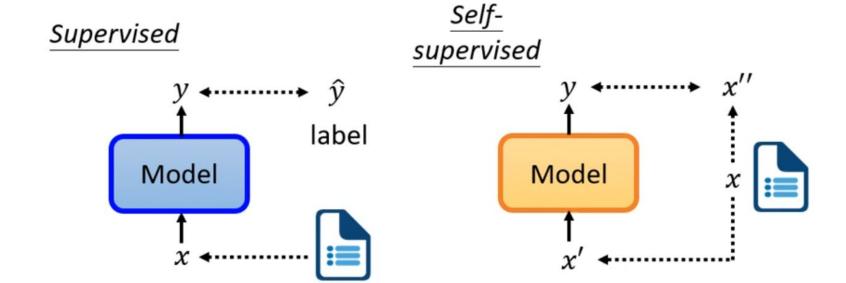
#### 大型語言模型

課程須知

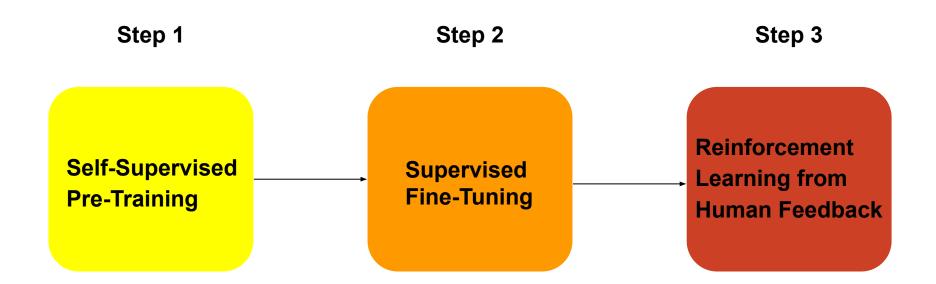
助教:蔡明晏、林言翰、紀宇烜、葉志銓、范姜伯軒、陳冠榮

#### What is LLM?

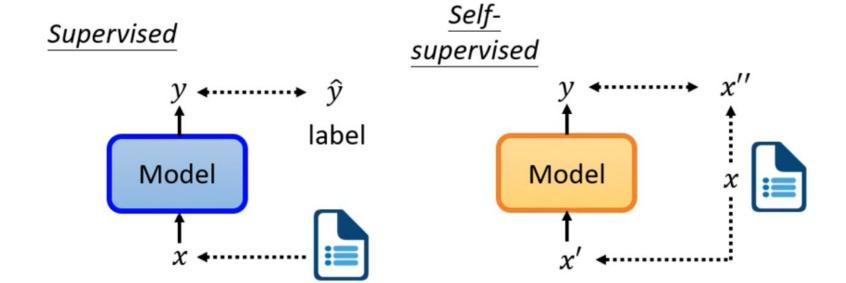
#### How to train LLM?



#### How to train LLM?



#### How to train model?



#### 大綱

- 評分方式
- 課程進度表
- 作業/專案規範
- 助教時間
- 聯絡方式
- 自學資源

#### 評分方式

- Homework (or Task Competition) (60%)
- Final Project (40%)
- Class Attendance (+10%)

#### 課程進度表

		每週進度表
週次	上課日期	課程進度、內容、主題
1	2025-09-05(五)	Feedforward and Convolutional Neural Networks
2	2025-09-12(五)	Regularization for Optimization in Deep Learning
3	2025-09-19(五)	Recurrent Neural Network and Sequential Learning
4	2025-09-26(五)	N-Gram Language Models and Topic-Based Language Models (Final project grouping)
5	2025-10-03(五)	RNN Language Models and Language Understanding $(HW1\ announcement)$
6	2025-10-10(五)	Attention Networks and Transformers
7	2025-10-17(五)	State-Space Language Models
8	2025-10-24(五)	BERT Encoder and GPT Decoder
9	2025-10-31(五)	Retrieval, Augmentation and Generation (Final project proposal)
10	2025-11-07(五)	Generation with Prompting Strategies (HW2 announcement)
11	2025-11-14(五)	LLMs with GPT and LLaMA
12	2025-11-21(五)	LLM Model Trends and Generative Al
13	2025-11-28(五)	Final Presentation
14	2025-12-05(五)	Al Computing Architecture for LLM
15	2025-12-12(五)	Final Presentation
16	2025-12-19(五)	DaVinci GAI Platform and Applications

#### 作業規範

- 使用Python撰寫
- 助教會幫忙釐清題意觀念, 但不會幫忙 Debug。
- 需要繳交程式碼、結報。
- 繳交期限大約兩星期,可以補交,但分數會打折
- 禁止抄襲

#### 專案規範

- 專題範圍為課程內容
- 3-4人一組
- 每組需要先繳交一份 proposal
- 助教會先針對各組提交的 proposal 給予實作上的建議讓同學參考
- 若提出的 proposal 在實作上有困難, 跟助教討論後可調整 proposal 的內容, 但 最晚須在 final project 報告前兩周寄信告知助教

#### 助教時間

- 實體詢問
  - 週三19:00-20:00 工程四館708
  - 欲前往請提前寄信告知, 並描述問題 內容, 以利助教能更快速的協助解惑

#### • 線上詢問

○ NewE3討論區優先,其次為助教電子信箱 nycumllab@gmail.com

#### 聯絡方式

- 實體
  - 工程四館708
- 電子信箱
  - o nycumllab@gmail.com

#### 自學資源

- Kaggle
- Pytorch Tutorials
- Hugging Face
- <u>iThome</u>
- 知乎
- scikit-learn
- NumPy Doc.
- Deep Learning.(Goodfellow)

