info",
 "message": "Created fine-tuning job"
```

List fine-tuning events

GET https://api.openai.com/v1/fine_tuning/jobs/{fine_tuning_job_id}/events Get status updates for a fine-tuning job.

Path parameters

```
fine_tuning_job_id
```

string

Required

The ID of the fine-tuning job to get events for.

Query parameters

after

string

Optional

Identifier for the last event from the previous pagination request.

limit

integer

Optional

Defaults to 20

Number of events to retrieve.

Returns

A list of fine-tuning event objects.

Example request

curl

Copy

1

curl https://api.openai.com/v1/fine_tuning/jobs/ftjob-abc123/events \

```
-H "Authorization: Bearer $OPENAI_API_KEY"
Response
Copy
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
 "object": "list",
 "data": [
  {
    "object": "fine_tuning.job.event",
   "id": "ft-event-ddTJfwuMVpfLXseO0Am0Gqjm",
   "created_at": 1692407401,
   "level": "info",
   "message": "Fine tuning job successfully completed",
   "data": null,
   "type": "message"
   "object": "fine_tuning.job.event",
   "id": "ft-event-tyiGuB72evQncpH87xe505Sv",
   "created_at": 1692407400,
   "level": "info",
   "message": "New fine-tuned model created: ft:gpt-3.5-turbo:openai::7p4lURel",
   "data": null,
   "type": "message"
  }
 "has_more": true
```

<u>Files</u>

Files are used to upload documents that can be used with features like **Assistants** and **Fine-tuning**.

The File object

The File object represents a document that has been uploaded to OpenAI.

id

string

The file identifier, which can be referenced in the API endpoints.

bytes

integer

The size of the file, in bytes.

created at

integer

The Unix timestamp (in seconds) for when the file was created.

filename

string

The name of the file.

object

string

The object type, which is always file.

purpose

string

The intended purpose of the file. Supported values are fine-tune, fine-tune-results, assistants, and assistants_output.

status

Deprecated

string

Deprecated. The current status of the file, which can be either uploaded, processed, or error.

status_details

Deprecated

string

Deprecated. For details on why a fine-tuning training file failed validation, see the error field on fine tuning.job.

```
The File object
```

```
Copy
1
2
3
4
5
6
7
8
{
"id": "file-BK7bzQj3FfZFXr7DbL6xJwfo",
"object": "file",
"bytes": 120000,
"created_at": 1677610602,
"filename": "salesOverview.pdf",
"purpose": "assistants",
}
```

List files

GET https://api.openai.com/v1/files

Returns a list of files that belong to the user's organization.

Query parameters

purpose

string

Optional

Only return files with the given purpose.

Returns

A list of **File** objects.

Example request

```
curl
```

Copy

1

curl https://api.openai.com/v1/files \

-H "Authorization: Bearer \$OPENAI_API_KEY"

```
Response
Copy
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
 "data": [
    "id": "file-abc123",
    "object": "file",
    "bytes": 175,
    "created_at": 1613677385,
    "filename": "salesOverview.pdf",
    "purpose": "assistants",
    "id": "file-abc123",
    "object": "file",
    "bytes": 140,
    "created at": 1613779121,
    "filename": "puppy.jsonl", "purpose": "fine-tune",
],
"object": "list"
```

Upload file

POST https://api.openai.com/v1/files

Upload a file that can be used across various endpoints/features. The size of all the files uploaded by one organization can be up to 100 GB.

The size of individual files for can be a maximum of 512MB. See the **Assistants Tools guide** to learn more about the types of files supported. The Fine-tuning API only supports .jsonl files.

Please **contact us** if you need to increase these storage limits.

Request body

file

string

Required

The File object (not file name) to be uploaded.

purpose

string

Required

The intended purpose of the uploaded file.

Use "fine-tune" for <u>Fine-tuning</u> and "assistants" for <u>Assistants</u> and <u>Messages</u>. This allows us to validate the format of the uploaded file is correct for fine-tuning.

Returns

The uploaded File object.

```
Example request
curl
Copy
1
2
3
curl <a href="https://api.openai.com/v1/files">https://api.openai.com/v1/files</a> \
-H "Authorization: Bearer $OPENAI_API_KEY" \
 -F purpose="fine-tune" \
-F file="@mydata.jsonl"
Response
Copy
2
3
4
5
6
7
8
 "id": "file-BK7bzQj3FfZFXr7DbL6xJwfo",
 "object": "file",
 "bytes": 120000,
 "created_at": 1677610602,
 "filename": "mydata.jsonl",
 "purpose": "fine-tune",
```

Delete file

DELETE https://api.openai.com/v1/files/{file_id}
Delete a file.

Path parameters

file id

string

Required

The ID of the file to use for this request.

Returns

Deletion status.

```
Example request
curl
Copy
1
2
3
curl <a href="https://api.openai.com/v1/files/file-abc123">https://api.openai.com/v1/files/file-abc123</a>\
-X DELETE \
-H "Authorization: Bearer $OPENAI_API_KEY"
Response
Copy
1
2
```

```
3
4
5
{
"id": "file-abc123",
"object": "file",
"deleted": true
```

Retrieve file

GET https://api.openai.com/v1/files/{file_id}
Returns information about a specific file.

Path parameters

file_id string

Required
The ID of the file to use for this request.

Returns

The **File** object matching the specified ID.

```
Example request
curl
Copy
curl https://api.openai.com/v1/files/file-BK7bzQj3FfZFXr7DbL6xJwfo \
 -H "Authorization: Bearer $OPENAI_API_KEY"
Response
Copy
1
2
3
4
5
6
7
8
 "id": "file-BK7bzQj3FfZFXr7DbL6xJwfo",
 "object": "file",
"bytes": 120000,
 "created_at": 1677610602,
 "filename": "mydata.jsonl",
 "purpose": "fine-tune",
```

Retrieve file content

GET https://api.openai.com/v1/files/{file_id}/content Returns the contents of the specified file.

Path parameters

file_id

string Required

The ID of the file to use for this request.

Returns

The file content.

Example request

curl

```
Copy
1
2
curl <a href="https://api.openai.com/v1/files/file-BK7bzQj3FfZFXr7DbL6xJwfo/content">https://api.openai.com/v1/files/file-BK7bzQj3FfZFXr7DbL6xJwfo/content</a> \
-H "Authorization: Bearer $OPENAI API KEY" > file.jsonI
```

Images

Given a prompt and/or an input image, the model will generate a new image.

Related guide: **Image generation**

The image object

Represents the url or the content of an image generated by the OpenAl API.

b64_json

string

The base64-encoded JSON of the generated image, if response format is b64 json.

url

string

The URL of the generated image, if response_format is url (default).

revised_prompt

string

The prompt that was used to generate the image, if there was any revision to the prompt.

The image object

```
Copy
1
2
3
4
{
    "url": "...",
    "revised_prompt": "..."
}
```

Create image

POST https://api.openai.com/v1/images/generations

Creates an image given a prompt.

Request body

prompt

string

Required

A text description of the desired image(s). The maximum length is 1000 characters for dall-e-2 and 4000 characters for dall-e-3.

model

string

Optional

Defaults to dall-e-2

The model to use for image generation.

n

integer or null

Optional

Defaults to 1

The number of images to generate. Must be between 1 and 10. For dall-e-3, only n=1 is supported.

quality

string

Optional

Defaults to standard

The quality of the image that will be generated. hd creates images with finer details and greater consistency across the image. This param is only supported for dall-e-3.

response_format

string or null

Optional

Defaults to url

The format in which the generated images are returned. Must be one of url or b64_json.

size

string or null

Optional

Defaults to 1024x1024

The size of the generated images. Must be one of 256x256, 512x512, or 1024x1024 for dall-e-2. Must be one of 1024x1024, 1792x1024, or 1024x1792 for dall-e-3 models.

style

string or null

Optional

Defaults to vivid

The style of the generated images. Must be one of vivid or natural. Vivid causes the model to lean towards generating hyper-real and dramatic images. Natural causes the model to produce more natural, less hyper-real looking images. This param is only supported for dall-e-3.

user

string

Optional

A unique identifier representing your end-user, which can help OpenAl to monitor and detect abuse. **Learn more**.

Returns

Returns a list of **image** objects.

Example request

```
curl
Copy
2
3
4
5
6
7
8
9
curl https://api.openai.com/v1/images/generations \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAL API KEY" \
 -d '{
  "model": "dall-e-3",
  "prompt": "A cute baby sea otter",
  "n": 1,
  "size": "1024x1024"
}'
Response
Copy
1
2
3
4
5
6
7
8
9
10
11
 "created": 1589478378,
 "data": [
    "url": "https://..."
  },
```

```
{
    "url": "<u>https://...</u>"
    }
]
```

Create image edit

POST https://api.openai.com/v1/images/edits

Creates an edited or extended image given an original image and a prompt.

Request body

image

string

Required

The image to edit. Must be a valid PNG file, less than 4MB, and square. If mask is not provided, image must have transparency, which will be used as the mask.

prompt

string

Required

A text description of the desired image(s). The maximum length is 1000 characters.

mask

string

Optional

An additional image whose fully transparent areas (e.g. where alpha is zero) indicate where image should be edited. Must be a valid PNG file, less than 4MB, and have the same dimensions as image.

model

string

Optional

Defaults to dall-e-2

The model to use for image generation. Only dall-e-2 is supported at this time.

n

integer or null

Optional

Defaults to 1

The number of images to generate. Must be between 1 and 10.

size

string or null

Optional

Defaults to 1024x1024

The size of the generated images. Must be one of 256x256, 512x512, or 1024x1024.

response format

string or null

Optional

Defaults to url

The format in which the generated images are returned. Must be one of url or b64_json.

user

string

Optional

A unique identifier representing your end-user, which can help OpenAI to monitor and detect abuse. **Learn more**.

Returns

Returns a list of image objects.

Example request

curl

Copy

1

2

3

4 5

```
7
curl https://api.openai.com/v1/images/edits \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -F image="@otter.png" \
 -F mask="@mask.png" \
 -F prompt="A cute baby sea otter wearing a beret" \
 -F n=2 \
 -F size="1024x1024"
Response
Copy
1
2
3
4
5
6
7
8
9
10
11
 "created": 1589478378,
 "data": [
   "url": "https://..."
  },
   "url": "https://..."
  }
]
```

Create image variation

POST https://api.openai.com/v1/images/variations

Creates a variation of a given image.

Request body

image

6

string

Required

The image to use as the basis for the variation(s). Must be a valid PNG file, less than 4MB, and square.

model

string

Optional

Defaults to dall-e-2

The model to use for image generation. Only dall-e-2 is supported at this time.

n

integer or null

Optional

Defaults to 1

The number of images to generate. Must be between 1 and 10. For dall-e-3, only n=1 is supported.

response_format

string or null

Optional

Defaults to url

The format in which the generated images are returned. Must be one of url or b64_json.

size

string or null

Optional

Defaults to 1024x1024

The size of the generated images. Must be one of 256x256, 512x512, or 1024x1024.

user

string Optional

A unique identifier representing your end-user, which can help OpenAl to monitor and detect abuse. **Learn more**.

Returns

Returns a list of **image** objects.

