Applied Rest APIs II



Agenda - Schedule

- 1. REST APIs Review
- 1. OpenAl API Intro
- 2. Break
- 3. OpenAl API Lab

```
{
    "Name": "Alex",
    "Age": 37,
    "Admin": true,
    "Contact": {
        "Site": "alexwebdevelop.com",
        "Phone": 123456789,
        "Address": null
    },
    "Tags": [
        "php",
        "web",
        "dev"
    ]
}
```

JavaScript Object Notation (JSON) is an open-standard data format or interchange for semi-structured data. It is text-based and readable by humans and machines. https://www.snowflake.com/guides/what-is-json

Agenda - Announcements

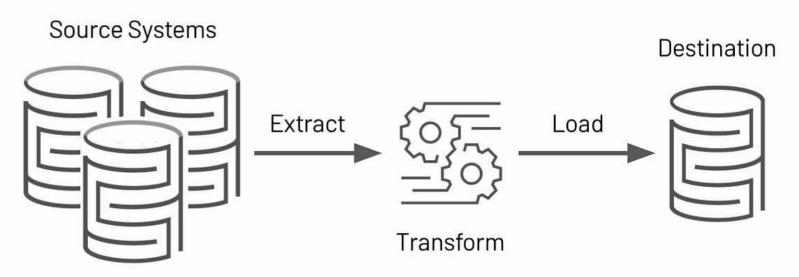
- Week 4 Pre-Class Quiz due 4/1
- TLAB #2 Due 4/21
- Office hours posted

Agenda - Goals

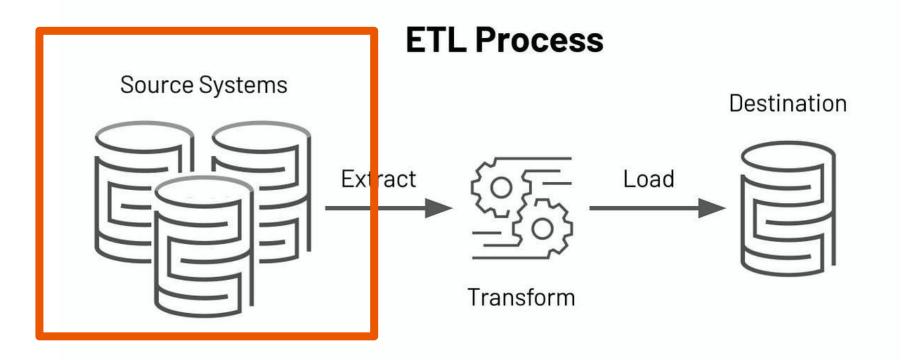
- Review Rest API usage
- Get introduced to LLM technology
- Learn how to use the OpenAl API
- Discover the principles of effective requests

Web APIs Review

ETL Process



As we discussed, one typical pipeline that in the data engineering process is the **extract-transform-load (ETL)** pipeline.



This week, we will go over various ways to extract data via **Web APIs** and **Web Scraping.** Afterwards, we will go over techniques we can use to transform data for easy analysis and prediction.

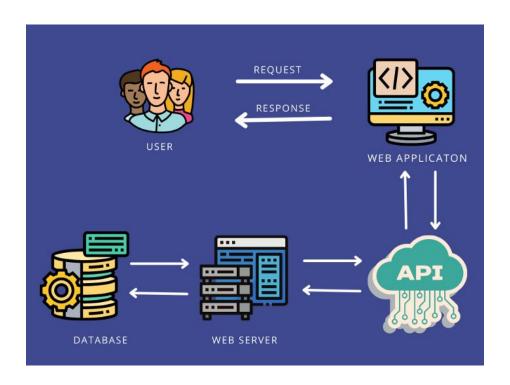
The term "API" can be used in multiple contexts:

pandas API:

An interface to use when coding in pandas

Web API:

An interface when requesting data over the web



Today, we will review the **RESTful Web API**. This describes a programmatic way to "ask respectfully" for data.

