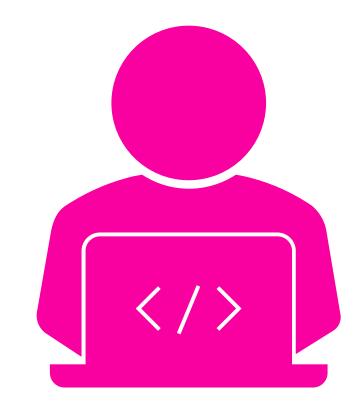
#### Create Machine Learning projects with Python (HuggingFace & Gradio)

Dave Appadoo

Presentation for PYMUG



Note: The slides with recording are available at:

https://daveappadoo.com/how-to-use-hugging-face-gradio-to-create-simple-machine-learning-ml-projects/

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#### About me

- +Software Developer at Accenture
- +Consultant Data Scientist
- +daveappadoo.com









#### Agenda

#### What is Hugging Face ??

- Hugging Face model zoo
- How to use the Hugging Face library
- Hugging Face demo

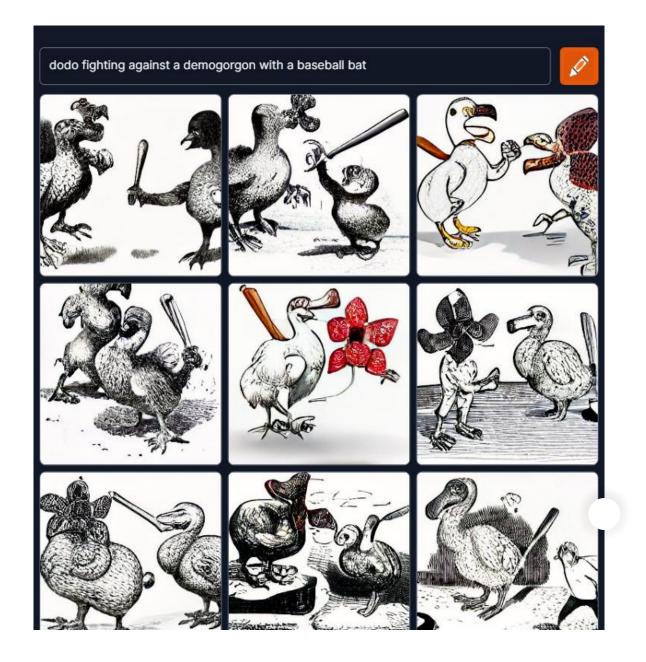
#### What's Gradio

- Using Gradio
- Gradio Demo

#### End

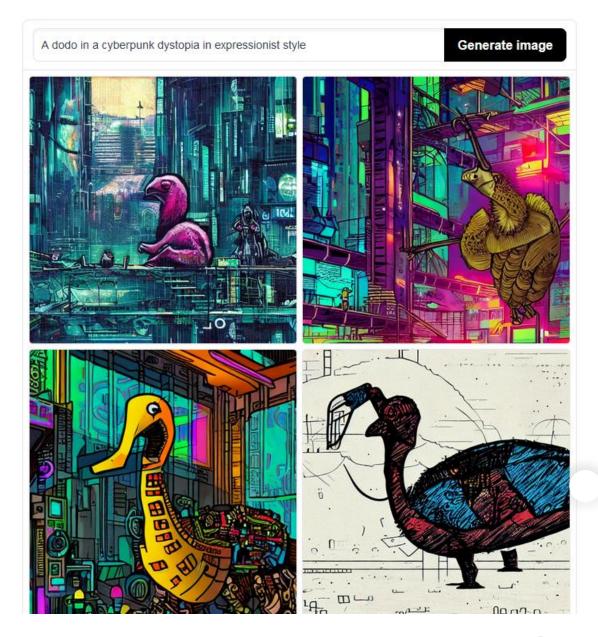
# Have you seen this recently?

- + Craiyon (formerly DALL·E mini)
- + an Al model that can draw images from any text prompt!



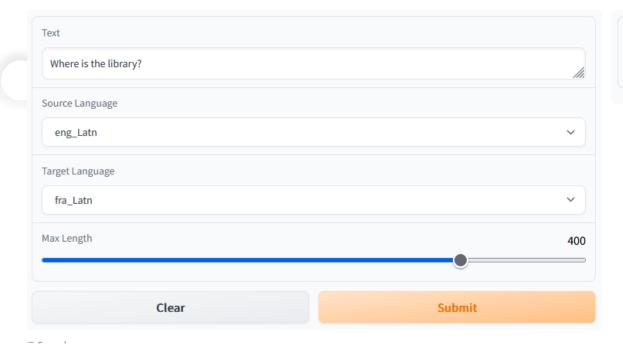
# Have you seen this recently?

