```
The Chat Format
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One of the exciting things about a large language model is you could use it to build a custom chatbot with only a modest amount of effort. ChatGPT, the web interface, is a way for you to have a conversational interface, a conversation via a large language model. But one of the cool things is you can also use a large language model to build your custom chatbot to maybe play the role of an AI customer service agent or an AI order taker for a restaurant. And in this lesson, you'll learn how to do that by yourself. I'm going to describe the components of the OpenAI chat completions format in more detail and then you're going to build a chatbot yourself. So let's get into it.

In this notebook, you will explore how you can utilize the chat format to have extended conversations with chatbots personalized or specialized for specific tasks or behaviors.

```
Setup
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```
So first we'll set up the OpenAI Python package as usual.
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```
import os
import openai
from dotenv import load_dotenv, find_dotenv
_ = load_dotenv(find_dotenv()) # read local .env file

openai.api kev = os.getenv('OPENAI API KEY')
```

openai.api_key = os.getenv('OPENAI_API_KEY')

So chat models like ChatGPT are actually trained to take a series of messages as input and return a model generated message as output. And so although the chat format is designed to make multi-turn conversations like this easy, we've kind of seen through the previous lessons that it's also just as useful for single-turn tasks without any conversation.

OpenAl API call

take a series of messages as input and then return a model generated message as output. So the user message is the input and then the assistant message is the output.

And so next we're going to define two helper functions. So get_completion function is the one that we've been using throughout all the lessons.

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Chat model

But if you look at it, we give a prompt but then inside the function what we're actually doing is putting this prompt into what looks like some kind of user message. And this is because the ChatGPT model is a chat model which means it's trained to

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output input input user

So in this lesson we're going to actually use a different helper function and instead of kind of putting a single prompt as input and getting a single completion, we're going to pass in a list of messages and these messages can be from a variety of
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different roles. So I'll describe those

In [2]: def get_completion(prompt, model="gpt-3.5-turbo"):
 messages = [{"role": "user", "content": prompt}]
 response = openai.ChatCompletion.create(

```
model=model,
messages=messages,
temperature=0, # this is the degree of randomness of the model's output
}
return response.choices[0].message["content"]

def get_completion_from_messages(messages, model="gpt-3.5-turbo", temperature=0):
    response = openai.ChatCompletion.create(
        model=model,
        messages=messages,
        temperature=temperature=temperature, # this is the degree of randomness of the model's output
}
# print(str(response.choices[0].message))
    return response.choices[0].message["content"]

Example 1

So here's an example of a list of messages and so the first message is a system message which gives an overall instruction and then after this message we have turns between the user and the assistant and this will continue to go on.
```

{"role": "system",

{"role": "assistant",

{"role": "user",

Role

messages =

```
And if you've ever used ChatGPT, the web interface, then your messages are the user messages and then ChatGPT's messages are the assistant messages. So the system message helps to set the behavior and persona of the assistant and it acts as a high-level instruction for the conversation. So you can think of it as whispering in the assistant's ear and kind of guiding its responses without the user being aware of the system message. So as the user, if you've ever used ChatGPT, you probably don't know what's in ChatGPT's system message. The benefit of the system message is that it provides you, the developer, with a way to frame the conversation without making the request itself part of the conversation. So you can guide the assistant and whisper in its ear and guide its responses without making the user aware.
```

print(response)

{'role':'user', 'content':'Hi, my name is Isa'}]

{'role':'system', 'content':'You are friendly chatbot.'},

{'role':'system', 'content':'You are friendly chatbot.'},

Is there anything I can help you with today?"},

print(response)

user

In [7]: def collect_messages(_):

inp.value = ''

panels.append(

import panel as pn # GUI
pn.extension()

greek salad 7.25 \

peppers 1.00 \ Drinks: \

Output

print(response)

"pizza": {

"size": "large",

extra cheese 2.00, \
mushrooms 1.50 \

Toppings: \

assistant

prompt = inp.value_input

{'role':'user', 'content':'Yes, can you remind me, What is my name?'}]

response = get_completion_from_messages(messages, temperature=1)

Output

"content": "You are an assistant... "},

"content": "Why did the chicken ... "},

"content": "tell me a joke "},