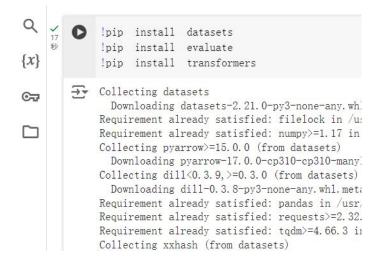
# Google COOCOO

● 筆記本都會建置在Google Drive中



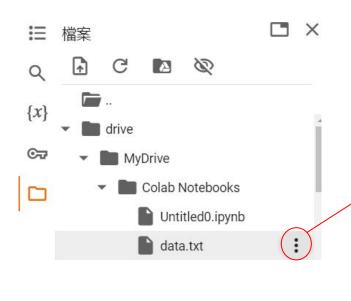


● 套件安裝 & 雲端存取



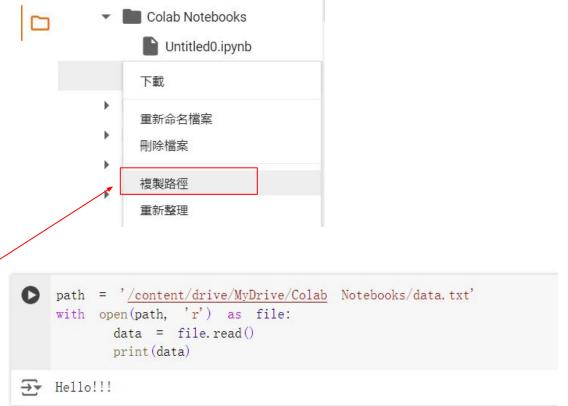


• 載入資料集

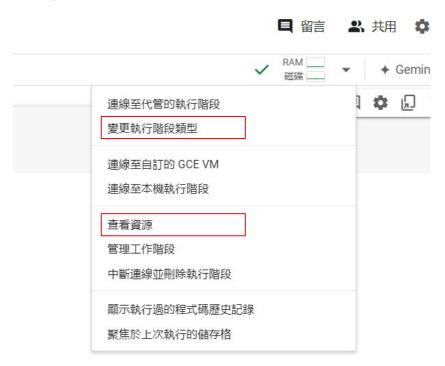


©₽

MyDrive



● 使用GPU

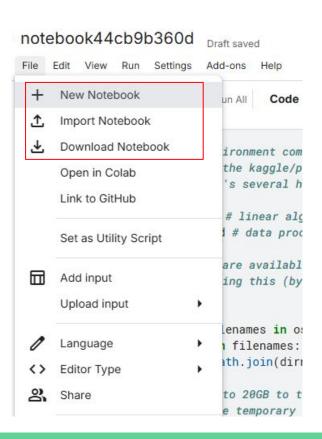




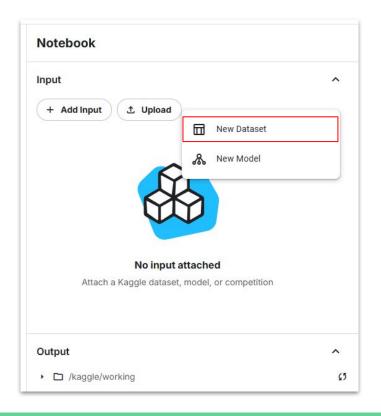


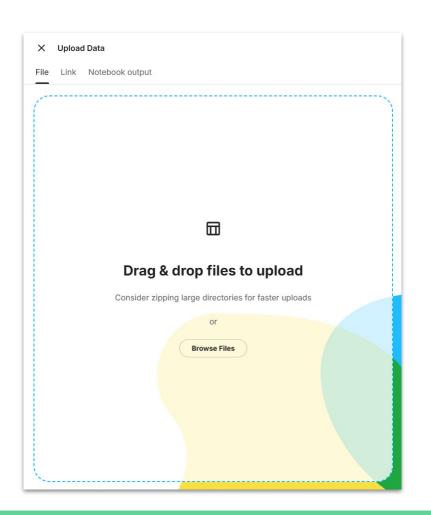


● 新增 / 上傳筆記本

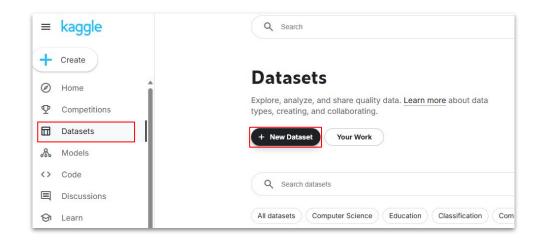


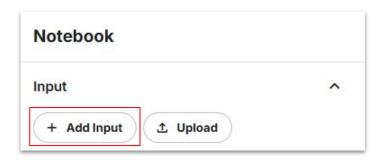
● 上傳資料集 (方法一)

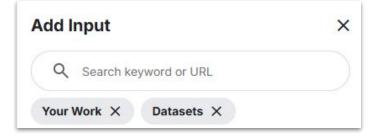




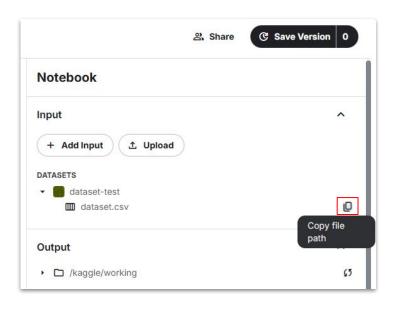
● 上傳資料集 (方法二)