Let's review our REST API terminology. Can anyone identify the components of this URL?

https://pokeapi.co/api/v2/pokemon/pikachu

The base URL (aka the Request URL) is the URL that points to your API. This serves as your initial entry-point.

Access the pokemon Web API...

https://pokeapi.co/api/v2/pokemon/pikachu

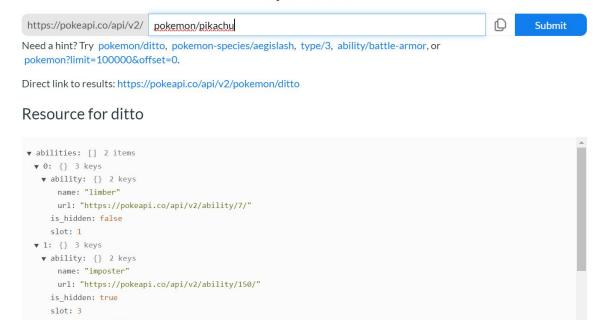
Next, we have the **endpoint** which describes where the data that you want to interact with is located.

Notice the similarities between the endpoint and file-paths... (e.g. TLAB2/data/averages.png)

...and give me information on the pokemon named pikachu

https://pokeapi.co/api/v2/pokemon/pikachu

Try it now!



But keep in mind, you will always use the API documentation to structure your URL: https://pokeapi.co/.

Which points back to the importance of good documentation: https://www.ncei.noaa.gov/support/access-data-service-api-user-documentation

The previous example describes a simple public API which entails a minimal URL.

This is rarely the type of URL you will be accessing for your projects, instead you will be accessing URLs that have **complex queries**, such as the following API:

We still have our base URL...

Access the polygon Web API...

As well as the **endpoint...**

...and give me aggregates on the AAPL ticker from Jan 1 2023 to Feb 10 2023

But this additional section describes the **query**, which provides additional parameters that allows you to **retrieve** a modified version of the resource that you are requesting, as well as **provide** required access information (for example **API keys**)

We combine multiple queries using the &:

The following queries provide the following parameters:

- adjusted=true: Return the adjusted price
- sort=asc: Sort these dates in descending order
- apiKey=123456: Here's my secret key

Aggregates (Bars)

| /v2/aggs/ticker/{stocksTicker}/range/{multiplier}/{timespan}/{from}/{to} | |
|--|-----------------------|
| Get aggregate bars for a stock over a given date range in custom time | window sizes. |
| For example, if timespan = 'minute' and multiplier = '5' then 5-minute b | oars will be returned |
| Parameters | |
| stocksTicker* AAPL | |
| Specify a case-sensitive ticker symbol. For example, AAPL represents Apple Inc. | |
| multiplier* 1 | |
| The size of the timespan multiplier. | |
| timespan* day • | |
| The size of the time window. | |
| from* 2023-01-09 | |
| The start of the aggregate time window. Either a date with the format YYYY-MM-DD or a milli- | second timestamp. |
| to* 2023-02-10 | |

The end of the aggregate time window. Either a date with the format YYYY-MM-DD or a millisecond timestamp

Read documentation for profit and success.

Once again you are not expected to construct these queries yourself. Instead, you will always use the API documentation: https://polygon.io/docs/stocks/get-v2 aggs ticker stocksticker range multiplier timespan from to

r = requests.get("https://pokeapi.co/api/v2/pokemon/pikachu")

data = r.json()

We can then use these API urls with our requests module to automatically extract information in our Python program.

```
for ab in data["abilities"]: print(ab)
```

OpenAl API

OpenAl API

For this next lab, we will have you to interact with the **OpenAl** API to send human-reviews for machine **sentiment labeling**.