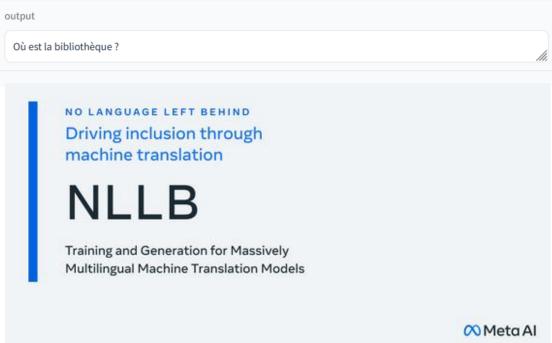
- + Stable Diffusion
- + state of the art text-to-image model that generates images from text!



# Have you seen this recently?

- + NLLB TRANSLATION Demo
- + state of the art text translation between ~200 languages!





## Creating simple ML projects back then

Steps	Framework	Language
Train model or use pretrained models	TensorFlow PYT ORCH	Python
2. Containerize and deploy the model	docker Flask web development, one drop at a time	Bash
3. Store incoming samples	MySQL	SQL
4. Build an interactive user interface	HTML JS CSS	HTML, JS, CSS

<sup>\*</sup>Taken from https://docs.google.com/presentation/d/1TXw48MjZFvkVur6rE0tUrEECf76SgUrglTjv0Kb0SWQ/edit?usp=sharing

### Creating simple ML projects now

Steps	Framework	Language
1. Train model or use pretrained models	TensorFlow PYT CH	Python
2. Containerize and deploy the model		Python
3. Store incoming samples	gradio	Python
4. Build an interactive user interface		Python

<sup>\*</sup>Taken from https://docs.google.com/presentation/d/1TXw48MjZFvkVur6rE0tUrEECf76SgUrglTjv0Kb0SWQ/edit?usp=sharing

## Step 1 in creating a ML project

- +Train a model or use a pre-trained model
- +Pre-trained models come in either TensorFlow or PyTorch.

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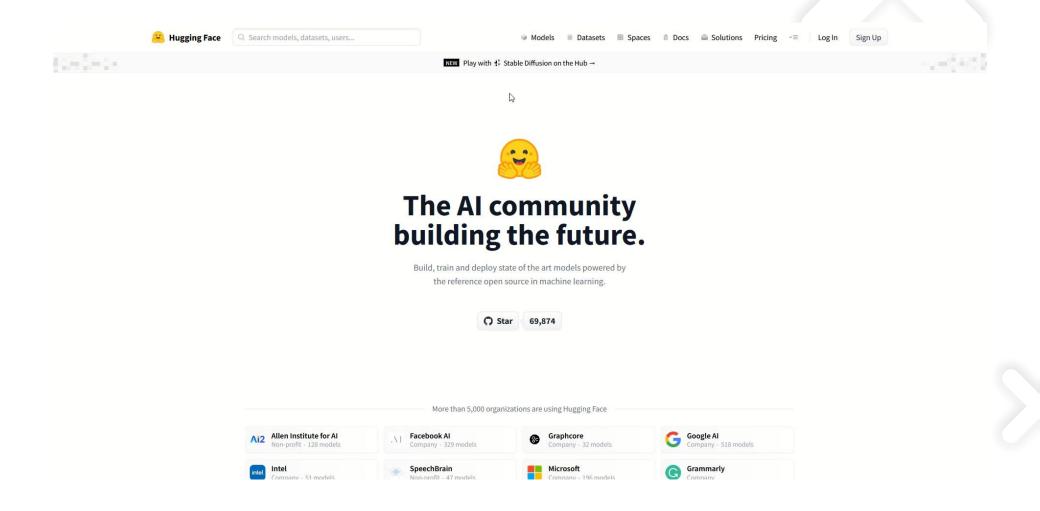
#### Step 1 in creating a ML project

- +Train a model or use a pre-trained model.
- +Pre-trained models come in either TensorFlow or PyTorch.
- +Murphy's law guaranteed:
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- +Hugging Face to the rescue with its model hub!