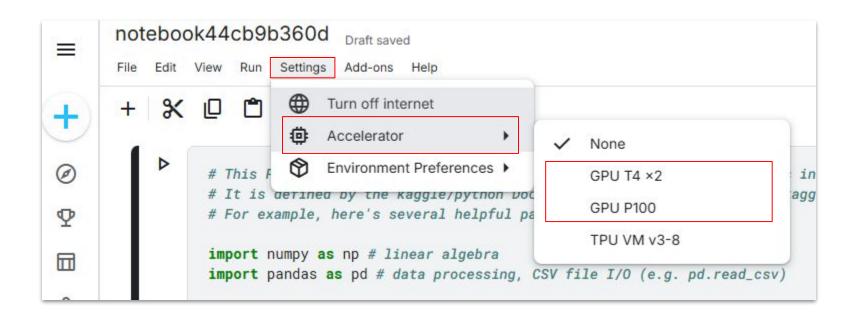


• 引入資料集



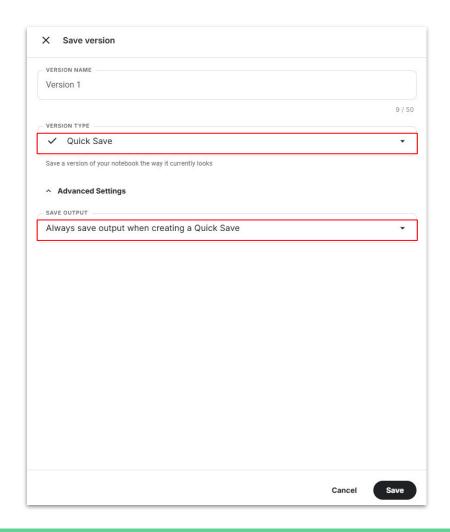
```
dataset = pd.read_csv("/kaggle/input/dataset-test/dataset.csv")
+ Code + Markdown
```

● GPU設定



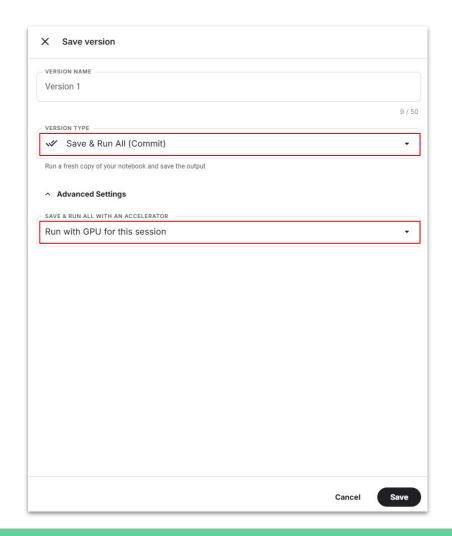
- 儲存版本 (Quick Save)
  - 保存當下所有的程式及輸出 (不執行)



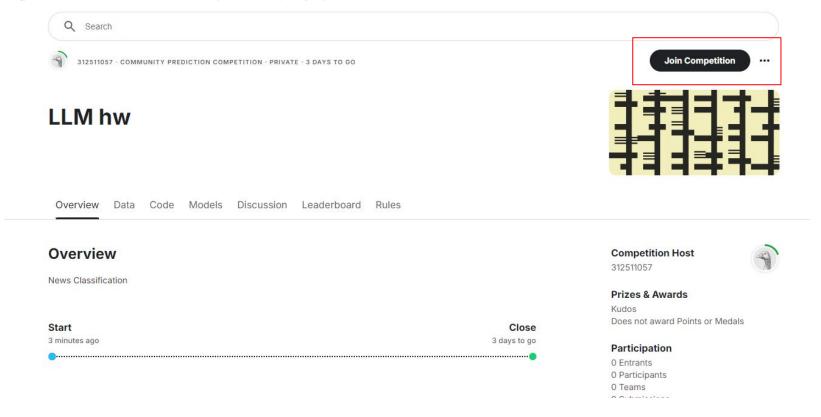


- 儲存版本 (Run & Save)
  - 執行所有程式且保存輸出 (離線執行)

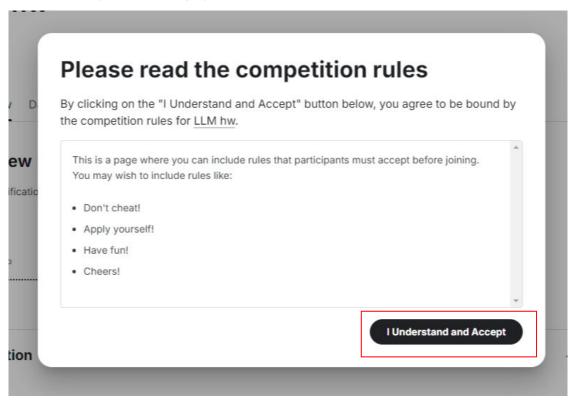




● 使用 Invitation Link, 加入競賽



● 使用 Invitation Link, 加入競賽



Team Name

#### LLM hw

Overview Data Code Models Discussion Leaderboard Rules Team Submissions **Your Team** Everyone that competes in a Competiton does so as a team - even if you're competing by yourself. Learn more. General TEAM NAME 123456789 This name will appear on your team's leaderboard position. Let others know you're looking for teammates Your team can't accept more team members. Team Members Your team is at maximum capacity. Great job! Save Changes

Data Download



#### Dataset Description

#### **File Description**

train.jsonl - training set

test.jsonl - testing set for the leaderboard ranking and final scoring

sample\_submission.csv - the sample submission file in the correct format

#### **Data Fields**

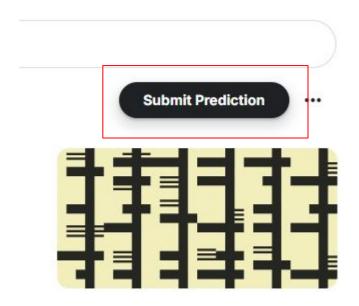
id - an anonymous id unique to a given news title