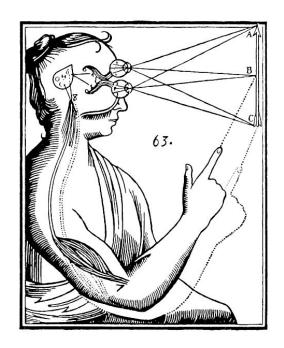
However, before we dive into the syntax, let's go over a brief overview of the context and expected usage of large language model technology.

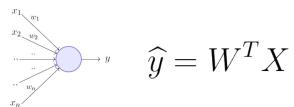
There's a lot of mixed messaging regarding LLMs, so we would like to emphasize a restrained look on how you should use LLMs in your workplace.

Neural Networks

In Phase 2 of this fellowship, we will learn about a type of machine learning algorithm called a **perceptron**.

This takes inspiration from the neurons in our brain, but purely exist as a mathematical function.



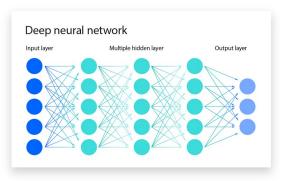


Neural Networks

We can organize these perceptrons in specific architecture to have them perform some impressive tasks such as:

- Computer vision
- Numeric prediction
- Classification
- Natural language processing



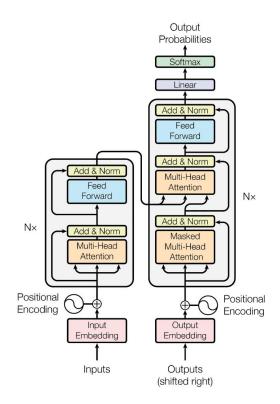


Neural Networks - Transformers

Back in 2017, Google researchers published a paper titled Attention is All you Need to introduce a neural network architecture to translate text between different languages.

In summary, this proposes a model for probabilistic **text prediction.** That is, the model is selecting the next word (*token*) based on **probabilities.**

This eventually became ChatGPT.



```
"Can I buy robux."

"Carrots are gross."

"I don't want to sleep."

"I hate carrots."

"Can I have my lpad."
```

For example, let's say you are babysitting your niece/nephew/cousin etc. You've spent a few days with them and there are certain phrases you hear them say over and over again.



You hear them begin a sentence again with "I hate..." Based on your historical data, what do you predict is going to be the next word of this sentence.



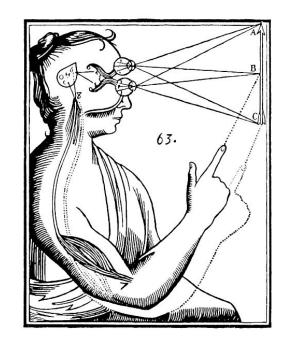
This is a massive over-simplification of ChatGPT, but this is a useful mental model to use going forward when we talk about large language models.

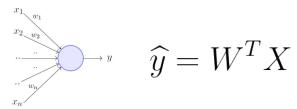
ChatGPT

This is why we emphasize that you **do not directly** use answers from ChatGPT in your TLABs/work.

By replacing your answers with LLM output, you open yourself up to:

- Hallucinations
- Not citing work
- Lack of authority in your answers





ChatGPT - Healthy Scepticism

We encourage you to have healthy **scepticism** of any new tool you come across.

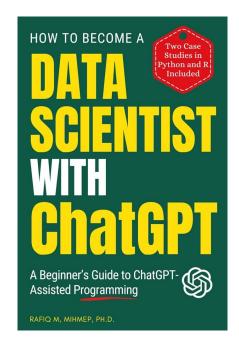
If a person/company/ceo/venture capitalist tells you that

- you need a product
- you will get left behind without it
- their product is **eternal**
- include an appeal to higher ideas (God, gov't, history)

They are either:

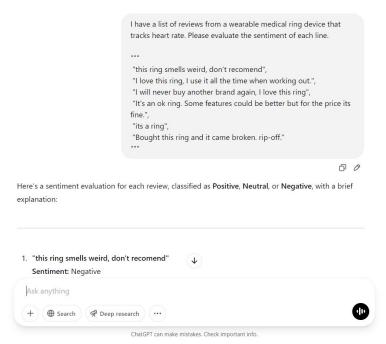
- Clueless to what they're talking about
- Trying to sell you something

Most of the time it's the **latter**.