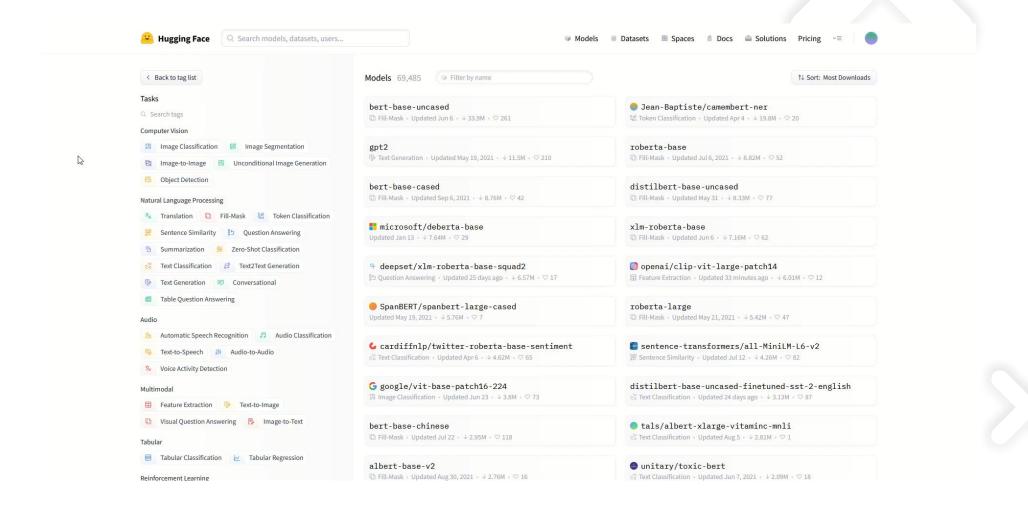
#### What is Hugging Face 😭 ?

- +Was known mostly for its Transformers library
  oit's open source, it supports both TensorFlow and PyTorch
- +Now a community and data science platform
  - oenable users to build, train and deploy ML models based on open-source code.
- +Think of what GitHub is to software development.
- +Hugging Face is like that but to ML.

### The Hugging Face model zoo



### Trying out a Object Detection model



#### Creating a simple UI for the model

```
from transformers import DetrFeatureExtractor, DetrForObjectDetection # The hugging face library
import torch # PyTorch library
from PIL import Image # Image module library
import requests # HTTP library
url = "http://images.cocodataset.org/val2017/000000039769.jpg"
image = Image.open(requests.get(url, stream=True).raw)
feature extractor = DetrFeatureExtractor.from pretrained("facebook/detr-resnet-50")
model = DetrForObjectDetection.from pretrained("facebook/detr-resnet-50")
inputs = feature extractor(images=image, return tensors="pt")
outputs = model(**inputs)
# convert outputs (bounding boxes and class logits) to COCO API.
target sizes = torch.tensor([image.size[::-1]])
results = feature extractor.post process(outputs, target sizes=target sizes)[0]
# Output the prediction.
for score, label, box in zip(results["scores"], results["labels"], results["boxes"]):
    box = [round(i, 2) for i in box.tolist()]
    # let's only keep detections with score > 0.9
   if score > 0.9:
        print(
            f"Detected {model.config.id2label[label.item()]} with confidence "
            f"{round(score.item(), 3)} at location {box}"
```



Detected remote with confidence 0.998 at location [40.16, 70.81, 175.55, 117.98]
Detected remote with confidence 0.996 at location [333.24, 72.55, 368.33, 187.66]
Detected couch with confidence 0.995 at location [-0.02, 1.15, 639.73, 473.76]
Detected cat with confidence 0.999 at location [13.24, 52.05, 314.02, 470.93]
Detected cat with confidence 0.999 at location [345.4, 23.85, 640.37, 368.72]

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- +Now we have the model
- +How to serve it to a user?
- +The user should not be editing source code to change image file.