category - the label category of the news title

headline - the title of the news

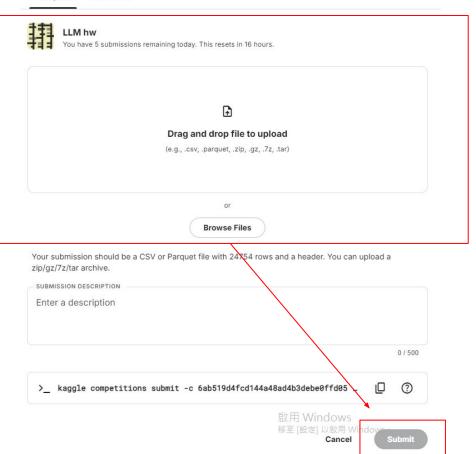
short\_description - news content

Submit Prediction

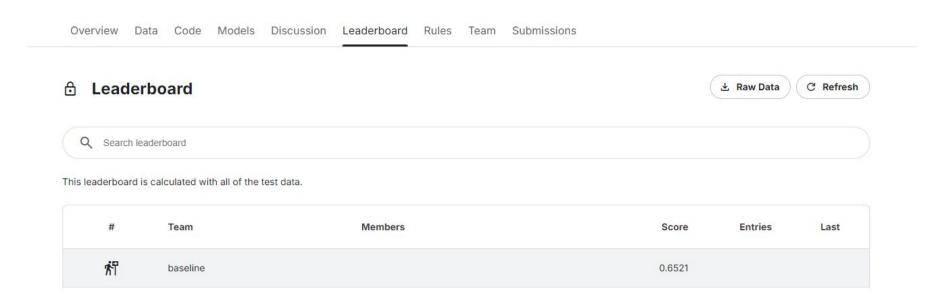


#### X Submit to Competition

File Upload Notebook



Leaderboard



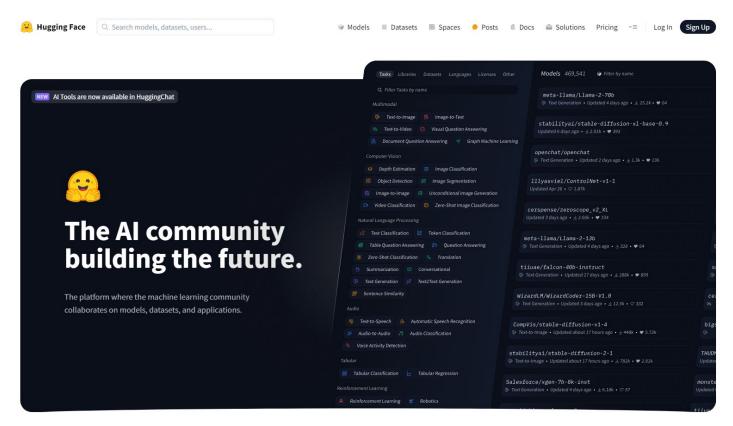
#### 常見問題

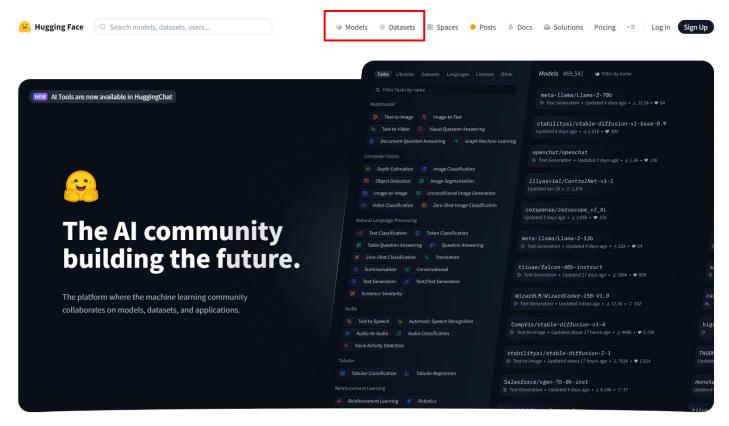
- 沒有GPU相關設定
  - 需註冊/登入帳號且綁定電話才能使用 GPU
- GPU使用時間限制
  - 一週30小時 (週六更新), 可至首頁右上角頭像 查詢
- 程式下載資源時出現錯誤
  - 上方Settings開啟網路功能