Absolute nonsense, but good cash-grab.

OpenAl API Syntax



Hopefully everyone is familiar with ChatGPT already. While we can open a chat-window and send requests to this LLM, how can we actually send **requests** over the internet programmatically?

If you haven't interacted with ChatGPT yet, we recommend you open the following URL and practice sending it queries: https://chatgpt.com/

OpenAl API



To do this, we need to utilize something *similar* to using the **requests** module.

Sometimes, specific tools have prebuilt Python packages which we can install and utilize in our projects.

In ChatGPT's case we can use the **openai** package.

Todays lab will entail the setup of this package in your **ds** environment. We recommend that you simply watch us interact with this package as opposed to following along.

```
from openai import OpenAl
client = OpenAI()
completion = client.chat.completions.create(
 model="gpt-4o-mini",
 messages=[
           "role": "user".
           "content": "Write a one-sentence bedtime story about a unicorn."
print(completion.choices[0].message.content)
```

This block of code represents a simple "chat completion" request to OpenAI. Note that you will not be able to run this code until **you've set up your API key**.

```
from openai import OpenAl client = OpenAl()
```

print(completion.choices[0].message.content)

To begin sending requests to ChatGPT, we must first import the OpenAI class from the openai package and create the **OpenAI** object. Here we name our object "client."

```
from openai import OpenAl
client = OpenAI()
completion = client.chat.completions.create()
 model- gpt-40-mm,
 messages=[
           "role": "user".
           "content": "Write a one-sentence bedtime story about a unicorn."
print(completion.choices[0].message.content)
```

Next we can begin our request that ChatGPT **generates a response** to a list of messages using the "**completions**" API endpoint. Here we simply need to define a series of objects **client.chat.completions.create**

```
from openai import OpenAl client = OpenAl()
```

print(completion.choices[0].message.content)

Inside of the **create** API call, we specify a series of **parameters**. Remember parameters are simply variables that only exist within a specific class or function. For the full list of parameters, we can check the docs: https://platform.openai.com/docs/api-reference/chat/create

```
from openai import OpenAl
client = OpenAI()
completion = client chat completions.create(
 model="gpt-4o-mini",
 messages=
           "role": "user".
           "content": "Write a one-sentence bedtime story about a unicorn."
print(completion.choices[0].message.content)
```

The first parameter we define is **model**. This is the **large language model** that we will be interfacing with. For our purposes, this will always be **gpt-4o-mini** due to API limitations.

Text tokens

| kens · Batch API price | | tokens | per 1M | Price |
|------------------------|--|--------|--------|-------|
|------------------------|--|--------|--------|-------|

| Model | Input | Cached input | Output |
|---|----------|--------------|----------|
| gpt-4.5-preview → gpt-4.5-preview-2025-02-27 | \$75.00 | \$37.50 | \$150.00 |
| gpt-4o → gpt-4o-2024-08-06 | \$2.50 | \$1.25 | \$10.00 |
| gpt-4o-audio-preview → gpt-4o-audio-preview-2024-12-17 | \$2.50 | - | \$10.00 |
| gpt-4o-realtime-preview → gpt-4o-realtime-preview-2024-12-17 | \$5.00 | \$2.50 | \$20.00 |
| gpt-4o-mini → gpt-4o-mini-2024-07-18 | \$0.15 | \$0.075 | \$0.60 |
| gpt-4o-mini-audio-preview → gpt-4o-mini-audio-preview-2024-12-17 | \$0.15 | - | \$0.60 |
| gpt-4o-mini-realtime-preview → gpt-4o-mini-realtime-preview-2024-12-17 | \$0.60 | \$0.30 | \$2.40 |
| o1 → o1-2024-12-17 | \$15.00 | \$7.50 | \$60.00 |
| o1-pro → o1-pro-2025-03-19 | \$150.00 | ÷ | \$600.00 |
| o3-mini → o3-mini-2025-01-31 | \$1.10 | \$0.55 | \$4.40 |

However there are other models that you could possibly input. These are restricted for our API, so we will not be able to use them.