```
Example request
curl
Copy
1
2
3
4
5
curl https://api.openai.com/v1/images/variations \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -F image="@otter.png" \
 -F n=2 \
-F size="1024x1024"
Response
Copy
1
2
3
4
5
6
7
8
9
10
11
 "created": 1589478378,
 "data": [
   "url": "https://..."
  },
   "url": "https://..."
]
```

Models

List and describe the various models available in the API. You can refer to the <u>Models</u> documentation to understand what models are available and the differences between them.

The model object

Describes an OpenAl model offering that can be used with the API.

id

string

The model identifier, which can be referenced in the API endpoints.

created

integer

The Unix timestamp (in seconds) when the model was created.

object

```
The object type, which is always "model".

owned_by
string
The organization that owns the model.

The model object
Copy
1
2
3
4
5
6
{
"id": "davinci",
"object": "model",
"created": 1686935002,
"owned_by": "openai"
}
```

List models

GET https://api.openai.com/v1/models

Lists the currently available models, and provides basic information about each one such as the owner and availability.

Returns

A list of **model** objects.

```
Example request
curl
Copy
1
2
curl <a href="https://api.openai.com/v1/models">https://api.openai.com/v1/models</a> \
 -H "Authorization: Bearer $OPENAI_API_KEY"
Response
Copy
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6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
{
    "object": "list",
```

```
"data": [

{
    "id": "model-id-0",
    "object": "model",
    "created": 1686935002,
    "owned_by": "organization-owner"
},

{
    "id": "model-id-1",
    "object": "model",
    "created": 1686935002,
    "owned_by": "organization-owner",
},

{
    "id": "model-id-2",
    "object": "model",
    "created": 1686935002,
    "owned_by": "openai"
},
],
"object": "list"
```

Retrieve model

GET https://api.openai.com/v1/models/{model}

Retrieves a model instance, providing basic information about the model such as the owner and permissioning.

Path parameters

model

string Required

The ID of the model to use for this request

Returns

The **model** object matching the specified ID.

```
Example request
gpt-3.5-turbo-instruct
curl
Copy
1
curl https://api.openai.com/v1/models/gpt-3.5-turbo-instruct \
 -H "Authorization: Bearer $OPENAI_API_KEY"
Response
gpt-3.5-turbo-instruct
Copy
1
2
3
4
5
6
 "id": "gpt-3.5-turbo-instruct",
 "object": "model",
 "created": 1686935002,
 "owned_by": "openai"
```

Delete fine-tune model

DELETE https://api.openai.com/v1/models/{model}

Delete a fine-tuned model. You must have the Owner role in your organization to delete a model.

Path parameters

```
model
```

string

Required

The model to delete

Returns

Deletion status.

```
Example request
curl
Copy
2
curl <a href="https://api.openai.com/v1/models/ft:gpt-3.5-turbo:acemeco:suffix:abc123">https://api.openai.com/v1/models/ft:gpt-3.5-turbo:acemeco:suffix:abc123</a>
 -X DELETE \
 -H "Authorization: Bearer $OPENAI_API_KEY"
Response
Copy
2
3
4
5
 "id": "ft:gpt-3.5-turbo:acemeco:suffix:abc123",
 "object": "model",
 "deleted": true
```

Moderations

Given a input text, outputs if the model classifies it as violating OpenAl's content policy.

Related guide: Moderations

The moderation object

Represents policy compliance report by OpenAl's content moderation model against a given input.

id

string

The unique identifier for the moderation request.

model

string

The model used to generate the moderation results.

results

array

A list of moderation objects.

Show properties

The moderation object

Copy

1

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7

```
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19
20
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22
23
24
25
26
27
28
29
30
31
32
33
34
35
 "id": "modr-XXXXX",
 "model": "text-moderation-005",
 "results": [
  {
    "flagged": true,
    `accries": {
    "categories": {
     "sexual": false,
     "hate": false,
      "harassment": false,
      "self-harm": false.
      "sexual/minors": false,
     "hate/threatening": false,
      "violence/graphic": false,
      "self-harm/intent": false,
      "self-harm/instructions": false,
     "harassment/threatening": true,
     "violence": true,
    "category_scores": {
     "sexual": 1.2282071e-06,
      "hate": 0.010696256,
      "harassment": 0.29842457,
      "self-harm": 1.5236925e-08,
      "sexual/minors": 5.7246268e-08,
     "hate/threatening": 0.0060676364, "violence/graphic": 4.435014e-06, "self-harm/intent": 8.098441e-10,
      "self-harm/instructions": 2.8498655e-11,
      "harassment/threatening": 0.63055265,
      "violence": 0.99011886,
```

Create moderation

POST https://api.openai.com/v1/moderations Classifies if text violates OpenAl's Content Policy

Request body

input

string or array Required

The input text to classify

model

string

Optional

Defaults to text-moderation-latest

Two content moderations models are available: text-moderation-stable and text-moderation-latest.

The default is text-moderation-latest which will be automatically upgraded over time. This ensures you are always using our most accurate model. If you use text-moderation-stable, we will provide advanced notice before updating the model. Accuracy of text-moderation-stable may be slightly lower than for text-moderation-latest.

Returns

24 25

A **moderation** object.

Example request

```
curl
Copy
2
3
4
5
curl <a href="https://api.openai.com/v1/moderations">https://api.openai.com/v1/moderations</a> \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -d '{
  "input": "I want to kill them."
Response
Copy
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
```

```
26
27
28
29
30
31
32
33
34
35
 "id": "modr-XXXXX",
 "model": "text-moderation-005",
 "results": [
    "flagged": true,
    "categories": {
     "sexual": false,
     "hate": false,
     "harassment": false,
     "self-harm": false,
     "sexual/minors": false,
     "hate/threatening": false,
     "violence/graphic": false,
     "self-harm/intent": false,
     "self-harm/instructions": false,
     "harassment/threatening": true,
     "violence": true,
    },
    "category_scores": {
     "sexual": 1.2282071e-06,
     "hate": 0.010696256,
     "harassment": 0.29842457,
     "self-harm": 1.5236925e-08,
     "sexual/minors": 5.7246268e-08,
     "hate/threatening": 0.0060676364,
     "violence/graphic": 4.435014e-06, 
"self-harm/intent": 8.098441e-10,
     "self-harm/instructions": 2.8498655e-11,
     "harassment/threatening": 0.63055265,
     "violence": 0.99011886,
```

Assistants

Reta

Build assistants that can call models and use tools to perform tasks.

Get started with the Assistants API

The assistant object

Beta

Represents an assistant that can call the model and use tools.

id

string

The identifier, which can be referenced in API endpoints.

object

string

The object type, which is always assistant.

created_at

integer

The Unix timestamp (in seconds) for when the assistant was created.

name

string or null

The name of the assistant. The maximum length is 256 characters.

description

string or null

The description of the assistant. The maximum length is 512 characters.

model

string

ID of the model to use. You can use the <u>List models</u> API to see all of your available models, or see our **Model overview** for descriptions of them.

instructions

string or null

The system instructions that the assistant uses. The maximum length is 32768 characters.

tools

array

A list of tool enabled on the assistant. There can be a maximum of 128 tools per assistant. Tools can be of types code_interpreter, retrieval, or function. Show possible types

file ids

array

A list of <u>file</u> IDs attached to this assistant. There can be a maximum of 20 files attached to the assistant. Files are ordered by their creation date in ascending order.

metadata

map

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

The assistant object

```
Copy
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
 "id": "asst_abc123",
 "obiect": "assistant".
 "created_at": 1698984975,
 "name": "Math Tutor",
 "description": null,
 "model": "gpt-4",
 "instructions": "You are a personal math tutor. When asked a question, write and run Python code to
answer the question.",
 "tools": [
   "type": "code_interpreter"
  }
 ],
 "file_ids": [],
 "metadata": {}
```

Create assistant

Beta

POST https://api.openai.com/v1/assistants

Create an assistant with a model and instructions.

Request body

model

Required

ID of the model to use. You can use the <u>List models</u> API to see all of your available models, or see our <u>Model overview</u> for descriptions of them.

name

string or null

Optional

The name of the assistant. The maximum length is 256 characters.

description

string or null

Optional

The description of the assistant. The maximum length is 512 characters.

instructions

string or null

Optional

The system instructions that the assistant uses. The maximum length is 32768 characters.

tools

array

Optional

Defaults to []

A list of tool enabled on the assistant. There can be a maximum of 128 tools per assistant. Tools can be of types code_interpreter, retrieval, or function.

Show possible types

file ids

array

Optional

Defaults to []

A list of <u>file</u> IDs attached to this assistant. There can be a maximum of 20 files attached to the assistant. Files are ordered by their creation date in ascending order.

metadata

map

Optional

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

Returns

An assistant object.

CODE INTERPRETERFILES

Example request

curl

Copy

1

2

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9 10

curl "https://api.openai.com/v1/assistants" \

-H "Content-Type: application/json" \

-H "Authorization: Bearer \$OPENAI_API_KEY" \

```
-H "OpenAI-Beta: assistants=v1" \
 -d '{
  "instructions": "You are a personal math tutor. When asked a question, write and run Python code to
answer the question.",
  "name": "Math Tutor"
  "tools": [{"type": "code_interpreter"}],
"model": "gpt-4"
Response
Copy
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
 "id": "asst_abc123",
 "object": "assistant",
 "created_at": 1698984975,
 "name": "Math Tutor",
 "description": null,
 "model": "gpt-4",
 "instructions": "You are a personal math tutor. When asked a question, write and run Python code to
answer the question.",
 "tools": [
    "type": "code_interpreter"
],
"file_ids": [],
'adata":
 "metadata": {}
```

Retrieve assistant

Beta

GET https://api.openai.com/v1/assistants/{assistant_id}

Retrieves an assistant.

Path parameters

assistant_id

string Required

The ID of the assistant to retrieve.

Returns

The **assistant** object matching the specified ID.

