

Serving Machine Learning Models in a Production Environment

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Serving Machine Learning Models in a Production Environment

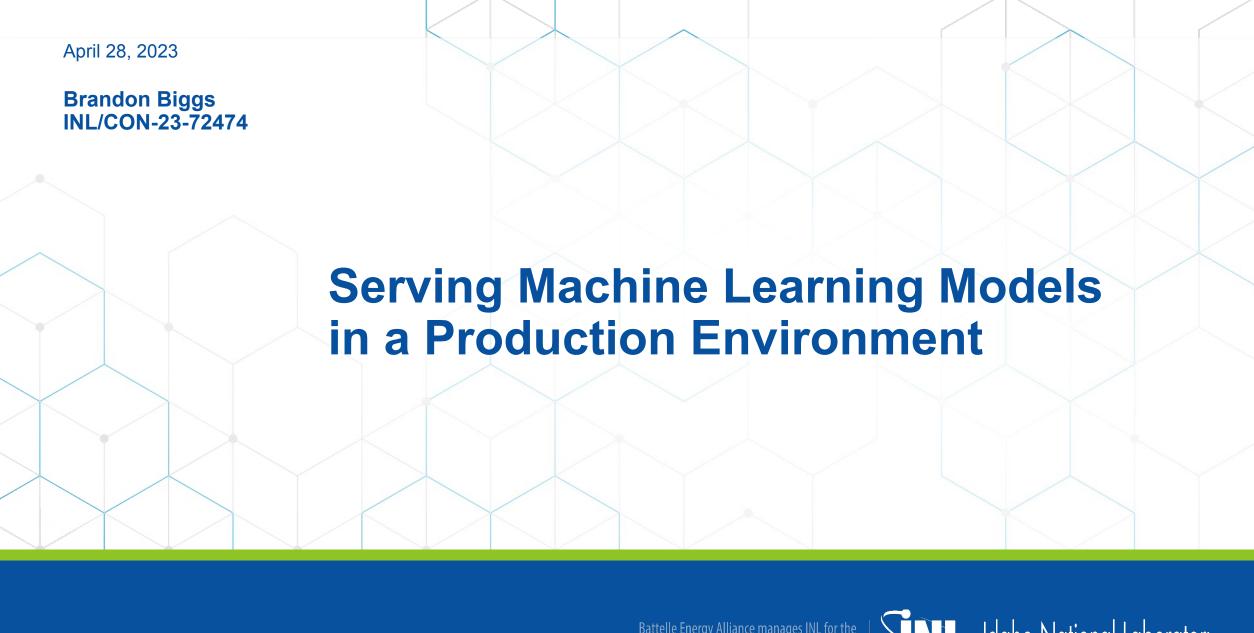
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Roadmap

- Motivation
- Challenges
- Goals
- Machine Learning Operations (MLOps)
- Model Repositories
- Hosting
 - Online vs Offline (Batch)
- Considerations
- Future Considerations



→ Hosted inference API ⑤

[™] Text-to-Image

A dog trying to fix a computer

Compute

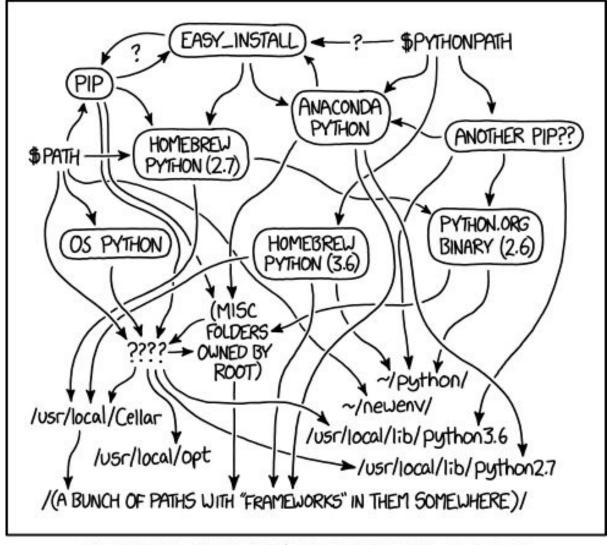
Computation time on gpu: 11.832 s

- Inference can be powerful (and fun!)
- Save time
- Collaborate
- Generate new ideas