The next parameter is **messages**. This specifies the messages that we are sending to the model. Take a look at the outermost brackets of the argument of this parameter. What kind of data-structure do we pass?

```
from openai import OpenAl
client = OpenAI()
completion = client.chat.completions.create(
 model="gpt-4o-mini",
 messages=[
            "role": "user".
            "content": "Write a one-sentence bedtime story about a unicorn."
                                                             Later on, we will also explore the
                                                             "developer" message to give the system
```

Notice that we are passing in a list! This will be a list composed of dictionaries. Note that so far we only specify the user message. This the prompt that ChatGPT receives in order to generate its output.

specific instructions.

```
from openai import OpenAl
client = OpenAI()
completion = client.chat.completions.create(
 model="gpt-4o-mini",
 messages=[
           "role": "user".
           "content": "Write a one-sentence bedtime story about a unicorn."
```

Finally we access the response generated from this prompt using our **completion** variable. Note that we access its choices using index notation, what does this tell you about the output ChatGPT generates?

```
from openai import OpenAl
                                                                                             "id": "chatcmpl-B9MBs8CjcvOU2jLn4n570S5qMJKcT",
client = OpenAI()
                                                                                             "object": "chat.completion",
                                                                                             "created": 1741569952,
                                                                                             "model": "gpt-4o-2024-08-06",
                                                                                             "choices": [
completion = client.chat.completions.create(
                                                                                                "index": 0,
  model="gpt-4o-mini",
                                                                                                "message": {
                                                                                                 "role": "assistant",
  messages=[
                                                                                                 "content": "Hello! How can I assist you today?",
                                                                                                 "refusal": null,
                                                                                                 "annotations": []
               "role": "user".
               "content": "Write a one-sentence bedtime story about a unicorn."
print(completion.choices[0].message.content)
```

Just like we've encountered before, our API request results in a JSON object which we can access using Python syntax. Since "choices" is an array of different choices, we use our index notation to get the first object. We then subsequently get the message, and then content of the response.

```
from openai import OpenAl
client = OpenAI()
completion = client.chat.completions.create(
 model="gpt-4o-mini",
 messages=[
           "role": "user".
           "content": "Write a one-sentence bedtime story about a unicorn."
```

Note that often we will have to transform our string **content** into another data-structure which we can later use in other parts of our program. This is especially true for our second tlab.

We will use this for TLAB 2 in order to tell our model what kind of responses it should craft.

OpenAl API



So far we've gone over simple requests to ChatGPT.

However, often times we also want to give directions that will be shown to the model, but not to the end-user themselves.

This is especially important if you would like to **generate some sort of Chat-bot** which will have unmoderated interaction with a person. For example, maybe you want to tell your LLM that it should always have a **cheery disposition.**

To provide these types of instructions to the model, you will have to include another message using the **developer** tag. Notice that this follows the same format as the original message, except this time we are setting up the context for the models response.

A question that you might have is, "What's the point of this type of prompt?" Well to answer that, we must delve into something called **prompt engineering.**

"Prompt Engineering"

Prompt Engineering

One of the jobs that were predicted to come out of LLMs was **prompt engineering**, which is the supposed job of solely interacting with an LLM model to get some specific output.

This is most likely not to going to be a job anytime soon, but it does emphasize a certain kind of skill in LLM interactions.