```
Example request
```

curl

Copy

1

2

```
curl https://api.openai.com/v1/assistants/asst_abc123 \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -H "OpenAI-Beta: assistants=v1"
Response
Copy
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
 "id": "asst_abc123",
 "object": "assistant",
 "created_at": 1699009709,
 "name": "HR Helper",
 "description": null,
 "model": "gpt-4",
 "instructions": "You are an HR bot, and you have access to files to answer employee questions about
company policies.",
 "tools": [
    "type": "retrieval"
],
"file_ids": [
  "file-abc123"
 "metadata": {}
```

Modify assistant

Beta

POST https://api.openai.com/v1/assistants/{assistant_id}

Modifies an assistant.

Path parameters

assistant id

string

Required

The ID of the assistant to modify.

Request body

model

Optional

ID of the model to use. You can use the <u>List models</u> API to see all of your available models, or see our **Model overview** for descriptions of them.

name

string or null Optional The name of the assistant. The maximum length is 256 characters.

description

string or null

Optional

The description of the assistant. The maximum length is 512 characters.

instructions

string or null

Optional

The system instructions that the assistant uses. The maximum length is 32768 characters.

tools

array

Optional

Defaults to []

A list of tool enabled on the assistant. There can be a maximum of 128 tools per assistant. Tools can be of types code_interpreter, retrieval, or function.

Show possible types

file ids

array

Optional

Defaults to []

A list of <u>File</u> IDs attached to this assistant. There can be a maximum of 20 files attached to the assistant. Files are ordered by their creation date in ascending order. If a file was previosuly attached to the list but does not show up in the list, it will be deleted from the assistant.

metadata

map

Optional

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

Returns

The modified assistant object.

Example request

```
curl
Copy
2
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9
curl https://api.openai.com/v1/assistants/asst_abc123 \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -H "OpenAI-Beta: assistants=v1" \
   "instructions": "You are an HR bot, and you have access to files to answer employee questions about
company policies. Always response with info from either of the files.",
   "tools": [{"type": "retrieval"}],
   "model": "gpt-4",
"file_ids": ["file-abc123", "file-abc456"]
  }'
Response
Copy
1
2
3
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```

```
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
 "id": "asst_abc123",
 "object": "assistant",
 "created_at": 1699009709,
 "name": "HR Helper",
 "description": null,
 "model": "gpt-4",
 "instructions": "You are an HR bot, and you have access to files to answer employee questions about
company policies. Always response with info from either of the files.",
 "tools": [
    "type": "retrieval"
 ],
 "file_ids": [
  "file-abc123",
  "file-abc456"
 "metadata": {}
Delete assistant
```

Beta

DELETE https://api.openai.com/v1/assistants/{assistant_id}

Delete an assistant.

Path parameters

```
assistant id
```

string

Required

The ID of the assistant to delete.

Returns

Deletion status

```
Example request
```

```
curl
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```

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curl https://api.openai.com/v1/assistants/asst_abc123 \

```
-H "Content-Type: application/json" \
```

- -H "Authorization: Bearer \$OPENAI_API_KEY" \
- -H "OpenAI-Beta: assistants=v1" \
- -X DELETE

Response

```
Copy
1
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{
"id": "asst_abc123",
"object": "assistant.deleted",
"deleted": true
}
```

List assistants

Beta

GET https://api.openai.com/v1/assistants

Returns a list of assistants.

Query parameters

limit

integer

Optional

Defaults to 20

A limit on the number of objects to be returned. Limit can range between 1 and 100, and the default is 20.

order

string

Optional

Defaults to desc

Sort order by the created_at timestamp of the objects. asc for ascending order and desc for descending order.

after

string

Optional

A cursor for use in pagination. after is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include after=obj_foo in order to fetch the next page of the list.

before

string

Optional

A cursor for use in pagination. before is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include before=obj_foo in order to fetch the previous page of the list.

Returns

A list of **assistant** objects.

Example request

```
curl
```

Сору

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curl "https://api.openai.com/v1/assistants?order=desc&limit=20" \

```
-H "Content-Type: application/json" \
```

-H "Authorization: Bearer \$OPENAI_API_KEY" \

-H "OpenAI-Beta: assistants=v1"

Response

Copy

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29
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33
34
35
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 "object": "list",
 "data": [
    "id": "asst_abc123",
    "object": "assistant",
    "created_at": 1698982736,
    "name": "Coding Tutor",
    "description": null,
    "model": "gpt-4", "instructions": "You are a helpful assistant designed to make me better at coding!",
    "tools": [],
    "file_ids": [],
    "metadata": {}
    "id": "asst_abc456",
    "object": "assistant",
    "created_at": 1698982718,
    "name": "My Assistant",
    "description": null,
    "model": "gpt-4",
    "instructions": "You are a helpful assistant designed to make me better at coding!",
    "tools": [],
    "file_ids": [],
    "metadata": {}
```

```
"id": "asst_abc789",
   "object": "assistant",
   "created_at": 1698982643,
   "name": null,
   "description": null,
   "model": "gpt-4",
   "instructions": null,
   "tools": [],
   "file_ids": [],
   "metadata": {}
  }
 "first_id": "asst_abc123",
 "last_id": "asst_abc789",
 "has_more": false
The assistant file object
Beta
A list of Files attached to an assistant.
id
string
The identifier, which can be referenced in API endpoints.
string
The object type, which is always assistant.file.
created at
integer
The Unix timestamp (in seconds) for when the assistant file was created.
assistant id
string
The assistant ID that the file is attached to.
The assistant file object
Copy
2
3
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5
6
 "id": "file-wB6RM6wHdA49HfS2DJ9fEyrH",
 "object": "assistant.file",
 "created at": 1699055364,
 "assistant_id": "asst_FBOFvAOHhwEWMghbMGseaPGQ"
```

Create assistant file

Beta

POST https://api.openai.com/v1/assistants/{assistant_id}/files

Create an assistant file by attaching a File to an assistant.

Path parameters

assistant_id

string

Required

The ID of the assistant for which to create a File.

Request body

file_id string Required

A <u>File</u> ID (with purpose="assistants") that the assistant should use. Useful for tools like retrieval and code interpreter that can access files.

Returns

An assistant file object.

```
Example request
curl
Copy
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curl https://api.openai.com/v1/assistants/asst_FBOFvAOHhwEWMghbMGseaPGQ/files \
  -H 'Authorization: Bearer $OPENAL APL KEY"' \
  -H 'Content-Type: application/json' \
  -H 'OpenAl-Beta: assistants=v1' \
  -d '{
   "file id": "file-wB6RM6wHdA49HfS2DJ9fEyrH"
Response
Copy
1
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5
6
 "id": "file-wB6RM6wHdA49HfS2DJ9fEyrH",
 "object": "assistant.file",
 "created_at": 1699055364,
 "assistant id": "asst FBOFvAOHhwEWMghbMGseaPGQ"
```

Retrieve assistant file

Beta

GET https://api.openai.com/v1/assistants/{assistant_id}/files/{file_id}

Retrieves an AssistantFile.

Path parameters

```
assistant id
```

string

Required

The ID of the assistant who the file belongs to.

file_id

string

Required

The ID of the file we're getting.

Returns

The **assistant file** object matching the specified ID.

Example request

curl

Copy

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```
curl https://api.openai.com/v1/assistants/asst_FBOFvAOHhwEWMqhbMGseaPGQ/files/file-
wB6RM6wHdA49HfS2DJ9fEyrH \
 -H 'Authorization: Bearer $OPENAI_API_KEY" \
-H 'Content-Type: application/json' \
 -H 'OpenAI-Beta: assistants=v1'
Response
Copy
1
2
3
4
5
6
 "id": "file-wB6RM6wHdA49HfS2DJ9fEyrH",
 "object": "assistant.file",
 "created_at": 1699055364,
 "assistant_id": "asst_FBOFvAOHhwEWMghbMGseaPGQ"
Delete assistant file
DELETE https://api.openai.com/v1/assistants/{assistant_id}/files/{file_id}
Delete an assistant file.
Path parameters
assistant id
string
Required
The ID of the assistant that the file belongs to.
string
Required
The ID of the file to delete.
Returns
Deletion status
Example request
curl
Copy
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https://api.openai.com/v1/assistants/asst_DUGk5I7sK0FpKeijvrO30z9J/files/file-9F1ex49ipEnKzyLUNnCA0
-H 'Authorization: Bearer $OPENAL APL KEY' \
-H 'Content-Type: application/json' \
-H 'OpenAI-Beta: assistants=v1' \
-X DELETE
Response
Copy
2
3
4
5
 id: "file-BK7bzQj3FfZFXr7DbL6xJwfo",
```

```
object: "assistant.file.deleted",
deleted: true
}
```

List assistant files

Beta

GET https://api.openai.com/v1/assistants/{assistant_id}/files

Returns a list of assistant files.

Path parameters

assistant id

string

Required

The ID of the assistant the file belongs to.

Query parameters

limit

integer

Optional

Defaults to 20

A limit on the number of objects to be returned. Limit can range between 1 and 100, and the default is 20.

order

string

Optional

Defaults to desc

Sort order by the created_at timestamp of the objects. asc for ascending order and desc for descending order.

after

string

Optional

A cursor for use in pagination. after is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include after=obj_foo in order to fetch the next page of the list.

before

string

Optional

A cursor for use in pagination. before is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include before=obj_foo in order to fetch the previous page of the list.

Returns

A list of assistant file objects.

Example request

curl

Copy

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curl https://api.openai.com/v1/assistants/asst_DUGk5I7sK0FpKeijvrO30z9J/files \

- -H 'Authorization: Bearer \$OPENAI_API_KEY' \
- -H 'Content-Type: application/json' \
- -H 'OpenAl-Beta: assistants=v1'

Response

Copy

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```
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20
 "object": "list",
 "data": [
  {
    "id": "file-dEWwUbt2UGHp3v0e0DpCzemP",
   "object": "assistant.file",
   "created_at": 1699060412,
   "assistant_id": "asst_DUGk5I7sK0FpKeijvrO30z9J"
   "id": "file-9F1ex49ipEnKzyLUNnCA0Yzx",
   "object": "assistant.file",
   "created_at": 1699060412,
   "assistant_id": "asst_DUGk5I7sK0FpKeijvrO30z9J"
 "first_id": "file-dEWwUbt2UGHp3v0e0DpCzemP",
 "last_id": "file-9F1ex49ipEnKzyLUNnCA0Yzx",
 "has_more": false
```

Threads

Beta

7

Create threads that assistants can interact with.