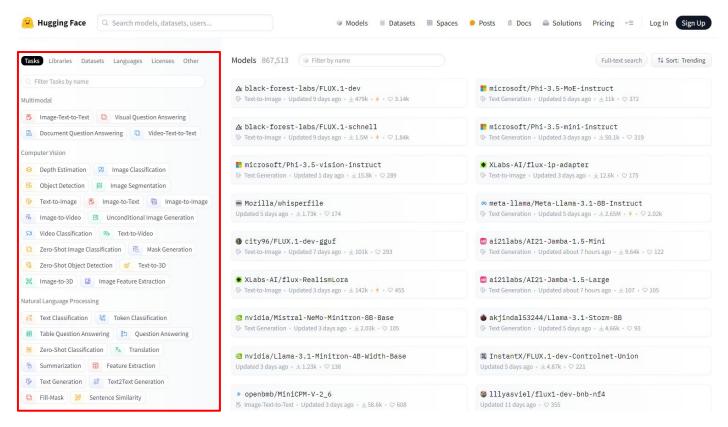
### Hugging Face

#### Hugging Face How to find the dataset/model

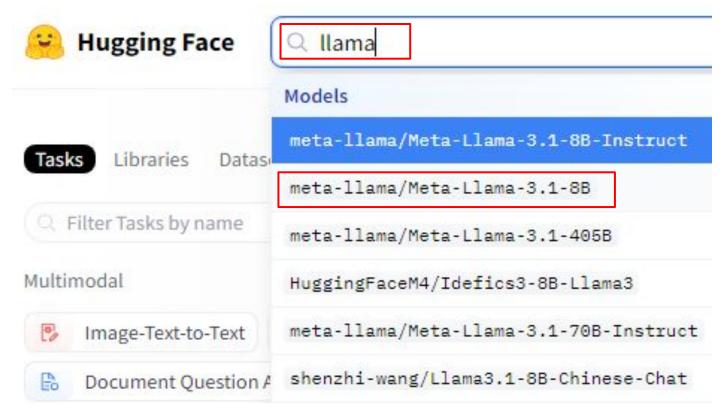




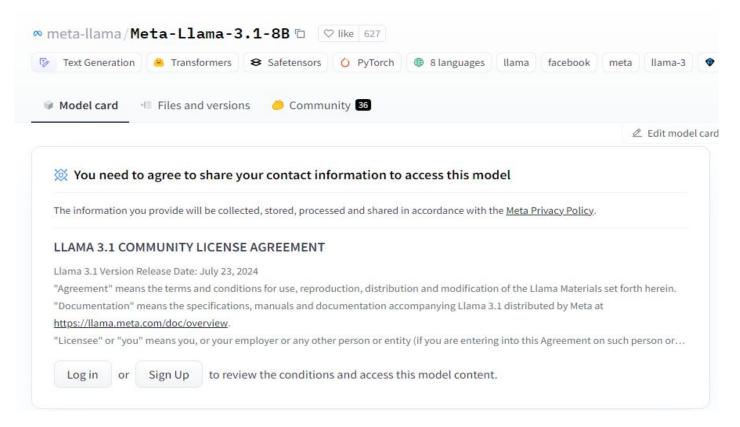
1-2 Click (Models) or (Datasets)



1-3 (Option) Click on the task option to filter your choices

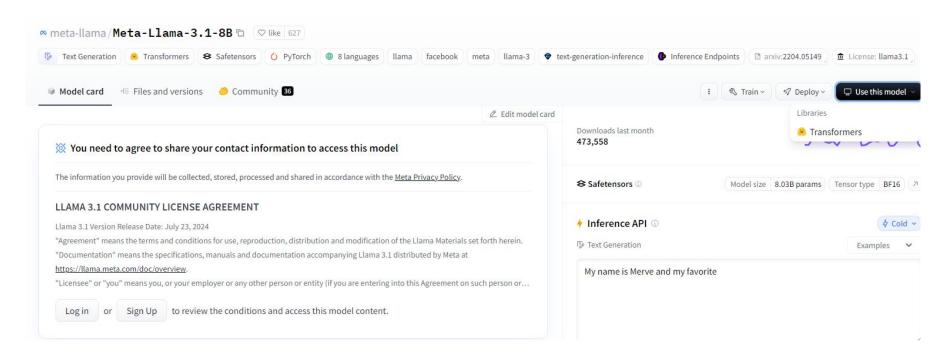


1-4 Enter your model/dataset name and click on the model you want to select



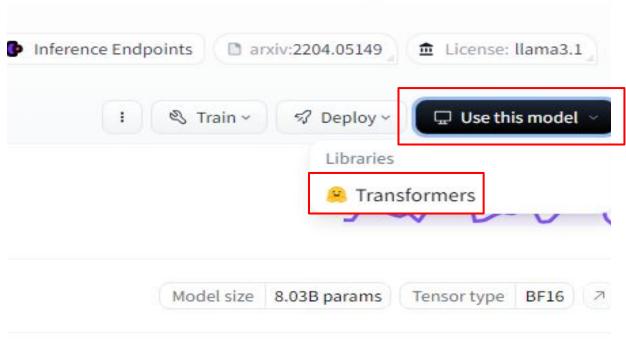
1-5 You successfully found the model/dataset you wanted!

#### Hugging Face How to use the model



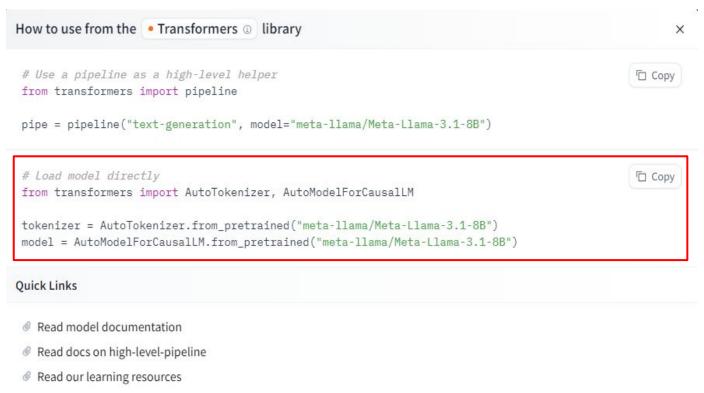
2-1 Switch to the model page

#### Hugging Face How to use the model



2-2 Click (Use this model) and click the libaraies

# Hugging Face How to use the model



2-3 You can copy the code below the words "Load model directly" to use it!

# Hugging Face Reference link

- https://huggingface.co/
- 官方教程
  - https://huggingface.co/learn/nlp-course/chapter1/1
  - https://www.youtube.com/watch?v=QEaBAZQCtwE

# Natural Language Processing

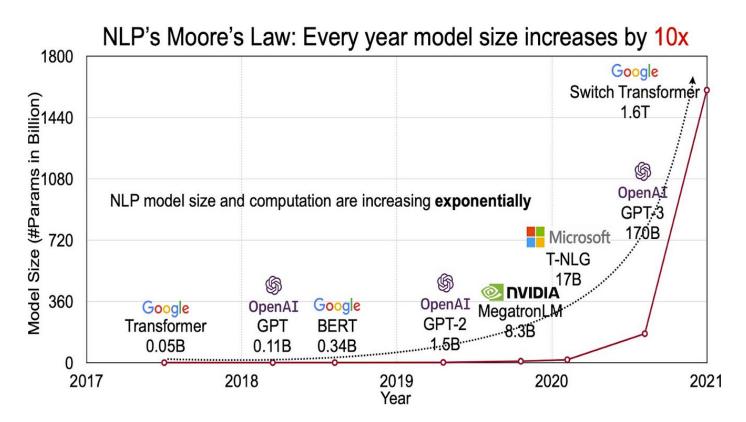
#### What is NLP?