It's important to note, we often **overestimate** the **effect of short-term** changes due to technological advances, but **underestimate** the long-term changes (Amara's Law).

Prompt Engineering

With this disclaimer let's go over what "prompt engineering" entails.

When you are interfacing with ChatGPT, you send text queries/requests which it then "responds to." In reality it generates a probabilistic series of words that it predicts is the most likely answer you would like to see, based on historical data (the internet).

The level of quality to these responses depends on the quality of your input.

Prompt Engineering

We will attempt to minimize the anthropomorphization of this tool, but generally speaking, prompt engineering entails some tips which involve being an **effective communicator as a manager**. This includes:

- Providing helpful examples
- Be **specific** in your request
- Do plenty of context setting, provide frameworks.

However there are other tips which only apply to LLMs. These include:

- "Gas up" your model: "You are an expert in XYZ. You can do ABC."
- Emphasize that you **need something ASAP**. Quality of responses tend to be better.

I have a list of reviews from a wearable medical ring device that tracks heart rate. Please evaluate the sentiment of each line. "this ring smells weird, don't recomend", "I love this ring, I use it all the time when working out.", "I will never buy another brand again, I love this ring", "It's an ok ring. Some features could be better but for the price its fine.", "its a ring", "Bought this ring and it came broken. rip-off." 1. "this ring smells weird, don't recomend" Sentiment: Negative Reason: Negative smell description and explicitly says "don't recommend." 2. "I love this ring, I use it all the time when working out." Sentiment: Positive Reason: Strong positive emotion ("love"), and describes frequent usage. 3. "I will never buy another brand again, I love this ring"

Reason: Expresses loyalty and strong satisfaction.

4. "It's an ok ring. Some features could be better but for the price its fine."

Reason: Balanced statement—acknowledges flaws but finds the value acceptable.

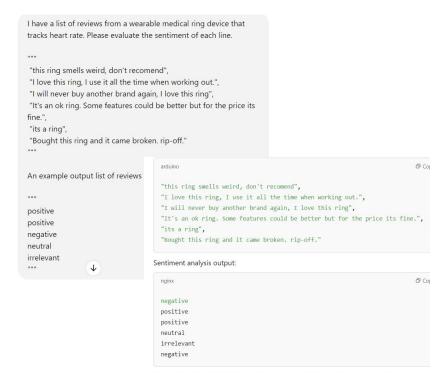
Sentiment: Positive

Sentiment: Neutral

I have a list of reviews from a wearable medical ring device that tracks heart rate. Please evaluate the sentiment of each line. "this ring smells weird, don't recomend", "I love this ring, I use it all the time when working out.", "I will never buy another brand again, I love this ring", "It's an ok ring. Some features could be better but for the price its fine.", "its a ring", "Bought this ring and it came broken. rip-off." A COL "this ring smells weird, don't recomend", An example output list "I love this ring, I use it all the time when working out.", "I will never buy another brand again, I love this ring", "It's an ok ring. Some features could be better but for the price its fine.", positive positive "Bought this ring and it came broken. rip-off." negative neutral Sentiment analysis output: irrelevant O Cor negative positive positive neutral irrelevant negative

Let me know if you'd like these categorized in a CSV or used in a visualization!

Let's say I'm asking ChatGPT to transform a list of text according to some changes. By providing **helpful examples**, I can get output closer to what I expect.



Let me know if you'd like these categorized in a CSV or used in a visualization!

This will be used in my Python project, so I need this output to be inside of a list which I will eventually plot via a matplotlib visualization.

Here's the sentiment evaluation for each review line, formatted as a Python list for project:

```
python

sentiments = [
    "negative",
    "positive",
    "neutral",
    "irrelevant",
    "negative"
]
```

Let me know if you want help generating a bar chart with matplotlib using this!

By **context setting** and specifying which language I'm working on/what my project should do, I can further refine my outputs.

This will be used in my Python project, so I need this output to be inside of a list which I will eventually plot via a matplotlib visualization.