Related guide: Assistants

The thread object

Beta

Represents a thread that contains **messages**.

id

string

The identifier, which can be referenced in API endpoints.

obiect

string

The object type, which is always thread.

created_at

integer

The Unix timestamp (in seconds) for when the thread was created.

metadata

map

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

The thread object

Copy

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{
"id": "thread_abc123",
"object": "thread",
"created_at": 1698107661,
"metadata": {}
}
```

Create thread

Beta

POST https://api.openai.com/v1/threads

Create a thread.

Request body

messages

array Optional

A list of **messages** to start the thread with.

Show properties

metadata

map

Optional

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

Returns

A thread object.

EMPTYMESSAGES

```
Example request
curl
Copy
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curl <a href="https://api.openai.com/v1/threads">https://api.openai.com/v1/threads</a> \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -H "OpenAI-Beta: assistants=v1" \
 -d "
Response
Copy
1
2
3
4
5
6
 "id": "thread_abc123",
 "object": "thread",
 "created_at": 1699012949,
 "metadata": {}
```

Retrieve thread

Beta

GET https://api.openai.com/v1/threads/{thread_id}

Retrieves a thread.

Path parameters

thread_id

string

Required

The ID of the thread to retrieve.

Returns

The **thread** object matching the specified ID.

```
Example request
```

```
curl
Copy
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curl <a href="https://api.openai.com/v1/threads/thread-abc123">https://api.openai.com/v1/threads/thread-abc123</a> \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -H "OpenAI-Beta: assistants=v1"
Response
Copy
1
2
3
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5
6
 "id": "thread_abc123",
 "object": "thread",
 "created_at": 1699014083,
 "metadata": {}
```

Modify thread

Beta

POST https://api.openai.com/v1/threads/{thread_id}

Modifies a thread.

Path parameters

thread_id

string

Required

The ID of the thread to modify. Only the metadata can be modified.

Request body

metadata

map

Optional

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

Returns

The modified **thread** object matching the specified ID.

Example request

curl

Copy

```
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curl <a href="https://api.openai.com/v1/threads/thread-abc123">https://api.openai.com/v1/threads/thread-abc123</a> \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -H "OpenAI-Beta: assistants=v1" \
 -d '{
    "metadata": {
     "modified": "true",
     "user": "abc123"
Response
Copy
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8
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 "id": "thread_abc123",
 "object": "thread",
 "created_at": 1699014083,
 "metadata": {
  "modified": "true",
  "user": "abc123"
Delete thread
```

Beta

DELETE https://api.openai.com/v1/threads/{thread_id}
Delete a thread.

Path parameters

thread_id

string Required

The ID of the thread to delete.

Returns

Deletion status

Example request

curl Copy 1 2

3 4

curl https://api.openai.com/v1/threads/thread_abc123 \

```
-H "Content-Type: application/json" \
-H "Authorization: Bearer $OPENAL_API_KEY" \
-H "OpenAl-Beta: assistants=v1" \
-X DELETE

Response
Copy
1
2
3
4
5
{
"id": "thread_abc123",
"object": "thread.deleted",
"deleted": true
}
```

<u>Messages</u>

Beta

Create messages within threads

Related guide: **Assistants**

The message object

Beta

Represents a message within a **thread**.

id

string

The identifier, which can be referenced in API endpoints.

object

string

The object type, which is always thread.message.

created_at

integer

The Unix timestamp (in seconds) for when the message was created.

thread id

string

The **thread** ID that this message belongs to.

role

strina

The entity that produced the message. One of user or assistant.

content

array

The content of the message in array of text and/or images.

Show possible types

assistant_id

string or null

If applicable, the ID of the **assistant** that authored this message.

run id

string or null

If applicable, the ID of the **run** associated with the authoring of this message.

file ids

array

A list of <u>file</u> IDs that the assistant should use. Useful for tools like retrieval and code_interpreter that can access files. A maximum of 10 files can be attached to a message.

metadata

map

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

The message object

```
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20
 "id": "msg_dKYDWyQvtjDBi3tudL1yWKDa",
 "object": "thread.message",
 "created_at": 1698983503,
 "thread id": "thread RGUhOuO9b2nrktrmsQ2uSR6I",
 "role": "assistant",
 "content": [
   "type": "text",
   "text": {
     "value": "Hi! How can I help you today?",
     "annotations": []
   }
  }
 ],
"file_ids": [],
 "assistant_id": "asst_ToSF7Gb04YMj8AMMm50ZLLtY",
 "run_id": "run_BjylUJgDqYK9bOhy4yjAiMrn",
 "metadata": {}
```

Create message

Beta

Copy

POST https://api.openai.com/v1/threads/{thread_id}/messages Create a message.

Path parameters

thread id

string

Required

The ID of the **thread** to create a message for.

Request body

role

string

Required

The role of the entity that is creating the message. Currently only user is supported.

content

string

Required

The content of the message.

file_ids

array

Optional

Defaults to []

A list of <u>File</u> IDs that the message should use. There can be a maximum of 10 files attached to a message. Useful for tools like retrieval and code_interpreter that can access and use files.

metadata

map

Optional

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

Returns

A **message** object.

```
Example request
curl
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curl https://api.openai.com/v1/threads/thread_abc123/messages \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
 -H "OpenAl-Beta: assistants=v1" \
 -d '{
   "role": "user",
   "content": "How does AI work? Explain it in simple terms."
Response
Copy
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20
 "id": "msg_abc123",
 "object": "thread.message",
 "created_at": 1699017614,
 "thread_id": "thread_abc123",
 "role": "user",
 "content": [
 {
"type": "text",
   "text": {
```

```
"value": "How does AI work? Explain it in simple terms.",
    "annotations": []
    }
    ],
    "file_ids": [],
    "assistant_id": null,
    "run_id": null,
    "metadata": {}
}
```

Retrieve message

Beta

GET https://api.openai.com/v1/threads/{thread_id}/messages/{message_id} Retrieve a message.

Path parameters

```
thread_id
string
Required
The ID of the thread to which this message belongs.
message_id
string
Required
The ID of the message to retrieve.
```

Returns

"id": "msg_abc123",

The **message** object matching the specified ID.

```
Example request
curl
Сору
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curl https://api.openai.com/v1/threads/thread_abc123/messages/msg_abc123 \
 -H "Content-Type: application/json" \
 -H "Authorization: Bearer $OPENAI_API_KEY" \
-H "OpenAI-Beta: assistants=v1"
Response
Copy
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```

Modify message

Beta

POST https://api.openai.com/v1/threads/{thread_id}/messages/{message_id} Modifies a message.

Path parameters

thread id

string

Required

The ID of the thread to which this message belongs.

message_id

string

Required

The ID of the message to modify.

Request body

metadata

map

Optional

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

Returns

The modified message object.

Example request

```
curl
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curl <a href="https://api.openai.com/v1/threads/thread_abc123/messages/msg_abc123">https://api.openai.com/v1/threads/thread_abc123/messages/msg_abc123</a>
-H "Content-Type: application/json" \
-H "Authorization: Bearer $OPENAI_API_KEY" \
-H "OpenAI-Beta: assistants=v1" \
-d '{
    "metadata": {
```

```
"modified": "true",
     "user": "abc123"
 }
}'
Response
Copy
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22
23
 "id": "msg_abc123",
 "object": "thread.message",
 "created_at": 1699017614,
 "thread_id": "thread_abc123",
 "role": "user",
 "content": [
    "type": "text",
     "value": "How does AI work? Explain it in simple terms.",
     "annotations": []
   }
  }
 "file_ids": [],
 "assistant_id": null,
 "run_id": null,
 "metadata": {
  "modified": "true",
  "user": "abc123"
```

List messages

Beta

GET https://api.openai.com/v1/threads/{thread_id}/messages
Returns a list of messages for a given thread.

Path parameters

thread_id string Required

The ID of the **thread** the messages belong to.

Query parameters

limit

integer

Optional

Defaults to 20

A limit on the number of objects to be returned. Limit can range between 1 and 100, and the default is 20.

order

string

Optional

Defaults to desc

Sort order by the created_at timestamp of the objects. asc for ascending order and desc for descending order.

after

string

Optional

A cursor for use in pagination. after is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include after=obj_foo in order to fetch the next page of the list.

before

string

Optional

A cursor for use in pagination. before is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include before=obj_foo in order to fetch the previous page of the list.

Returns

A list of **message** objects.

Example request

curl

Сору

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curl https://api.openai.com/v1/threads/thread_abc123/messages \

- -H "Content-Type: application/json" \
- -H "Authorization: Bearer \$OPENAL APL KEY" \
- -H "OpenAl-Beta: assistants=v1"

Response

Copy

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50
 "object": "list",
 "data": [
    "id": "msg_abc123",
    "object": "thread.message",
    "created_at": 1699016383,
"thread_id": "thread_abc123",
    "role": "user",
    "content": [
       "type": "text",
       "text": {
         "value": "How does AI work? Explain it in simple terms.",
         "annotations": []
       }
     }
    "file_ids": [],
    "assistant_id": null,
    "run_id": null,
    "metadata": {}
    "id": "msg_abc456",
    "object": "thread.message",
"created_at": 1699016383,
"thread_id": "thread_abc123",
    "role": "user",
    "content": [
       "type": "text",
       "text": {
         "value": "Hello, what is AI?",
         "annotations": []
       }
     }
    ],
```

```
"file_ids": [
    "file-abc123"
],
    "assistant_id": null,
    "run_id": null,
    "metadata": {}
}
],
"first_id": "msg_abc123",
"last_id": "msg_abc456",
"has_more": false
```

The message file object

Beta

A list of files attached to a message.

id

string

The identifier, which can be referenced in API endpoints.

object

string

The object type, which is always thread.message.file.

created at

integer

The Unix timestamp (in seconds) for when the message file was created.

message_id

string

The ID of the **message** that the **File** is attached to.

```
The message file object
```

```
Copy

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{
"id": "file-BK7bzQj3FfZFXr7DbL6xJwfo",
"object": "thread.message.file",
"created_at": 1698107661,
"message_id": "message_QLoltBbqwyAJEzlTy4y9kOMM",
"file_id": "file-BK7bzQj3FfZFXr7DbL6xJwfo"
}
```

Retrieve message file

Beta

GET https://api.openai.com/v1/threads/{thread_id}/messages/{message_id}/files/{file_id} Retrieves a message file.

Path parameters

thread id

string

Required

The ID of the thread to which the message and File belong.

message_id

string

Required

The ID of the message the file belongs to.

file_id

string

Required

The ID of the file being retrieved.

Returns

The message file object.