- Classifying each word in a sentence
- Classifying whole sentences
- Extracting an answer from a text
- Generating a new sentence from an input text
- Generating text content

#### LM History

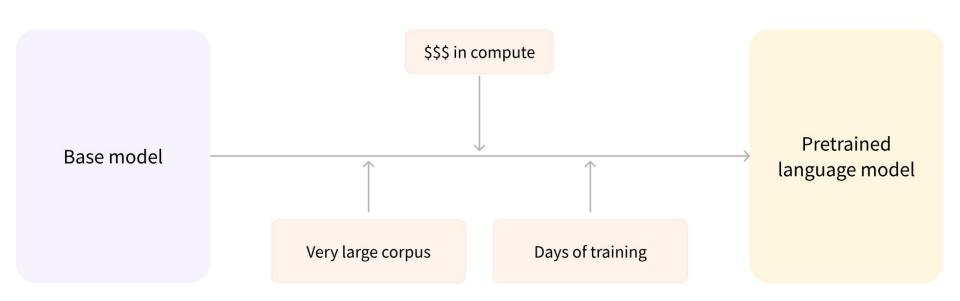


#### LM Size



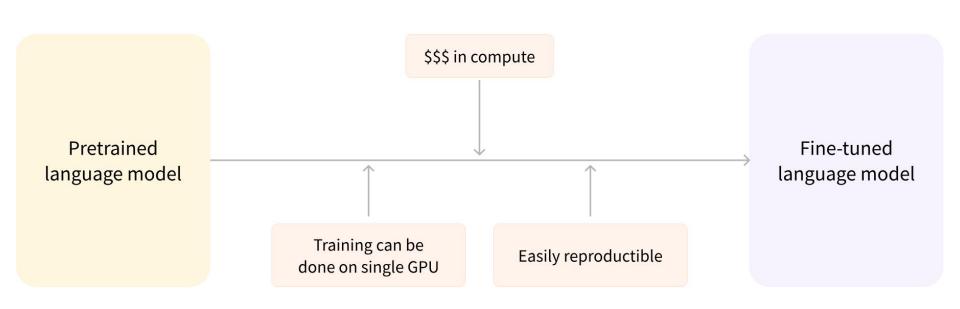
#### Transfer Learning

• Pretraining: training a model from scratch

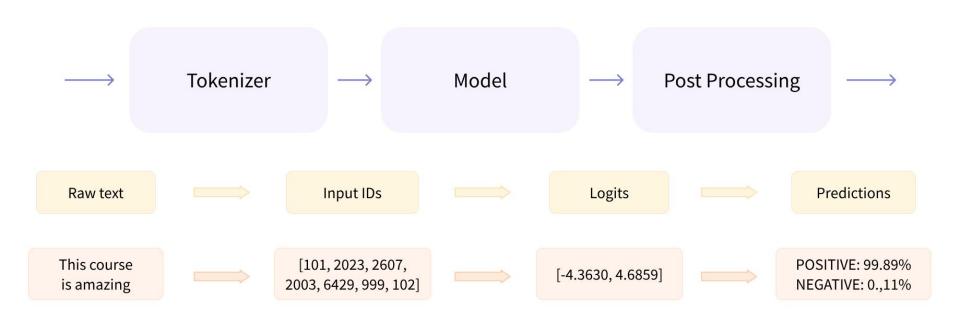


#### Transfer Learning

• Fine-tuning: the training done after a model has been pretrained

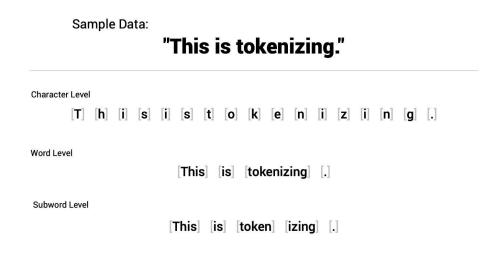


# NLP Task (Sentiment Analysis)



#### Tokenizer

- Splitting the input into words, subwords, or symbols (like punctuation) that are called tokens
- Mapping each token to an integer
- Adding additional inputs that may be useful to the model

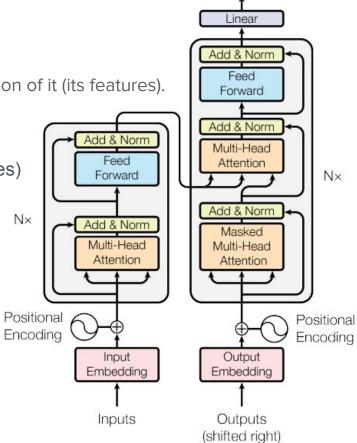


#### Tokenizer

```
from transformers import AutoTokenizer
checkpoint = "distilbert-base-uncased-finetuned-sst-2-english"
tokenizer = AutoTokenizer.from_pretrained(checkpoint)
raw inputs = [
   "I've been waiting for a HuggingFace course my whole life.",
   "I hate this so much!",
inputs = tokenizer(raw_inputs, padding=True, truncation=True, return_tensors="pt")
print(inputs)
  'input_ids': tensor([
     [ 101, 1045, 1005, 2310, 2042, 3403, 2005, 1037, 17662, 12172, 2607, 2026, 2878, 2166, 1012, 102],
     [ 101, 1045, 5223, 2023, 2061, 2172, 999, 102, 0, 0, 0, 0, 0, 0,
                                                                                               07
  ]),
  'attention mask': tensor([
     [1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0]
  1)
```

#### Transformer

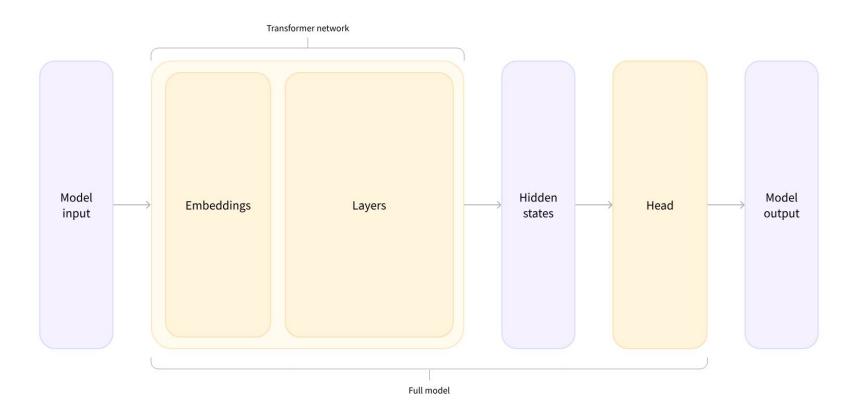
- Encoder
  - The encoder receives an input and builds a representation of it (its features).
  - BERT, RoBERTa
- Decoder
  - The decoder uses the encoder's representation (features) along with other inputs to generate a target sequence.
  - o GPT, GPT-2