```
{'role':'assistant', 'content':'Why did the chicken cross the road'},
{'role':'user', 'content':'I don\'t know'} ]

In []: # response = get_completion_from_messages(messages, temperature=1)
# print(response)

So now let's try to use these messages in a conversation. So we'll use our new helper function to get the completion from the messages. And we're also using a higher temperature (temperature=1).

So the system message says, "You are an assistant that speaks like Shakespeare". So this is us kind of describing to the assistant how it should behave. And then the first user message is, "Tell me a joke". The next is, "Why did the chicken cross
```

It is an old and classic that never fails. So there's our Shakespearean response.

Example 2

the road?" And then the final user message is, "I don't know." So if we run this, the response is:

response = get_completion_from_messages(messages, temperature=1)

Verily, the reason for the fowl's crossing is this: To reach the other si

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So now let's do another example. So here our messages are, the system message is "You are a friendly chatbot", and the first user message is, "Hi, my name is Isa". And we want to get the first user message. So let's execute this for the first assistant message. And so the first message is, "Hello Isa! It's nice to meet you. How can I assist you today?"

In [4]: messages = [
{'role':'system', 'content':'You are friendly chatbot.'},
```

de, in pursuit of adventures new and diverse!

response = get_completion_from_messages(messages, temperature=1)
print(response)

actually know my name. So each conversation with a language model is a standalone interaction, which means that you must provide all relevant messages for the model to draw from in the current conversation.

Now let's try another example. So here our messages are system message is, "You are a friendly chatbot" and the first user message is, "Yes, can you remind me what is my "name? And let's get the response. And as you can see, the model doesn't

Output

messages = [

```
messages = [
    {'role':'system', 'content':'You are friendly chatbot.'},
    {'role':'user', 'content':'Yes, can you remind me, What is my name?'} ]
response = get_completion_from_messages(messages, temperature=1)
print(response)

I'm sorry, but as a chatbot, I don't have access to personal information
or memory.
```

{'role':'user', 'content':'Hi, my name is Isa'},
{'role':'assistant', 'content': "Hi Isa! It's nice to meet you. \
Is there anything I can help you with today?"},
{'role':'user', 'content':'Yes, you can remind me, what is my name

Example 4

In []: messages = [

In [6]: messages = [

print(response)

{'role':'user', 'content':'Yes, you can remind me, What is my name?'}]
response = get_completion_from_messages(messages, temperature=1)
print(response)

We can automate the collection of user prompts and assistant responses to build a OrderBot. The OrderBot will take orders at a pizza restaurant.

If you want the model to draw from, or quote unquote remember earlier parts of a conversation, you must provide the earlier exchanges in the input to the model. And so we'll refer to this as context. So let's try this.

```
# print(response)

So now we've kind of given the context that the model needs, which is my name, in the previous messages, and we'll ask the same question, so we'll ask what my name is. And the model is able to respond because it has all of the context it needs in this kind of list of messages that we input to it.

Output

| messages = [
| ('role':'system', 'content':'You are friendly chatbot.'),
| ('role':'system', 'content':'Hi, my name is Isa'),
| ('role':'assistant', 'content': 'Hi Isal It's nice to meet you. \
```

OrderBot

So now you're going to build your own chatbot. This chatbot is going to be called "OrderBot", and we're going to automate the collection of user prompts and assistant responses in order to build this "OrderBot". And it's going to take orders at a pizza

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restaurant, so first we're going to define the helper function ( collect_messages ), and what it is doing is that it's going to kind of collect our user messages so we can avoid typing them in by hand in the same way that we did above, and it is going to collect prompts from a user interface that we'll build below, and then append it to a list called "context", and then it will call the model with that context every time. And the model response is then also added to the context, so the kind of model message is added to the context, so on, so it just kind of grows longer and longer.