### Creating a simple UI for the model

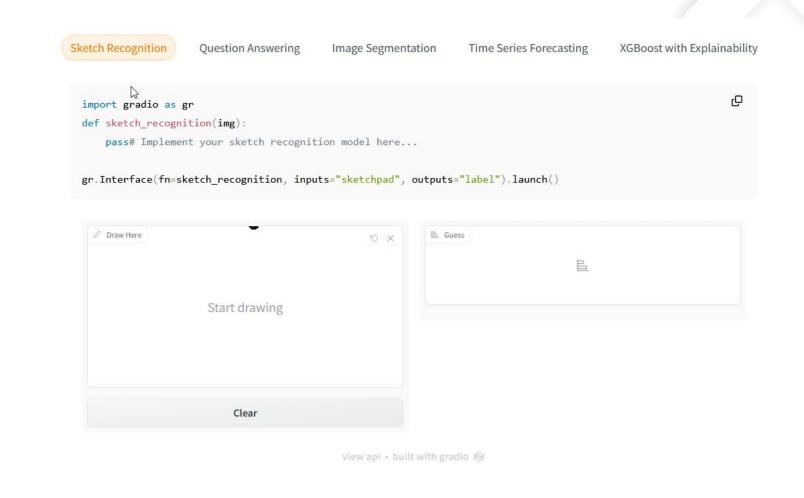
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- +Now we have the model
- +How to serve it to a user?
- +The user should not be editing source code to change image file.
- +Gradio as a solution!

#### What is Gradio?

- +Gradio is the fastest way to demo a machine learning model.
- +It provides a friendly web interface.
- +Gradio can be installed with pip.
- +Gradio can be embedded:
  - o in Python notebooks or
  - o presented as a webpage.
- +It can be permanently hosted on Hugging Face.

#### Some Gradio Examples



#### Let's create a simple UI for this model

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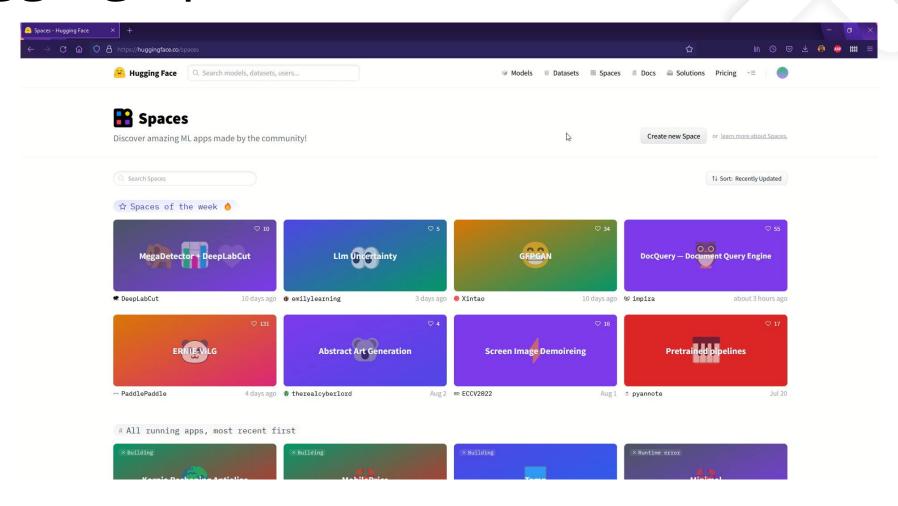