Output Probabilities

Softmax

#### Model



#### Model

```
from transformers import AutoModelForSequenceClassification

checkpoint = "distilbert-base-uncased-finetuned-sst-2-english"

model = AutoModelForSequenceClassification.from_pretrained(checkpoint)

outputs = model(**inputs)
```

```
print(outputs.logits)
```

```
tensor([[-1.5607, 1.6123],
        [ 4.1692, -3.3464]], grad_fn=<AddmmBackward>)
```

#### Postprocessing the Output

- "I've been waiting for a HuggingFace course my whole life."
  - NEGATIVE: 0.0402, POSITIVE: 0.9598
- "I hate this so much!"
  - NEGATIVE: 0.9995, POSITIVE: 0.0005

```
import torch

predictions = torch.nn.functional.softmax(outputs.logits, dim=-1)
print(predictions)
```

```
tensor([[4.0195e-02, 9.5980e-01],
        [9.9946e-01, 5.4418e-04]], grad_fn=<SoftmaxBackward>)
```

#### Fine-tuning a model with the Trainer API

Loading a dataset

```
from datasets import load dataset
from transformers import AutoTokenizer, DataCollatorWithPadding
raw_datasets = load_dataset("glue", "mrpc")
checkpoint = "bert-base-uncased"
tokenizer = AutoTokenizer.from pretrained(checkpoint)
def tokenize function(example):
    return tokenizer(example["sentence1"], example["sentence2"], truncation=True)
tokenized_datasets = raw_datasets.map(tokenize_function, batched=True)
data_collator = DataCollatorWithPadding(tokenizer=tokenizer)
```

# Fine-tuning a model with the Trainer API

#### Training

```
from transformers import TrainingArguments
training_args = TrainingArguments("test-trainer")
from transformers import AutoModelForSequenceClassification
model = AutoModelForSequenceClassification.from_pretrained(checkpoint, num_labels=2)
from transformers import Trainer
trainer = Trainer(
    model,
    training_args,
    train dataset=tokenized datasets["train"],
    eval_dataset=tokenized_datasets["validation"],
    data_collator=data_collator,
    tokenizer=tokenizer,
trainer.train()
```

#### Fine-tuning a model with the Trainer API

Evaluation

```
predictions = trainer.predict(tokenized datasets["validation"])
print(predictions.predictions.shape, predictions.label_ids.shape)
import numpy as np
preds = np.argmax(predictions.predictions, axis=-1)
import evaluate
metric = evaluate.load("glue", "mrpc")
metric.compute(predictions=preds, references=predictions.label ids)
{'accuracy': 0.8578431372549019, 'f1': 0.8996539792387542}
```