```
Example request
curl
Copy
2
3
4
curl
https://api.openai.com/v1/threads/thread RGUhOuO9b2nrktrmsQ2uSR6I/messages/msg g3XhbGmMzsgE
Fa81gMLBDAVU/files/file-dEWwUbt2UGHp3v0e0DpCzemP \
-H 'Authorization: Bearer $OPENAL APL KEY' \
-H 'Content-Type: application/json' \
-H 'OpenAI-Beta: assistants=v1'
Response
Copy
1
2
3
4
5
6
 "id": "file-dEWwUbt2UGHp3v0e0DpCzemP",
 "object": "thread.message.file",
 "created_at": 1699061776,
 "message_id": "msg_q3XhbGmMzsqEFa81gMLBDAVU"
```

List message files

Beta

GET <a href="https://api.openai.com/v1/threads/{thread_id}/messages/{message_id}/files Returns a list of message files.

Path parameters

thread_id

string

Required

The ID of the thread that the message and files belong to.

message_id

string

Required

The ID of the message that the files belongs to.

Query parameters

limit

integer

Optional

Defaults to 20

A limit on the number of objects to be returned. Limit can range between 1 and 100, and the default is 20.

order

string

Optional

Defaults to desc

Sort order by the created_at timestamp of the objects. asc for ascending order and desc for descending order.

after

string

Optional

A cursor for use in pagination. after is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include after=obj_foo in order to fetch the next page of the list.

before

string

Optional

A cursor for use in pagination. before is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include before=obj_foo in order to fetch the previous page of the list.

Returns

A list of **message file** objects.

```
Example request
curl
Copy
2
3
4
curl
https://api.openai.com/v1/threads/thread_RGUhOuO9b2nrktrmsQ2uSR6I/messages/msg_q3XhbGmMzsqE
Fa81gMLBDAVU/files \
 -H 'Authorization: Bearer $OPENAL APL KEY' \
-H 'Content-Type: application/json' \
-H 'OpenAl-Beta: assistants=v1'
Response
Copy
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
 "object": "list",
 "data": [
 {
    "id": "file-dEWwUbt2UGHp3v0e0DpCzemP",
   "object": "thread.message.file",
   "created_at": 1699061776,
   "message_id": "msg_q3XhbGmMzsqEFa81gMLBDAVU"
  },
   "id": "file-dEWwUbt2UGHp3v0e0DpCzemP",
   "object": "thread.message.file",
   "created_at": 1699061776,
   "message_id": "msg_q3XhbGmMzsqEFa81gMLBDAVU"
```

```
],
"first_id": "file-dEWwUbt2UGHp3v0e0DpCzemP",
"last_id": "file-dEWwUbt2UGHp3v0e0DpCzemP",
"has_more": false
}
```

<u>Runs</u>

Beta

Represents an execution run on a thread.

Related guide: Assistants

The run object

Beta

Represents an execution run on a thread.

id

string

The identifier, which can be referenced in API endpoints.

object

string

The object type, which is always assistant.run.

created_at

integer

The Unix timestamp (in seconds) for when the run was created.

thread id

string

The ID of the **thread** that was executed on as a part of this run.

assistant id

string

The ID of the **assistant** used for execution of this run.

status

string

The status of the run, which can be

either queued, in_progress, requires_action, cancelling, cancelled, failed, completed, or expired.

required_action

object or null

Details on the action required to continue the run. Will be null if no action is required.

Show properties

last_error

object or null

The last error associated with this run. Will be null if there are no errors.

Show properties

expires_at

integer

The Unix timestamp (in seconds) for when the run will expire.

started at

integer or null

The Unix timestamp (in seconds) for when the run was started.

cancelled at

integer or null

The Unix timestamp (in seconds) for when the run was cancelled.

failed at

integer or null

The Unix timestamp (in seconds) for when the run failed.

completed at

integer or null

The Unix timestamp (in seconds) for when the run was completed.

model

string

The model that the **assistant** used for this run.

instructions

string

The instructions that the **assistant** used for this run.

tools

array

The list of tools that the <u>assistant</u> used for this run. Show possible types

file_ids

array

The list of File IDs the assistant used for this run.

metadata

map

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

The run object

```
Copy
2
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16
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18
19
 "id": "run_example123",
 "object": "thread.run",
 "created_at": 1698107661,
 "assistant_id": "asst_gZ1aOomboBuYWPcXJx4vAYB0",
 "thread_id": "thread_adOpf7Jbb5Abymz0QbwxAh3c",
 "status": "completed",
 "started at": 1699073476,
 "expires at": null.
 "cancelled_at": null,
 "failed_at": null,
 "completed at": 1699073498,
 "last_error": null,
 "model": "gpt-4",
 "instructions": null,
 "tools": [{"type": "retrieval"}, {"type": "code_interpreter"}],
 "file_ids": [],
 "metadata": {}
```

Create run

Beta

POST https://api.openai.com/v1/threads/{thread_id}/runs Create a run.

Path parameters

thread id

string

Required

The ID of the thread to run.

Request body

assistant_id

string

Required

The ID of the **assistant** to use to execute this run.

model

string or null

Optional

The ID of the <u>Model</u> to be used to execute this run. If a value is provided here, it will override the model associated with the assistant. If not, the model associated with the assistant will be used.

instructions

string or null

Optional

Override the default system message of the assistant. This is useful for modifying the behavior on a per-run basis.

tools

array or null

Optional

Override the tools the assistant can use for this run. This is useful for modifying the behavior on a per-run basis.

Show possible types

metadata

map

10 11 12

Optional

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

Returns

A run object.

```
Example request
curl
Copy
1
2
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6
curl https://api.openai.com/v1/threads/thread BDDwlgM4KgHibXX3mgmN3Lgs/runs \
-H 'Authorization: Bearer $OPENAI_API_KEY' \
-H 'Content-Type: application/json' \
 -H 'OpenAI-Beta: assistants=v1' \
 -d '{
  "assistant_id": "asst_nGl00s4xa9zmVY6Fvuvz9wwQ"
}'
Response
Copy
1
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6
7
8
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```

```
13
14
15
16
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18
19
20
21
22
23
24
25
26
 "id": "run_UWvV94U0FQYiT2rlbBrdEVmC",
 "object": "thread.run",
 "created_at": 1699063290,
 "assistant_id": "asst_nGl00s4xa9zmVY6Fvuvz9wwQ",
 "thread_id": "thread_BDDwlqM4KgHibXX3mqmN3Lgs",
 "status": "queued",
 "started_at": 1699063290,
 "expires_at": null,
 "cancelled_at": null,
 "failed_at": null,
 "completed_at": 1699063291,
 "last_error": null,
 "model": "gpt-4",
 "instructions": null,
 "tools": [
   "type": "code_interpreter"
  }
 "file_ids": [
  "file-9F1ex49ipEnKzyLUNnCA0Yzx",
  "file-dEWwUbt2UGHp3v0e0DpCzemP"
],
"metadata": {}
```

Retrieve run

GET https://api.openai.com/v1/threads/{thread_id}/runs/{run_id} Retrieves a run.

Path parameters

thread_id

string

Required

The ID of the **thread** that was run.

run_id

string

Required

The ID of the run to retrieve.

Returns

The **run** object matching the specified ID.

Example request

curl

Сору

1

```
3
curl https://api.openai.com/v1/threads/thread_BDDwlqM4KgHibXX3mqmN3Lgs/runs/run_
5pyUEwhaPk11vCKiDneUWXXY \
 -H 'Authorization: Bearer $OPENAI_API_KEY' \
 -H 'OpenAl-Beta: assistants=v1'
Response
Copy
1
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8
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10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
25
26
 "id": "run_5pyUEwhaPk11vCKiDneUWXXY",
 "object": "thread.run",
 "created_at": 1699075072,
 "assistant_id": "asst_nGl00s4xa9zmVY6Fvuvz9wwQ",
 "thread_id": "thread_BDDwlqM4KgHibXX3mqmN3Lgs",
 "status": "completed",
 "started_at": 1699075072,
 "expires_at": null,
 "cancelled_at": null,
 "failed_at": null,
 "completed_at": 1699075073,
 "last_error": null,
 "model": "gpt-3.5-turbo",
 "instructions": null,
 "tools": [
    "type": "code_interpreter"
  }
 "file_ids": [
  "file-9F1ex49ipEnKzyLUNnCA0Yzx",
  "file-dEWwUbt2UGHp3v0e0DpCzemP"
 ],
"metadata": {}
```

Modify run

Beta

POST https://api.openai.com/v1/threads/{thread_id}/runs/{run_id} Modifies a run.

Path parameters

```
thread_id
string
Required
The ID of the thread that was run.
run_id
string
Required
The ID of the run to modify.
```

Request body

metadata

map

16 17 18

Optional

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

Returns

The modified **run** object matching the specified ID.

```
Example request
curl
Copy
2
3
4
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6
7
8
curl https://api.openai.com/v1/threads/thread BDDwlgM4KgHibXX3mgmN3Lgs/runs/run
5pyUEwhaPk11vCKiDneUWXXY \
-H 'Authorization: Bearer $OPENAI_API_KEY' \
 -H 'Content-Type: application/json' \
 -H 'OpenAl-Beta: assistants=v1' \
 -d '{
  "metadata": {
   "user_id": "user_zmVY6FvuBDDwlqM4KgH"
Response
Copy
1
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5
6
7
8
9
10
11
12
13
14
15
```

```
20
21
22
23
24
25
26
27
28
 "id": "run_5pyUEwhaPk11vCKiDneUWXXY",
 "object": "thread.run",
 "created at": 1699075072,
 "assistant_id": "asst_nGl00s4xa9zmVY6Fvuvz9wwQ",
 "thread_id": "thread_BDDwlqM4KgHibXX3mqmN3Lgs",
 "status": "completed",
 "started_at": 1699075072,
 "expires_at": null,
 "cancelled_at": null,
 "failed at": null,
 "completed_at": 1699075073,
 "last_error": null,
 "model": "gpt-3.5-turbo",
 "instructions": null,
 "tools": [
  {
    "type": "code_interpreter"
  }
 ],
 "file_ids": [
  "file-9F1ex49ipEnKzyLUNnCA0Yzx",
  "file-dEWwUbt2UGHp3v0e0DpCzemP"
 "metadata": {
  "user_id": "user_zmVY6FvuBDDwlqM4KgH"
```

<u>List runs</u>

Beta

19

GET https://api.openai.com/v1/threads/{thread_id}/runs

Returns a list of runs belonging to a thread.