Challenges

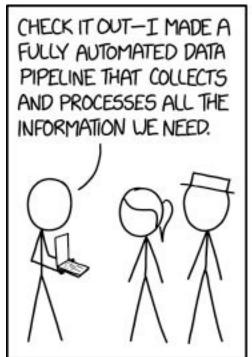
- Machine learning is rapidly advancing.
- Managing dependencies is hard.
 - Pip, pipx, conda, poetry, pipenv for Python
 - Containers?
 - Accelerator packages?
- Between 8 and 90 days for companies to deploy a single model [1]
- 2,473 organizations surveyed and found that a significant portion of their attempted AI deployments fail [1]
- Can we make machine learning useful?

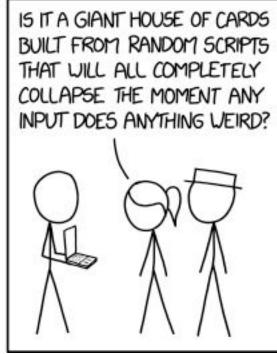


MY PYTHON ENVIRONMENT HAS BECOME SO DEGRADED THAT MY LAPTOP HAS BEEN DECLARED A SUPERFUND SITE.

Goals

- Make machine learning more accessible.
- Provide a way for people to utilize the efforts of data scientists, researchers, etc.



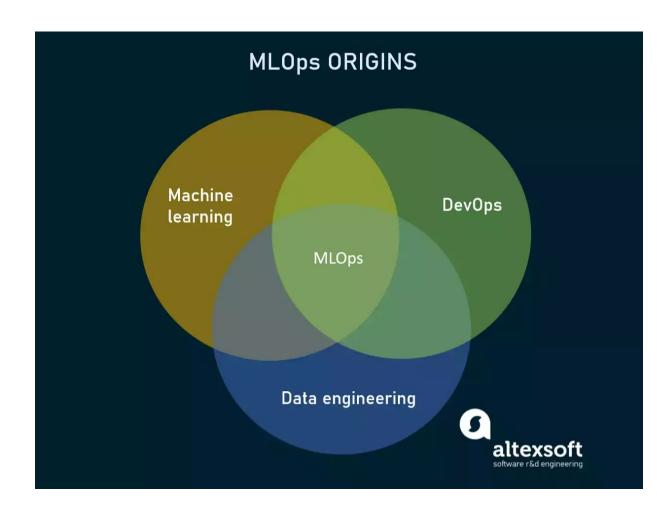






Machine Learning Operations - MLOps

- Like DevOps but for machine learning.
- Has a lot of the same practices but adapted for ML specific challenges.
 - Data
 - Model Training
 - Model Serving



Hosting

- Online and/or Offline
- Cloud vs On-Prem



Hosting Tools - Cloud

- AWS SageMaker
- Google Vertex Al
- Azure Machine Learning Endpoints
- Many more

Hosting Tools - Local

- NVIDIA Triton
- BentoML
- TensorFlow Serving
- Torch Serve
- Mlflow
- Kubeflow
- Building your own with an API framework!

```
from transformers import ViTImageProcessor, ViTForImageClassification
from PIL import Image
import requests
from fastapi import FastAPI
processor = ViTImageProcessor.from_pretrained('google/vit-base-patch16-224')
model = ViTForImageClassification.from_pretrained('google/vit-base-patch16-224')
app = FastAPI()
@app.post("/classify_via_url/")
async def classify_via_url(image_url):
    image = Image.open(requests.get(image_url).raw)
    inputs = processor(images=image, return_tensors="pt")
    outputs = model(**inputs).logits
    prediction_id = logits.argmax(-1).item()
    label = model.config.id2label[prediction_id]
    return label
```

Image Classification Projects description..