messages =

[
system
user
assistant
```

{'role':'user', 'content':'Yes, you can remind me, What is my name?'}]

response = get_completion_from_messages(messages, temperature=1)

return pn.Column(*panels)

And so now we'll set up and run this kind of UI to display the Autobot. And so here's the context, and it contains the system message that contains the menu. And note that every time we call the language model, we're going to use the same context, and the context is building up over time.

This way the model has the information it needs to determine what to do next.

context.append({'role':'user', 'content':f"{prompt}"})

context.append({'role':'assistant', 'content':f"{response}"})

pn.Row('User:', pn.pane.Markdown(prompt, width=600)))

pn.Row('Assistant:', pn.pane.Markdown(response, width=600, style={'background-color': '#F6F6F6'})))

response = get_completion_from_messages(context)

```
panels = [] # collect display
context = [ {'role':'system', 'content':"""
You are OrderBot, an automated service to collect orders for a pizza restaurant. \
You first greet the customer, then collects the order, \
and then asks if it's a pickup or delivery. \
You wait to collect the entire order, then summarize it and check for a final \
time if the customer wants to add anything else. \
If it's a delivery, you ask for an address. \
Finally you collect the payment.\
Make sure to clarify all options, extras and sizes to uniquely \
identify the item from the menu.\
You respond in a short, very conversational friendly style. \
The menu includes \
pepperoni pizza 12.95, 10.00, 7.00 \
cheese pizza 10.95, 9.25, 6.50 \
eggplant pizza 11.95, 9.75, 6.75 \
fries 4.50, 3.50 \
```

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sausage 3.00 \
canadian bacon 3.50 \
AI sauce 1.50 \
peppers 1.00 \
Drinks: \
coke 3.00, 2.00, 1.00 \
sprite 3.00, 2.00, 1.00 \
bottled water 5.00 \
"""} ] # accumulate messages
# inp = pn.widgets.TextInput(value="Hi", placeholder='Enter text here...')
# button_conversation = pn.widgets.Button(name="Chat!")
# interactive_conversation = pn.bind(collect_messages, button_conversation)
# dashboard = pn.Column(
       pn.Row(button_conversation),
       pn.panel(interactive_conversation, loading_indicator=True, height=300),
# dashboard
And then let's execute this.
   import panel as pn # GUI
   pn.extension()
   panels = [] # collect display
   context = [ {'role':'system', 'content':"""
   You are OrderBot, an automated service to collect orders for a pizza restaurant. \
   You first greet the customer, then collects the order, \
   and then asks if it's a pickup or delivery. \
   You wait to collect the entire order, then summarize it and check for a final \
   time if the customer wants to add anything else. \
   If it's a delivery, you ask for an address. \
   Finally you collect the payment.\
   Make sure to clarify all options, extras and sizes to uniquely \
   identify the item from the menu.\
   You respond in a short, very conversational friendly style. \
   The menu includes \
   pepperoni pizza 12.95, 10.00, 7.00 \
   cheese pizza 10.95, 9.25, 6.50 \
   eggplant pizza 11.95, 9.75, 6.75 \
   fries 4.50, 3.50 \
   greek salad 7.25 \
   Toppings: \
    extra cheese 2.00, '
   mushrooms 1.50 \
   sausage 3.00 \
   canadian bacon 3.50 \
   AI sauce 1.50 \
```

```
coke 3.00, 2.00, 1.00 \
                sprite 3.00, 2.00, 1.00 \
                bottled water 5.00 \
                """} ] # accumuLate messages
                inp = pn.widgets.TextInput(value="Hi", placeholder='Enter text here...')
                button_conversation = pn.widgets.Button(name="Chat!")
                interactive_conversation = pn.bind(collect_messages, button_conversation)
                dashboard = pn.Column(
                    pn.Row(button_conversation),
                    pn.panel(interactive_conversation, loading_indicator=True, height=300),
                dashboard
                  Enter text here...
                Assistant: Hello! Welcome to our pizza restaurant. How can I assist you today?
                User: Hi I would like to order a pizza
                Assistant: Great! I'd be happy to help you with that. What kind of pizza would you like to order? We have
                         pepperoni, cheese, and eggplant pizza.
                User: how much are they
                Assistant: Sure! The prices for our pizzas are as follows:

    Pepperoni Pizza: $12.95 for a large, $10.00 for a medium, and $7.00 for a small.

                          . Cheese Pizza: $10.95 for a large, $9.25 for a medium, and $6.50 for a small.
                          . Eggplant Pizza: $11.95 for a large, $9.75 for a medium, and $6.75 for a small.
                         Let me know which size and type you'd like to order!
                User: large cheese pizza please
                Assistant: Great choice! A large cheese pizza is $10.95. Is there anything else you would like to add to your
                         order, such as toppings, sides, or drinks?
                User: i would like to add pepperoni and olives as topping with extra cheese and a drink please
                Assistant: Absolutely! So, you'd like a large cheese pizza with pepperoni, olives, and extra cheese. And for
                         the drink, which one would you like? We have coke, sprite, and bottled water available.
                User: yes, that would be it. Give me a receipt for it please
                Assistant: Of course! Here's a summary of your order:

    Large Cheese Pizza with Pepperoni, Olives, and Extra Cheese: $10.95

    Drink: Coke: $3.00

                         Your total comes to $13.95. Is there anything else I can assist you with?
                User: Thank you OrderBot
                Assistant: You're welcome! If you have any other questions or need further assistance, feel free to ask.
            Now we can ask the model to create a JSON summary that we could send to the order system based on the conversation. So, we're now appending another system message, which is an instruction, and we're saying:
             Create a JSON summary of the previous food order. Itemize the price for each item. The fields should be 1) pizza, include side 2) list of toppings 3) list of drinks, include size 4)
            list of sides 5) and finally, the total price.
            And you could also use a user message here ('role':'system'). This ('role':'system') does not have to be a system message.
In [10]: messages = context.copy()
             {'role':'system', 'content':'create a json summary of the previous food order. Itemize the price for each item\
             The fields should be 1) pizza, include size 2) list of toppings 3) list of drinks, include size 4) list of sides include size 5)total price '},
```