Path parameters

thread_id

string

Required

The ID of the thread the run belongs to.

Query parameters

limit

integer

Optional

Defaults to 20

A limit on the number of objects to be returned. Limit can range between 1 and 100, and the default is 20.

order

string

Optional

Defaults to desc

Sort order by the created_at timestamp of the objects. asc for ascending order and desc for descending order.

after

string

Optional

A cursor for use in pagination. after is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include after=obj_foo in order to fetch the next page of the list.

before

string

Optional

A cursor for use in pagination. before is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include before=obj_foo in order to fetch the previous page of the list.

Returns

A list of **run** objects.

Example request curl Copy curl https://api.openai.com/v1/threads/thread_BDDwlqM4KgHibXX3mqmN3Lgs/runs \ -H 'Authorization: Bearer \$OPENAI_API_KEY' \ -H 'Content-Type: application/json' \ -H 'OpenAl-Beta: assistants=v1' Response Copy

```
38
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40
41
42
43
44
45
46
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48
49
50
51
52
53
54
55
56
57
58
59
60
 "object": "list",
 "data": [
   "id": "run_5pyUEwhaPk11vCKiDneUWXXY",
   "object": "thread.run",
   "created at": 1699075072,
   "assistant_id": "asst_nGl00s4xa9zmVY6Fvuvz9wwQ",
   "thread_id": "thread_BDDwlqM4KgHibXX3mqmN3Lgs",
   "status": "completed",
   "started_at": 1699075072,
   "expires_at": null,
   "cancelled_at": null,
   "failed_at": null,
   "completed_at": 1699075073,
   "last_error": null,
   "model": "gpt-3.5-turbo",
   "instructions": null,
   "tools": [
      "type": "code_interpreter"
    }
   "file_ids": [
    "file-9F1ex49ipEnKzyLUNnCA0Yzx",
     "file-dEWwUbt2UGHp3v0e0DpCzemP"
   "metadata": {}
  },
   "id": "run_UWvV94U0FQYiT2rlbBrdEVmC",
   "object": "thread.run",
   "created_at": 1699063290,
   "assistant_id": "asst_nGl00s4xa9zmVY6Fvuvz9wwQ",
   "thread_id": "thread_BDDwlqM4KgHibXX3mqmN3Lgs",
   "status": "completed",
   "started_at": 1699063290,
   "expires_at": null,
   "cancelled_at": null,
   "failed_at": null,
   "completed_at": 1699063291,
   "last_error": null,
   "model": "gpt-3.5-turbo",
```

```
"instructions": null,
  "tools": [
      {
            "type": "code_interpreter"
      }
      ],
      "file_ids": [
            "file-9F1ex49ipEnKzyLUNnCA0Yzx",
            "file-dEWwUbt2UGHp3v0e0DpCzemP"
      ],
      "metadata": {}
    }
    ],
    "first_id": "run_5pyUEwhaPk11vCKiDneUWXXY",
    "last_id": "run_UWvV94U0FQYiT2rlbBrdEVmC",
    "has_more": false
```

Submit tool outputs to run

Beta

POST https://api.openai.com/v1/threads/{thread_id}/runs/{run_id}/submit_tool_outputs

When a run has the status: "requires_action" and required_action.type is submit_tool_outputs, this endpoint can be used to submit the outputs from the tool calls once they're all completed. All outputs must be submitted in a single request.

Path parameters

thread id

string

Required

The ID of the **thread** to which this run belongs.

run id

string

Required

The ID of the run that requires the tool output submission.

Request body

tool outputs

array

Required

A list of tools for which the outputs are being submitted.

Show properties

Returns

The modified **run** object matching the specified ID.

Example request

curl Copy

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12

curl

https://api.openai.com/v1/threads/thread_EdR8UvCDJ035LFEJZMt3AxCd/runs/run_PHLyHQYIQn4F7JrSXsIEYWwh/submit_tool_outputs\

⁻H 'Authorization: Bearer \$OPENAI_API_KEY' \

```
-H 'Content-Type: application/json' \
 -H 'OpenAl-Beta: assistants=v1' \
  "tool_outputs": [
     \label{locallid} \begin{tabular}{ll} "tool\_call\_id": "call\_MbELIQcB72cq35Yzo2MRw5qs", \\ "output": "28C" \end{tabular}
Response
Copy
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26
27
28
29
30
31
32
33
34
35
36
37
38
39
40
41
42
43
44
45
46
 "id": "run_PHLyHQYIQn4F7JrSXslEYWwh",
 "object": "thread.run",
 "created_at": 1699075592,
 "assistant_id": "asst_IgmpQTah3ZfPHCVZjTqAY8Kv",
 "thread_id": "thread_EdR8UvCDJ035LFEJZMt3AxCd",
 "status": "queued",
```

```
"started_at": 1699075592,
"expires_at": 1699076192,
"cancelled_at": null,
"failed_at": null,
"completed_at": null,
"last_error": null,
"model": "gpt-4",
"instructions": "You tell the weather.",
"tools": [
   "type": "function",
   "function": {
    "name": "get_weather",
    "description": "Determine weather in my location",
     "parameters": {
      "type": "object",
      "properties": {
       "location": {
         "type": "string",
"description": "The city and state e.g. San Francisco, CA"
      },
"unit": {
"pe"
         "type": "string",
"enum": [
          "c",
          "f"
       }
      },
      "required": [
       "location"
],
"file_ids": [],
"metadata": {}
```

Cancel a run

Beta

POST https://api.openai.com/v1/threads/{thread_id}/runs/{run_id}/cancel Cancels a run that is in_progress.

Path parameters

```
thread_id
```

string

Required

The ID of the thread to which this run belongs.

run_id

string

Required

The ID of the run to cancel.

Returns

The modified **run** object matching the specified ID.

Example request

curl

Сору

1

2

```
curl https://api.openai.com/v1/threads/thread_
1cjnJPXj8MFigTx58jU9TivC/runs/run BeRGmpGt2wb1VI22ZRniOkrR/cancel \
 -H 'Authorization: Bearer $OPENAI_API_KEY' \
 -H 'OpenAl-Beta: assistants=v1' \
 -X POST
Response
Copy
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5
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7
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10
11
12
13
14
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16
17
18
19
20
21
22
23
 "id": "run_BeRGmpGt2wb1VI22ZRniOkrR",
 "object": "thread.run",
 "created_at": 1699076126,
 "assistant_id": "asst_IgmpQTah3ZfPHCVZjTqAY8Kv",
 "thread_id": "thread_1cjnJPXj8MFiqTx58jU9TivC",
 "status": "cancelling",
 "started_at": 1699076126,
 "expires_at": 1699076726,
 "cancelled_at": null,
 "failed_at": null,
 "completed_at": null,
 "last_error": null,
 "model": "gpt-4",
 "instructions": "You summarize books.",
 "tools": [
    "type": "retrieval"
  }
 "file_ids": [],
 "metadata": {}
```

Create thread and run

Beta

POST https://api.openai.com/v1/threads/runs

Create a thread and run it in one request.

Request body

assistant_id

string

Required

The ID of the **assistant** to use to execute this run.

thread

object Optional

Show properties

model

string or null

Optional

The ID of the <u>Model</u> to be used to execute this run. If a value is provided here, it will override the model associated with the assistant. If not, the model associated with the assistant will be used.

instructions

string or null

Optional

Override the default system message of the assistant. This is useful for modifying the behavior on a per-run basis.

tools

array or null

Optional

Override the tools the assistant can use for this run. This is useful for modifying the behavior on a per-run basis.

metadata

map

5

Optional

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

Returns

A run object.

Example request

```
curl
Copy
2
3
4
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6
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10
11
curl https://api.openai.com/v1/threads/runs \
 -H 'Authorization: Bearer $OPENAI_API_KEY' \
 -H 'Content-Type: application/json' \
 -H 'OpenAI-Beta: assistants=v1' \
   "assistant_id": "asst_lgmpQTah3ZfPHCVZjTqAY8Kv",
   "thread": {
    "messages": [
      {"role": "user", "content": "Explain deep learning to a 5 year old."}
Response
Copy
2
3
4
```

```
7
8
9
10
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12
13
14
15
16
17
18
19
 "id": "run_3Qudf05GGhCleEg9ggwfJQih",
 "object": "thread.run",
 "created_at": 1699076792,
 "assistant_id": "asst_IgmpQTah3ZfPHCVZjTqAY8Kv",
 "thread_id": "thread_Ec3eKZcWI00WDZRC7FZci8hP",
 "status": "queued",
 "started_at": null,
 "expires_at": 1699077392,
 "cancelled_at": null,
 "failed_at": null,
 "completed_at": null,
 "last_error": null,
 "model": "gpt-4",
 "instructions": "You are a helpful assistant.",
 "tools": [],
 "file_ids": [],
 "metadata": {}
```

The run step object

Beta

6

Represents a step in execution of a run.

```
string
```

The identifier of the run step, which can be referenced in API endpoints.

object

string

The object type, which is always `assistant.run.step``.

created at

integer

The Unix timestamp (in seconds) for when the run step was created.

assistant_id

string

The ID of the **assistant** associated with the run step.

thread_id

string

The ID of the **thread** that was run.

run_id

string

The ID of the <u>run</u> that this run step is a part of.

type

strina

The type of run step, which can be either message creation or tool calls.

status

strina

The status of the run, which can be either in progress, cancelled, failed, completed, or expired.

step_details

object

The details of the run step.

Show possible types

last error

object or null

The last error associated with this run step. Will be null if there are no errors.

Show properties

expired_at

integer or null

The Unix timestamp (in seconds) for when the run step expired. A step is considered expired if the parent run is expired.

cancelled at

integer or null

The Unix timestamp (in seconds) for when the run step was cancelled.

failed at

integer or null

The Unix timestamp (in seconds) for when the run step failed.

completed_at

integer or null

The Unix timestamp (in seconds) for when the run step completed.

metadata

map

Set of 16 key-value pairs that can be attached to an object. This can be useful for storing additional information about the object in a structured format. Keys can be a maximum of 64 characters long and values can be a maximum of 512 characters long.