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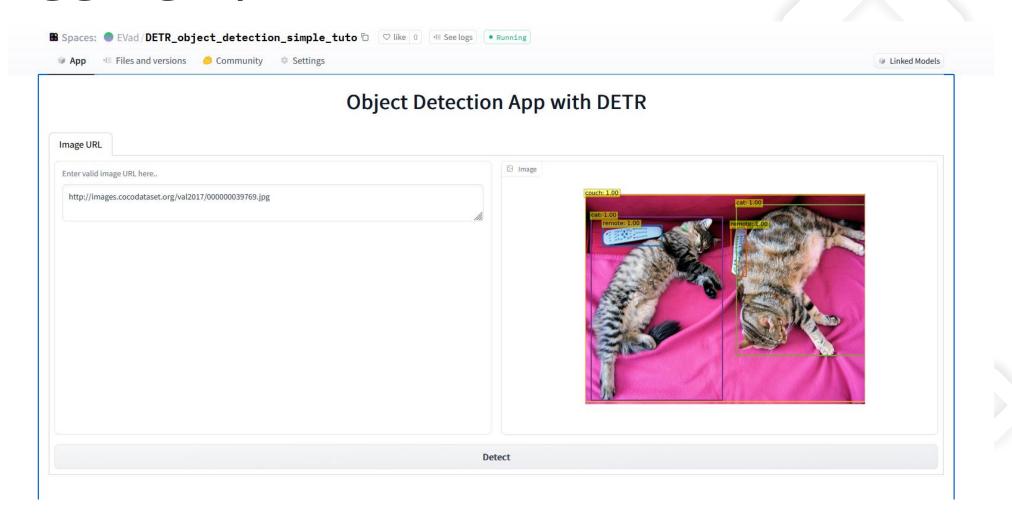
#### Let's create a simple UI for this model

```
▲ DETR object detection simple tuto.ipynb ☆
                                                                                                                                                                                                   Comment Share
      + Code + Text
                                                                                                                                                                                                               Editing
    - Creating the Gradio Interface
                                                                                                                                                                                                      イ ψ ⑤ 目 🌣 🎧 📋
      0
3 def make prediction(img, feature extractor, model):
            4 inputs = feature extractor(img, return tensors="pt")
                 outputs = model(**inputs)
                 img_size = torch.tensor([tuple(reversed(img.size))])
                 processed outputs = feature extractor.post process(outputs, img size)
                 return processed outputs[0]
           10 def detect objects(url input):
                 #Extract model and feature extractor
                 feature extractor = AutoFeatureExtractor.from pretrained("facebook/detr-resnet-50")
                 model = DetrForObjectDetection.from pretrained("facebook/detr-resnet-50")
                 # if image comes from URL
                 if validators.url(url input):
                     image = Image.open(requests.get(url_input, stream=True).raw)
                 processed outputs = make prediction(image, feature extractor, model)
                 viz_img = visualize_prediction(image, processed_outputs, 0.7, model.config.id2label)
                 return viz img
            2 COLORS = [
                 [0.000, 0.447, 0.741],
                 [0.850, 0.325, 0.098],
                 [0.929, 0.694, 0.125],
                 [0.494, 0.184, 0.556],
                 [0.466, 0.674, 0.188],
                 [0.301, 0.745, 0.933]
```

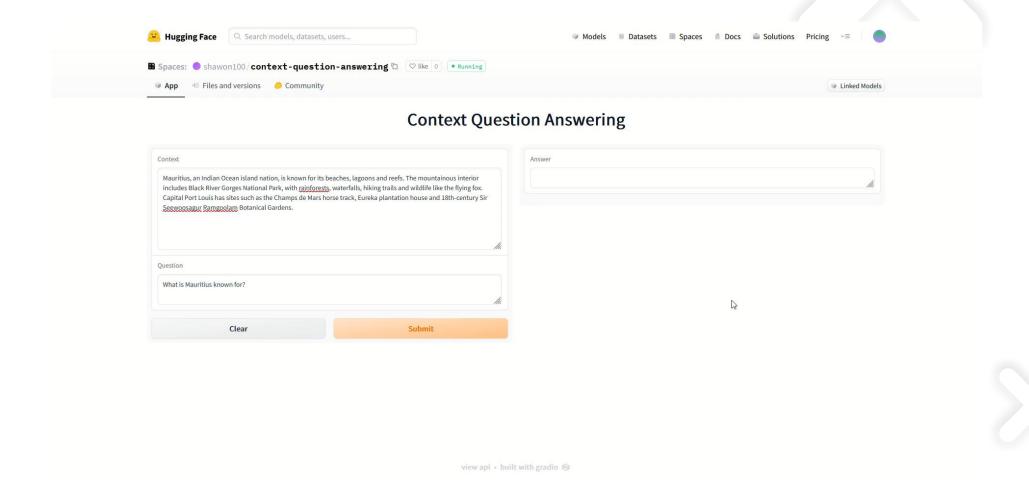
# Hosting our Object Detection app on Hugging Space



# Hosting our Object Detection app on Hugging Space



### What else does Spaces offer?



#### Resources

- +https://github.com/azezezaaa/DETR-object-detection-tutorial
- +https://huggingface.co
- +https://gradio.app/
- + <a href="https://huggingface.co/spaces/anaxagoras7/gauravgs-text-summarizer">https://huggingface.co/spaces/anaxagoras7/gauravgs-text-summarizer</a>
- +https://huggingface.co/spaces/Narrativaai/NLLB-Translator
- +https://huggingface.co/spaces/stabilityai/stable-diffusion
- +https://huggingface.co/spaces/dalle-mini/dalle-mini

#### End



Thank you!



Connect with me