a lower temperature as well because for a customer's assistant chatbot, you might want the output to be a bit more predictable as well. So, let's execute this.

#The fields should be 1) pizza, price 2) list of toppings 3) list of drinks, include size include price 4) list of sides include size include price, 5)total price '},

response = get_completion_from_messages(messages, temperature=0)

print(response)

And notice, in this case, we're using a lower temperature because for these kinds of tasks, we want the output to be fairly predictable. For a conversational agent, you might want to use a higher temperature. However, in this case, I would maybe use

```
messages = context.copy()
messages.append(
    {'role':'system', 'content':'create a json summary of the previous food order. Itemize the price
    The fields should be 1) pizza, include size 2) list of toppings 3) list of drinks, include size
)
    #The fields should be 1) pizza, price 2) list of toppings 3) list of drinks, include size inclu
response = get_completion_from_messages(messages, temperature=0)
```

```
"type": "cheese"
}
toppings": [
"pepperoni",
"olives",
"extra cheese"
},
"drinks": [
"type": "coke",
"size": "regular"
],
"stdes": [],
"total_price": 13.95

And so, here we have the summary of our order. And so, we could submit this to the order system if we wanted to. So there we have it, you've built your very own order chatbot. Feel free to kind of customise it yourself and play around with the system message to change the behaviour of the chatbot and kind of get it to act as different personas with different knowledge.
```

Try experimenting on your own!

You can modify the menu or instructions to create your own orderbot!