GET

/classify/{image-url:path} Image By Url

Text-to-Speech this is a description

GET

/text-to-speech/{text} Get Audio From Text

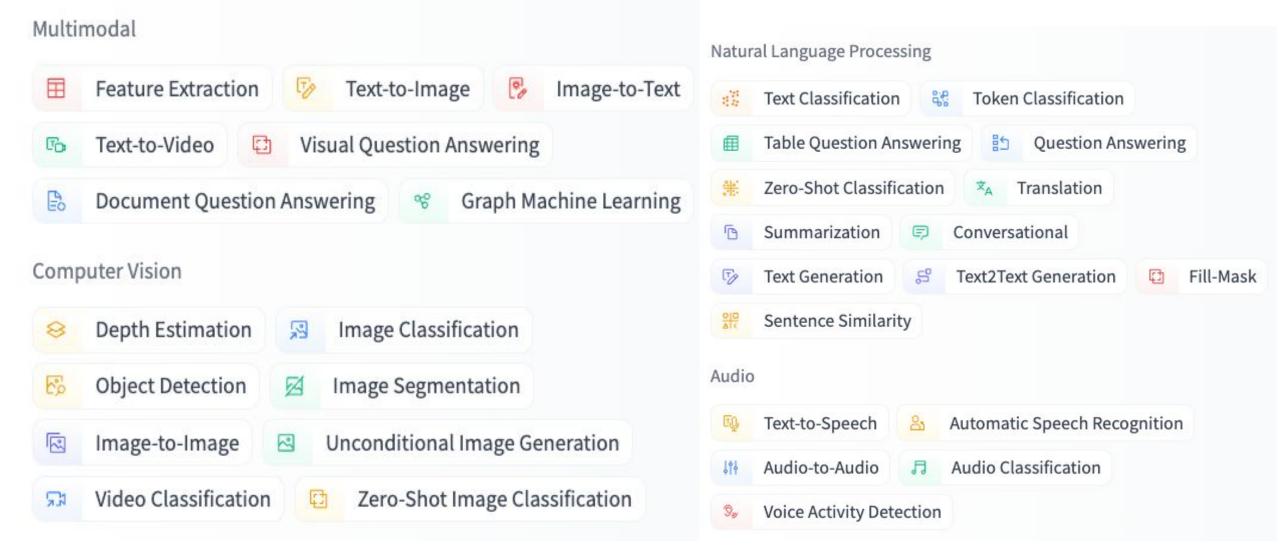
Image Generation Projects description..

GET

/stable-diffusion/text-to-image/{text} Image From Text

Model Repositories

- Hugging Face (cloud)
- BentoML (on prem)
- Weights and Biases (Paid)
- DagsHub
- MLRun
- More!



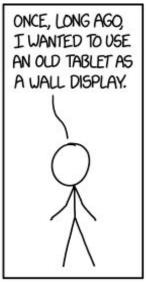
Considerations

- Containers
- Hardware
- Security
- Monitoring

Considerations - Containers

- Very useful for dependencies
- Lots of already created containers with dependencies
 - NVIDIA (many)
 - AMD Infinity Hub (few)
- Don't have to use Docker, Singularity/Apptainer have been well tested.









...THEN I REALIZED IT WOULD BE WAY EASIER TO GET TWO SMALLER PHONES ON EBAY AND GLUE THEM TOGETHER.



ON THAT DAY, I ACHIEVED SOFTWARE ENLIGHTENMENT.



Considerations - Hardware

- A lot of work in the NLP space going into optimizing large language models for CPUs/non-datacenter GPUs
- Some models just require a lot of GPU memory
- Tools exist for testing on GPUs before deploying models somewhere
 - Google Colab

Considerations – Security and Monitoring

- You're building an API for people to access. Watch out for abuse and unauthorized access.
- Keep an eye on models being used to generate toxic content

Future Considerations

- Data Versioning
- Experiment Tracking
- Model Registry
- API Scaling

Questions?

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