The run step object

```
Copy
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 "id": "step_QyjyrsVsysd7F4K894BZHG97",
 "object": "thread.run.step",
 "created at": 1699063291,
 "run_id": "run_UWvV94U0FQYiT2rlbBrdEVmC",
 "assistant id": "asst nGl00s4xa9zmVY6Fvuvz9wwQ",
 "thread id": "thread BDDwlgM4KgHibXX3mgmN3Lgs",
 "type": "message_creation",
 "status": "completed",
 "cancelled_at": null,
 "completed_at": 1699063291,
 "expired_at": null,
 "failed at": null,
 "last error": null,
 "step details": {
  "type": "message_creation",
  "message_creation": {
```

```
"message_id": "msg_6YmiCRmMbbE6FALYNePPHqwm"
Retrieve run step
GET https://api.openai.com/v1/threads/{thread_id}/runs/{run_id}/steps/{step_id}
Retrieves a run step.
Path parameters
thread_id
string
Required
The ID of the thread to which the run and run step belongs.
run_id
string
Required
The ID of the run to which the run step belongs.
step_id
string
Required
The ID of the run step to retrieve.
Returns
The run step object matching the specified ID.
Example request
curl
Copy
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https://api.openai.com/v1/threads/thread_BDDwlqM4KgHibXX3mqmN3Lgs/runs/run_UWvV94U0FQYiT2rlb
BrdEVmC/steps/step_QyjyrsVsysd7F4K894BZHG97 \
 -H 'Authorization: Bearer $OPENAI_API_KEY' \
-H 'Content-Type: application/json' \
-H 'OpenAI-Beta: assistants=v1'
Response
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```

```
"id": "step_QyjyrsVsysd7F4K894BZHG97",
"object": "thread.run.step",
"created_at": 1699063291,
"run_id": "run_UWvV94U0FQYiT2rlbBrdEVmC",
"assistant_id": "asst_nGl00s4xa9zmVY6Fvuvz9wwQ",
"thread_id": "thread_BDDwlqM4KgHibXX3mqmN3Lgs",
"type": "message_creation",
"status": "completed",
"cancelled_at": null,
"completed_at": 1699063291,
"expired_at": null,
"failed_at": null,
"last error": null,
"step details": {
"type": "message_creation",
"message_creation": {
  "message id": "msg 6YmiCRmMbbE6FALYNePPHqwm"
```

List run steps

Beta

GET https://api.openai.com/v1/threads/{thread_id}/runs/{run_id}/steps

Returns a list of run steps belonging to a run.

Path parameters

thread id

string

Required

The ID of the thread the run and run steps belong to.

run_id

string

Required

The ID of the run the run steps belong to.

Query parameters

limit

integer

Optional

Defaults to 20

A limit on the number of objects to be returned. Limit can range between 1 and 100, and the default is 20.

order

string

Optional

Defaults to desc

Sort order by the created_at timestamp of the objects. asc for ascending order and desc for descending order.

after

string

Optional

A cursor for use in pagination. after is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include after=obj_foo in order to fetch the next page of the list.

before

string

Optional

A cursor for use in pagination. before is an object ID that defines your place in the list. For instance, if you make a list request and receive 100 objects, ending with obj_foo, your subsequent call can include before=obj_foo in order to fetch the previous page of the list.

Returns

A list of **run step** objects.

```
Example request
curl
Copy
1
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https://api.openai.com/v1/threads/thread BDDwlgM4KgHibXX3mgmN3Lgs/runs/run UWvV94U0FQYiT2rlb
BrdEVmC/steps \
 -H 'Authorization: Bearer $OPENAI_API_KEY' \
 -H 'Content-Type: application/json' \
-H 'OpenAI-Beta: assistants=v1'
Response
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29
 "object": "list",
 "data": [
   "id": "step_QyjyrsVsysd7F4K894BZHG97",
   "object": "thread.run.step",
   "created at": 1699063291,
   "run id": "run UWvV94U0FQYiT2rlbBrdEVmC",
   "assistant_id": "asst_nGl00s4xa9zmVY6Fvuvz9wwQ",
   "thread_id": "thread_BDDwlqM4KgHibXX3mqmN3Lgs",
   "type": "message_creation",
   "status": "completed",
   "cancelled_at": null,
   "completed_at": 1699063291,
   "expired_at": null,
   "failed_at": null,
   "last_error": null,
   "step_details": {
    "type": "message_creation",
    "message_creation": {
```

```
"message_id": "msg_6YmiCRmMbbE6FALYNePPHqwm"
}
}
],
"first_id": "step_QyjyrsVsysd7F4K894BZHG97",
"last_id": "step_QyjyrsVsysd7F4K894BZHG97",
"has_more": false
```

Fine-tunes

Deprecated

Manage legacy fine-tuning jobs to tailor a model to your specific training data.

We recommend transitioning to the updating fine-tuning API

The fine-tune object

Deprecated

The FineTune object represents a legacy fine-tune job that has been created through the API.

id

string

The object identifier, which can be referenced in the API endpoints.

created at

integer

The Unix timestamp (in seconds) for when the fine-tuning job was created.

events

array

The list of events that have been observed in the lifecycle of the FineTune job.

Show properties

fine_tuned_model

string or null

The name of the fine-tuned model that is being created.

hyperparams

object

The hyperparameters used for the fine-tuning job. See the **fine-tuning guide** for more details. Show properties

model

string

The base model that is being fine-tuned.

object

string

The object type, which is always "fine-tune".

organization id

string

The organization that owns the fine-tuning job.

result files

arrav

The compiled results files for the fine-tuning job.

status

string

The current status of the fine-tuning job, which can be either created, running, succeeded, failed, or cancelled.

training_files

array

The list of files used for training.

updated at

integer

The Unix timestamp (in seconds) for when the fine-tuning job was last updated.

validation_files

array

The list of files used for validation.

```
64
65
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69
 "id": "ft-AF1WoRqd3aJAHsqc9NY7iL8F",
 "object": "fine-tune",
 "model": "curie",
 "created_at": 1614807352,
 "events": [
    "object": "fine-tune-event",
   "created at": 1614807352,
   "level": "info",
   "message": "Job enqueued. Waiting for jobs ahead to complete. Queue number: 0."
   "object": "fine-tune-event",
   "created_at": 1614807356,
   "level": "info",
   "message": "Job started."
    "object": "fine-tune-event",
   "created_at": 1614807861,
   "level": "info",
   "message": "Uploaded snapshot: curie:ft-acmeco-2021-03-03-21-44-20."
   "object": "fine-tune-event",
   "created_at": 1614807864,
   "level": "info",
   "message": "Uploaded result files: file-abc123."
   "object": "fine-tune-event",
   "created_at": 1614807864,
   "level": "info",
   "message": "Job succeeded."
 "fine_tuned_model": "curie:ft-acmeco-2021-03-03-21-44-20",
 "hyperparams": {
  "batch_size": 4,
  "learning_rate_multiplier": 0.1,
  "n_epochs": 4,
  "prompt_loss_weight": 0.1,
 "organization_id": "org-123",
 "result_files": [
   "id": "file-abc123",
   "object": "file",
   "bytes": 81509,
   "created_at": 1614807863,
   "filename": "compiled_results.csv",
    "purpose": "fine-tune-results"
 "status": "succeeded",
 "validation_files": [],
 "training_files": [
   "id": "file-abc123",
```

```
"object": "file",
"bytes": 1547276,
"created_at": 1610062281,
"filename": "my-data-train.jsonl",
"purpose": "fine-tune"
}
],
"updated_at": 1614807865,
```

Create fine-tune

Deprecated

POST https://api.openai.com/v1/fine-tunes

Creates a job that fine-tunes a specified model from a given dataset.

Response includes details of the enqueued job including job status and the name of the fine-tuned models once complete.

Learn more about fine-tuning

Request body

training_file

string

Required

The ID of an uploaded file that contains training data.

See **upload file** for how to upload a file.

Your dataset must be formatted as a JSONL file, where each training example is a JSON object with the keys "prompt" and "completion". Additionally, you must upload your file with the purpose fine-tune.

See the **fine-tuning guide** for more details.

batch_size

integer or null

Optional

Defaults to null

The batch size to use for training. The batch size is the number of training examples used to train a single forward and backward pass.

By default, the batch size will be dynamically configured to be ~0.2% of the number of examples in the training set, capped at 256 - in general, we've found that larger batch sizes tend to work better for larger datasets.

classification betas

array or null

Optional

Defaults to null

If this is provided, we calculate F-beta scores at the specified beta values. The F-beta score is a generalization of F-1 score. This is only used for binary classification.

With a beta of 1 (i.e. the F-1 score), precision and recall are given the same weight. A larger beta score puts more weight on recall and less on precision. A smaller beta score puts more weight on precision and less on recall.

classification n classes

integer or null

Optional

Defaults to null

The number of classes in a classification task.

This parameter is required for multiclass classification.

classification positive class

string or null

Optional

Defaults to null

The positive class in binary classification.

This parameter is needed to generate precision, recall, and F1 metrics when doing binary classification.

compute_classification_metrics

boolean or null

Optional

Defaults to false

If set, we calculate classification-specific metrics such as accuracy and F-1 score using the validation set at the end of every epoch. These metrics can be viewed in the **results file**.

In order to compute classification metrics, you must provide a validation_file. Additionally, you must specify classification_n_classes for multiclass classification or classification_positive_class for binary classification.

hyperparameters

object

Optional

The hyperparameters used for the fine-tuning job.

Show properties

learning_rate_multiplier

number or null

Optional

Defaults to null

The learning rate multiplier to use for training. The fine-tuning learning rate is the original learning rate used for pretraining multiplied by this value.

By default, the learning rate multiplier is the 0.05, 0.1, or 0.2 depending on final batch_size (larger learning rates tend to perform better with larger batch sizes). We recommend experimenting with values in the range 0.02 to 0.2 to see what produces the best results.

model

string

Optional

Defaults to curie

The name of the base model to fine-tune. You can select one of "ada", "babbage", "curie", "davinci", or a fine-tuned model created after 2022-04-21 and before 2023-08-22. To learn more about these models, see the **Models** documentation.

prompt_loss_weight

number or null

Optional

Defaults to 0.01

The weight to use for loss on the prompt tokens. This controls how much the model tries to learn to generate the prompt (as compared to the completion which always has a weight of 1.0), and can add a stabilizing effect to training when completions are short.

If prompts are extremely long (relative to completions), it may make sense to reduce this weight so as to avoid over-prioritizing learning the prompt.

suffix

string or null

Optional

Defaults to null

A string of up to 40 characters that will be added to your fine-tuned model name.

For example, a suffix of "custom-model-name" would produce a model name like ada:ft-your-org:custom-model-name-2022-02-15-04-21-04.

validation file

string or null

Optional

The ID of an uploaded file that contains validation data.

If you provide this file, the data is used to generate validation metrics periodically during fine-tuning. These metrics can be viewed in the <u>fine-tuning results file</u>. Your train and validation data should be mutually exclusive.

Your dataset must be formatted as a JSONL file, where each validation example is a JSON object with the keys "prompt" and "completion". Additionally, you must upload your file with the purpose fine-tune.

See the **fine-tuning guide** for more details.

Returns

A **fine-tune** object.