This will be used in my Python project, so I need this output to be inside of a list which I will eventually plot via a matplotlib visualization. Please only output the list with no additional text.

Here's the sentiment evaluation for each review line, formatted as a Python list for project:

```
python

sentiments = [
    "negative",
    "positive",
    "neutral",
    "irrelevant",
    "negative"
]
```

Let me know if you want help generating a bar chart with matplotlib using this!

Finally, by being **specific in my request**, I can get my final answer **which only contains a Python list**. This will be useful for when I am interacting with the **gpt-4o-mini** API programmatically.

Prompt Engineering - Odd Tricks

As we've explained, you can also include additional context to further refine your answers.

"Gas up" your model:

- Tell the model how good it is at some specific task.
- Ex: "You are an expert software engineer with 20 years of experience in the Python language. You are aware of test-driven development..."

You need something ASAP.

• I cannot find the literature supporting this yet, but there's an odd tip that if you tell ChatGPT that you need something immediately, it performs "better"

Prompt Engineering - Odd Tricks

There are a few other odd tricks with ChatGPT that seem like they might improve performance:

- https://news.ycombinator.com/item?id=39495476
 - Pretending to tip ChatGPT might cause it to write longer text
- https://www.reddit.com/r/ChatGPT/comments/18gg6sr/tell gpt its may and i tll perform better/
 - If you tell ChatGPT it's May, it might perform better.

Take these with a grain of salt, none of this is established research.

OpenAl API Exercise

OpenAl API Exercise

For this exercise you will take set up your API token and begin practicing interacting with the OpenAI API.

However before you get started there are a couple of things to note:

- NEVER write your API key in publicly accessible code
- Be responsible with API usage

Keeping APIs Safe

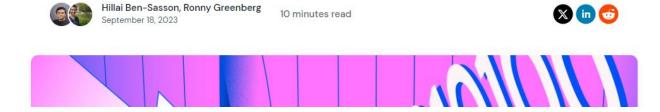
For this project, we will send you all API keys. By tonight, we will delete these keys from the Slack channel. Please be present.

You will set up your API keys in your conda environment (ds). Follow the listed directions carefully.

If you complete tonights lab, you will get ahead on TLAB #2.

38TB of data accidentally exposed by Microsoft Al researchers

Wiz Research found a data exposure incident on Microsoft's Al GitHub repository, including over 30,000 internal Microsoft Teams messages – all caused by one misconfigured SAS token



NOTE: You should always keep your key a secret. Never push your API key to GitHub. Always use the environment variable instead.

Therefore, you should make API requests responsibly:

- A few test runs are ok
- 100 test runs are not

Responsible use of APIs

Because OpenAl API is a **paid** service, there incurs a charge every time you use the API.

To prevent costs from exploding we've set up our project to limit the number of requests that can occur per minute (100) and per day (1500).

If you encounter a 429 error while running your project you must either:

- Wait 1 minute for the requests per minute (RPM) to reset
- Wait till the next day for the requests per day (RPD) to reset (this will almost never happen unless you collude to run as many requests as possible)



Using this idea, complete the following API exercise as a group. Join a group to receive an API key.

Wrap-Up

Lab (Due 04/21)



Vancouver, Canada

You are a growth analyst at a Vancouver-based consulting firm called Monica Group. Your manager is spearheading the completion of a a new analytical tool which will automatically label if a review is positive, neutral, negative, or irrelevant.

You will be kicking off completion of this milestone by independently implementing a minimal-viable-product. This will be a Python pipeline that ingests a text-file of review data and interfaces with the Open AI API in order to automatically label each review.

We will release API keys on 4/1

Wednesday

Wednesday will entail:

- A review of probability
- Applied probability
- Work on the OpenAl Lab



Jupyter: scratchpad of the data scientist

If you understand what you're doing, you're not learning anything. - Anonymous