# RAG and LangChain

# 檢索增強生成(RAG)

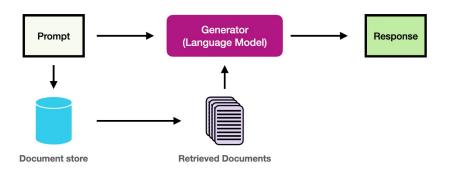
目的:解決特定外部知識問題

**你** 請你介紹今年**4/3**發生在台灣的大地震

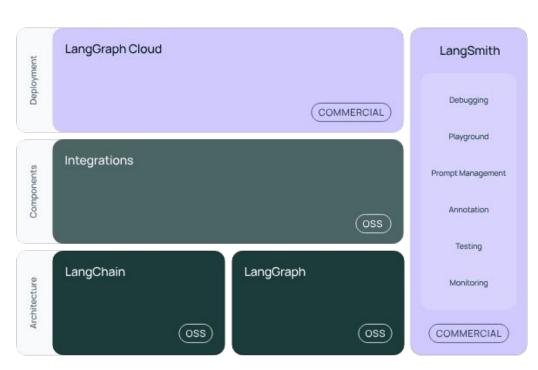
#### 

抱歉,我無法提供關於最近的自然災害或其他事件的即時資訊。你可以通過查閱新聞來獲取有關任何事件的最新信息,或者你有其他想知道的事情嗎?我很樂意幫助你。

#### Naïve RAG

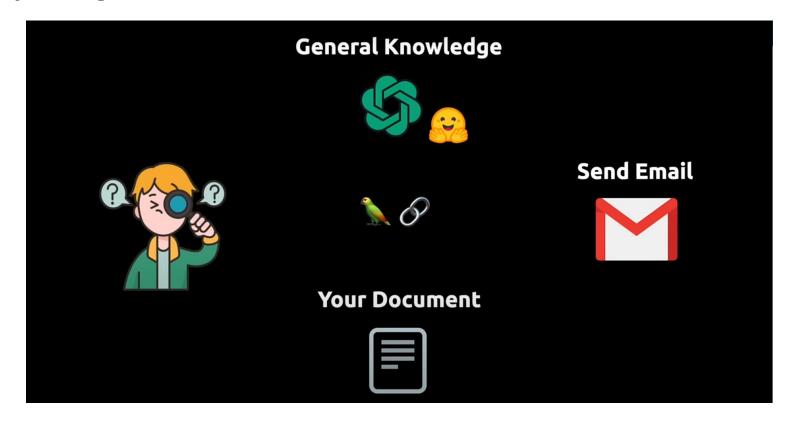


#### Langchain v0.2

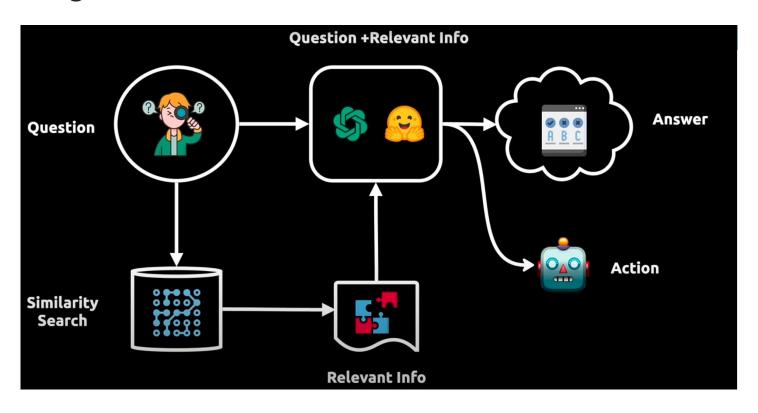


- 開發:使用 LangChain 的開源元件、LangGraph和第三方的package來構建應用程式。
- **生產化**:使用 LangSmith 檢查、監控和評估 langchain 程式,以便持續優化並部署。
- **部署**:通過 LangGraph Cloud 將 LangGraph 應用程式轉換為 API。

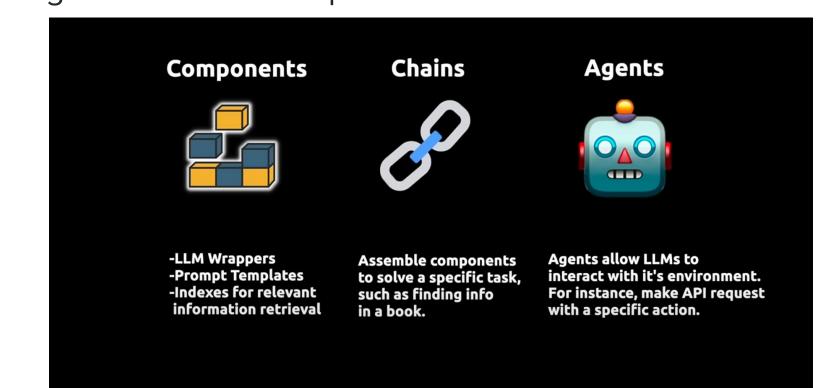
# Why Langchain?



# Why Langchain?



#### LangChain main concepts



#### RAG two main concepts

 Indexing: a pipeline for ingesting data from a source and indexing it. This usually happens offline.

2. **Retrieval** and **generation**: the actual RAG chain, which takes the user query at run time and retrieves the relevant data from the index, then passes that to the model.

#### Indexing: Load

```
import bs4
from langchain_community.document_loaders import WebBaseLoader
# Only keep post title, headers, and content from the full HTML.
bs4_strainer = bs4.SoupStrainer(class_=("post-title", "post-header", "post-content"))
loader = WebBaseLoader(
    web_paths=("https://lilianweng.github.io/posts/2023-06-23-agent/",),
    bs kwargs={"parse only": bs4 strainer},
docs = loader.load()
len(docs[0].page_content)
```

#### Indexing: Load

```
print(docs[0].page_content[:500])
      LLM Powered Autonomous Agents
Date: June 23, 2023 | Estimated Reading Time: 31 min | Author: Lilian Weng
Building agents with LLM (large language model) as its core controller is a cool concept. Several proof-
Agent System Overview#
In
```

## Indexing: Split

```
from langchain_text_splitters import RecursiveCharacterTextSplitter

text_splitter = RecursiveCharacterTextSplitter(
    chunk_size=1000, chunk_overlap=200, add_start_index=True
)
all_splits = text_splitter.split_documents(docs)

len(all_splits)
```

# Indexing: Split

```
66
len(all_splits[0].page_content)
969
all_splits[10].metadata
{'source': 'https://lilianweng.github.io/posts/2023-06-23-agent/',
 'start_index': 7056}
```

## Indexing: Store

```
from langchain_chroma import Chroma
from langchain_openai import OpenAIEmbeddings

vectorstore = Chroma.from_documents(documents=all_splits, embedding=OpenAIEmbeddings())
```

#### Retrieval and Generation: Retrieve

```
retriever = vectorstore.as_retriever(search_type="similarity", search_kwargs={"k": 6})
retrieved docs = retriever.invoke("What are the approaches to Task Decomposition?")
len(retrieved docs)
6
print(retrieved_docs[0].page_content)
Tree of Thoughts (Yao et al. 2023) extends CoT by exploring multiple reasoning possibilities at each ste
Task decomposition can be done (1) by LLM with simple prompting like "Steps for XYZ.\n1.", "What are the
```

#### Retrieval and Generation: Generate

```
import getpass
import os

os.environ["OPENAI_API_KEY"] = getpass.getpass()

from langchain_openai import ChatOpenAI

llm = ChatOpenAI(model="gpt-4o-mini")
```

#### Retrieval and Generation: Generate

```
from langchain import hub
prompt = hub.pull("rlm/rag-prompt")
example_messages = prompt.invoke(
    {"context": "filler context", "question": "filler question"}
).to_messages()
example messages
[HumanMessage(content="You are an assistant for question-answering tasks. Use the following pieces of re
print(example_messages[0].content)
You are an assistant for question-answering tasks. Use the following pieces of retrieved context to answering
Question: filler question
Context: filler context
Answer:
```

#### Retrieval and Generation: Generate

```
def format_docs(docs):
    return "\n\n".join(doc.page_content for doc in docs)

rag_chain = (
    {"context": retriever | format_docs, "question": RunnablePassthrough()}
    | prompt
    | llm
    | StrOutputParser()
)

rag_chain.invoke("What is Task Decomposition?")
```

'Task Decomposition is a process where a complex task is broken down into smaller, simpler steps or sub-

# 加簽

- 會從表單中抽十位同學加簽
- 注意加簽的同學期中不能退選,請同學 慎重考慮