```
Example request
curl
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curl <a href="https://api.openai.com/v1/fine-tunes">https://api.openai.com/v1/fine-tunes</a> \
 -H "Content-Type: application/json" \
-H "Authorization: Bearer $OPENAI_API_KEY" \
   "training_file": "file-abc123"
 }'
Response
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18
19
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22
23
24
25
26
27
28
29
30
31
32
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36
 "id": "ft-AF1WoRqd3aJAHsqc9NY7iL8F",
"object": "fine-tune",
"model": "curie",
"created_at": 1614807352,
  "events": [
     "object": "fine-tune-event",
     "created_at": 1614807352,
```

```
"level": "info",
    "message": "Job enqueued. Waiting for jobs ahead to complete. Queue number: 0."
 ],
"fine_tuned_model": null,
"crams": {
   "batch_size": 4,
   "learning_rate_multiplier": 0.1,
   "n_epochs": 4,
   "prompt_loss_weight": 0.1,
 },
"organization_id": "org-123",
 "result_files": [],
 "status": "pending",
 "validation_files": [],
 "training_files": [
  {
    "id": "file-abc123",
    "object": "file",
    "bytes": 1547276,
    "created_at": 1610062281,
    "filename": "my-data-train.jsonl", "purpose": "fine-tune-results"
 "updated_at": 1614807352,
List fine-tunes
Deprecated
GET <a href="https://api.openai.com/v1/fine-tunes">https://api.openai.com/v1/fine-tunes</a>
List your organization's fine-tuning jobs
Returns
A list of fine-tune objects.
Example request
curl
Copy
1
curl <a href="https://api.openai.com/v1/fine-tunes">https://api.openai.com/v1/fine-tunes</a> \
 -H "Authorization: Bearer $OPENAI_API_KEY"
Response
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 "object": "list",
 "data": [
    "id": "ft-AF1WoRqd3aJAHsqc9NY7iL8F",
    "object": "fine-tune",
"model": "curie",
    "created_at": 1614807352,
    "fine_tuned_model": null,
    "hyperparams": { ... },
    "organization_id": "org-123",
    "result_files": [],
    "status": "pending",
    "validation_files": [],
    "training_files": [ { ... } ],
    "updated_at": 1614807352,
  { ... },
  { ... }
```

Retrieve fine-tune

Deprecated

GET https://api.openai.com/v1/fine-tunes/{fine_tune_id}

Gets info about the fine-tune job.

Learn more about fine-tuning

Path parameters

fine_tune_id string Required The ID of the fine-tune job

Returns

The **fine-tune** object with the given ID.

```
Example request
curl
```

Copy 1

curl https://api.openai.com/v1/fine-tunes/ft-AF1WoRqd3aJAHsqc9NY7iL8F\

-H "Authorization: Bearer \$OPENAI_API_KEY"

Response

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69
 "id": "ft-AF1WoRqd3aJAHsqc9NY7iL8F",
 "object": "fine-tune",
"model": "curie",
"created_at": 1614807352,
 "events": [
  "created_at": 1614807352,
   "level": "info",
   "message": "Job enqueued. Waiting for jobs ahead to complete. Queue number: 0."
  },
```

```
"object": "fine-tune-event",
  "created_at": 1614807356,
  "level": "info",
  "message": "Job started."
  "object": "fine-tune-event",
  "created_at": 1614807861,
  "level": "info",
  "message": "Uploaded snapshot: curie:ft-acmeco-2021-03-03-21-44-20."
  "object": "fine-tune-event",
  "created at": 1614807864,
  "level": "info",
  "message": "Uploaded result files: file-abc123."
  "object": "fine-tune-event",
  "created_at": 1614807864,
  "level": "info",
  "message": "Job succeeded."
"fine_tuned_model": "curie:ft-acmeco-2021-03-03-21-44-20",
"hyperparams": {
 "batch_size": 4,
 "learning_rate_multiplier": 0.1,
 "n_epochs": 4,
 "prompt_loss_weight": 0.1,
"organization_id": "org-123",
"result_files": [
  "id": "file-abc123",
  "object": "file",
  "bytes": 81509,
  "created_at": 1614807863,
  "filename": "compiled_results.csv",
"purpose": "fine-tune-results"
"status": "succeeded",
"validation_files": [],
"training_files": [
  "id": "file-abc123",
  "object": "file",
  "bytes": 1547276,
  "created_at": 1610062281,
  "filename": "my-data-train.jsonl",
  "purpose": "fine-tune"
"updated_at": 1614807865,
```

Cancel fine-tune

Deprecated

POST https://api.openai.com/v1/fine-tunes/{fine_tune_id}/cancel Immediately cancel a fine-tune job.

Path parameters

fine_tune_id string The ID of the fine-tune job to cancel

Returns

The cancelled **fine-tune** object.

```
Example request
curl
Copy
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curl https://api.openai.com/v1/fine-tunes/ft-AF1WoRqd3aJAHsqc9NY7iL8F/cancel \
 -H "Authorization: Bearer $OPENAI_API_KEY"
Response
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 "id": "ft-xhrpBbvVUzYGo8oUO1FY4nI7",
 "object": "fine-tune",
"model": "curie",
 "created_at": 1614807770,
 "events": [ { ... } ],
 "fine_tuned_model": null,
 "hyperparams": { ... },
"organization_id": "org-123",
 "result_files": [],
 "status": "cancelled",
 "validation_files": [],
 "training_files": [
    "id": "file-abc123",
    "object": "file",
"bytes": 1547276,
    "created_at": 1610062281,
"filename": "my-data-train.jsonl",
"purpose": "fine-tune"
 "updated_at": 1614807789,
```

The fine-tune event object

Deprecated

Fine-tune event object

```
created at
integer
level
string
message
string
object
string
The fine-tune event object
Copy
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 "object": "fine-tune-event",
 "created at": 1677610602,
 "level": "info".
 "message": "Created fine-tune job"
```

List fine-tune events

Deprecated

GET https://api.openai.com/v1/fine-tunes/{fine_tune_id}/events Get fine-grained status updates for a fine-tune job.

Path parameters

fine_tune_id

string

Required

The ID of the fine-tune job to get events for.

Query parameters

stream

boolean

Optional

Defaults to false

Whether to stream events for the fine-tune job. If set to true, events will be sent as data-only **server-sent events** as they become available. The stream will terminate with a data: [DONE] message when the job is finished (succeeded, cancelled, or failed).

If set to false, only events generated so far will be returned.

Returns

A list of fine-tune event objects.

```
Example request

curl

Copy

1

2

curl <a href="https://api.openai.com/v1/fine-tunes/ft-AF1WoRqd3aJAHsqc9NY7iL8F/events">https://api.openai.com/v1/fine-tunes/ft-AF1WoRqd3aJAHsqc9NY7iL8F/events</a> \
-H "Authorization: Bearer $OPENAI_API_KEY"

Response

Copy

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24
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35
 "object": "list",
 "data": [
    "object": "fine-tune-event",
   "created_at": 1614807352, "level": "info",
   "message": "Job enqueued. Waiting for jobs ahead to complete. Queue number: 0."
    "object": "fine-tune-event",
    "created_at": 1614807356,
    "level": "info",
    "message": "Job started."
    "object": "fine-tune-event",
    "created_at": 1614807861,
    "level": "info",
    "message": "Uploaded snapshot: curie:ft-acmeco-2021-03-03-21-44-20."
    "object": "fine-tune-event",
    "created_at": 1614807864,
    "level": "info",
    "message": "Uploaded result files: file-abc123"
    "object": "fine-tune-event",
    "created_at": 1614807864,
    "level": "info",
    "message": "Job succeeded."
```

```
Edits
Deprecated
Given a prompt and an instruction, the model will return an edited version of the prompt.
The edit object
Deprecated
choices
A list of edit choices. Can be more than one if n is greater than 1.
Show properties
object
The object type, which is always edit.
created
integer
The Unix timestamp (in seconds) of when the edit was created.
usage
object
Usage statistics for the completion request.
Show properties
The edit object
Copy
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11
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13
14
15
 "object": "edit",
 "created": 1589478378,
```

Create edit

"prompt_tokens": 25,
"completion_tokens": 32,
"total_tokens": 57

Deprecated

"choices": [

], "usage": {

}

"index": 0,

POST https://api.openai.com/v1/edits

"text": "What day of the week is it?",

Creates a new edit for the provided input, instruction, and parameters.

Request body

instruction

string

Required

The instruction that tells the model how to edit the prompt.

string

Required

ID of the model to use. You can use the text-dayinci-edit-001 or code-dayinci-edit-001 model with this endpoint.

input

string or null

Optional

Defaults to "

The input text to use as a starting point for the edit.

integer or null

Optional

Defaults to 1

How many edits to generate for the input and instruction.

temperature

number or null

Optional

Defaults to 1

What sampling temperature to use, between 0 and 2. Higher values like 0.8 will make the output more random, while lower values like 0.2 will make it more focused and deterministic.

We generally recommend altering this or top_p but not both.

top_p

number or null

Optional

Defaults to 1

An alternative to sampling with temperature, called nucleus sampling, where the model considers the results of the tokens with top_p probability mass. So 0.1 means only the tokens comprising the top 10% probability mass are considered.

We generally recommend altering this or temperature but not both.

Returns

Returns an edit object.

Example request

text-davinci-edit-001

curl

Copy

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curl https://api.openai.com/v1/edits \ -H "Content-Type: application/json" \

-H "Authorization: Bearer \$OPENAI_API_KEY" \

"model": "text-davinci-edit-001", "input": "What day of the wek is it?", "instruction": "Fix the spelling mistakes"

Response

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{
    "object": "edit",
    "created": 1589478378,
    "created": [
    {
"text": "What day of the week is it?",
       "index": 0,
"usage": {
  "prompt_tokens": 25,
  "completion_tokens": 32,
  "total_tokens": 57
 }
```

From < https://platform.openai.com/docs/api-reference/edits/create>