



COMMUNITY DAY



Avoiding Common Pitfalls with Hosting Machine Learning Models

Max De Jong | June 13, 2024



Who Am I?

Applied scientist with academic background

Realized that a “full ML stack” understanding required for maximum impact

Beware the data science pin
factory: The power of the full-
stack data science generalist and
the perils of division of labor
through function



ERIC COLSON

March 11, 2019 - San Francisco, CA

Explosion of Open Source Models

There has never been a better moment to build with machine learning



The image shows a screenshot of the Hugging Face Model Hub. The interface features a large, smiling emoji icon on the left. To its right, the text "Hugging Face" is displayed in a large, bold, dark blue font. Below this, there are several sections of model cards. The first section shows a card for "MIT/ast-finetuned-audio-set-10-10-0.4593" with a subtitle "Audio Classification" and a download count of 105M. The second section shows a card for "openai/clip-vit-large-patch14" with a subtitle "Zero-Shot Image Classification" and a download count of 62.7M. The third section shows a card for "jonatasgrosman/wav2vec2-large-xlsr-53-english" with a subtitle "Automatic Speech Recognition" and a download count of 53.5M. The fourth section shows a card for "google-bert/bert-base-uncased" with a subtitle "Fill-Mask" and a download count of 49.2M. The fifth section shows a card for "tohoku-nlp/bert-base-japanese" with a subtitle "Fill-Mask" and a download count of 37.6M. The sixth section shows a card for "sentence-transformers/all-MiniLM-L6-v2" with a subtitle "Sentence Similarity" and a download count of 31.6M. The seventh section shows a card for "CAMEL-Lab/bert-base-arabic-camelbert-mix-pos-egy" with a subtitle "Token Classification" and a download count of 35.4M. The eighth section shows a card for "mrm8488/distilroberta-finetuned-financial-news-sent..." with a subtitle "Text Classification" and a download count of 29.3M. The ninth section shows a card for "distilbert/distilbert-base-uncased" with a subtitle "Fill-Mask" and a download count of 16.4M. The tenth section shows a card for "FacebookAI/roberta-base" with a subtitle "Fill-Mask" and a download count of 16M. At the top of the page, there is a search bar with the text "Models 608,679" and a "Filter by name" button. To the right of the search bar are buttons for "new", "Full-text search", and "Sort: Most downloads".

Model	Description	Downloads	Stars
MIT/ast-finetuned-audio-set-10-10-0.4593	Audio Classification	105M	126
openai/clip-vit-large-patch14	Zero-Shot Image Classification	62.7M	1.1k
jonatasgrosman/wav2vec2-large-xlsr-53-english	Automatic Speech Recognition	53.5M	405
google-bert/bert-base-uncased	Fill-Mask	49.2M	1.49k
tohoku-nlp/bert-base-japanese	Fill-Mask	37.6M	27
CAMEL-Lab/bert-base-arabic-camelbert-mix-pos-egy	Token Classification	35.4M	2
mrm8488/distilroberta-finetuned-financial-news-sent...	Text Classification	29.3M	220
distilbert/distilbert-base-uncased	Fill-Mask	16.4M	408
FacebookAI/roberta-base	Fill-Mask	16M	327

Yet Something Is Missing...

Breakthroughs in models don't translate to ML democratization



Blog

Find something...

Log in

Sign up

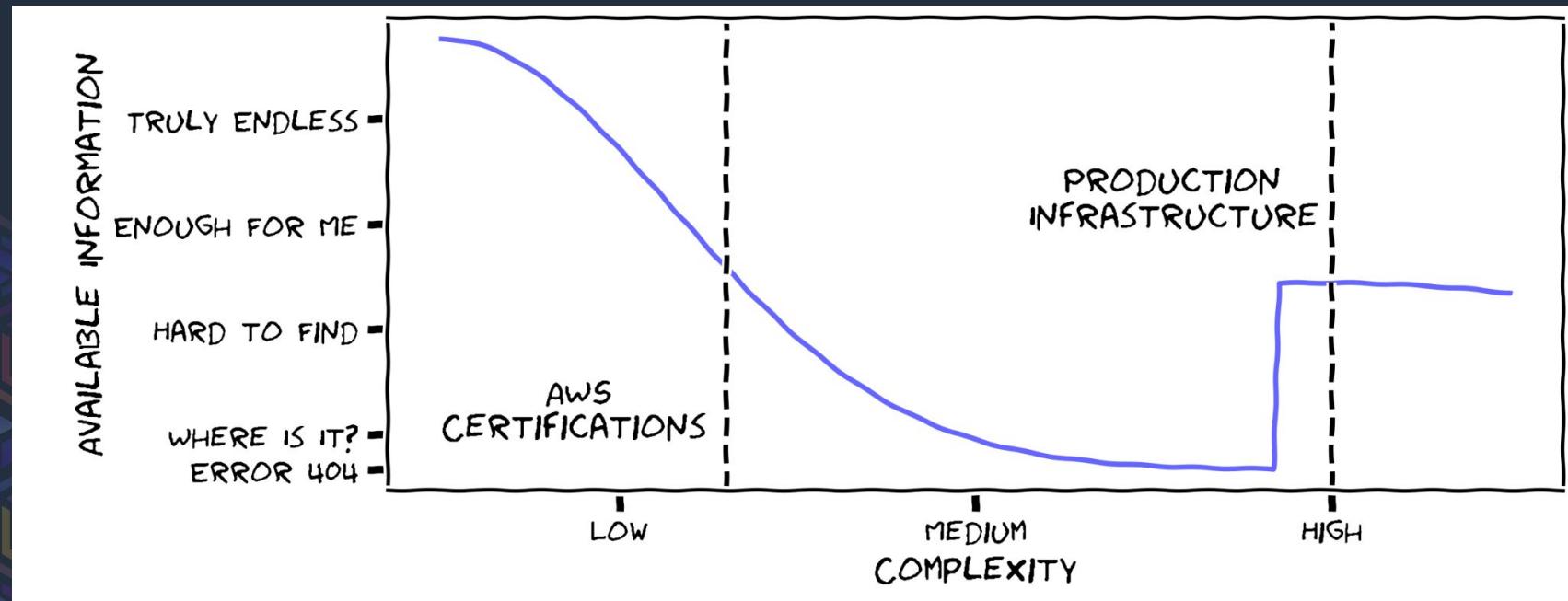
APRIL 17, 2024

If everyone is building AI, why aren't more projects in production?



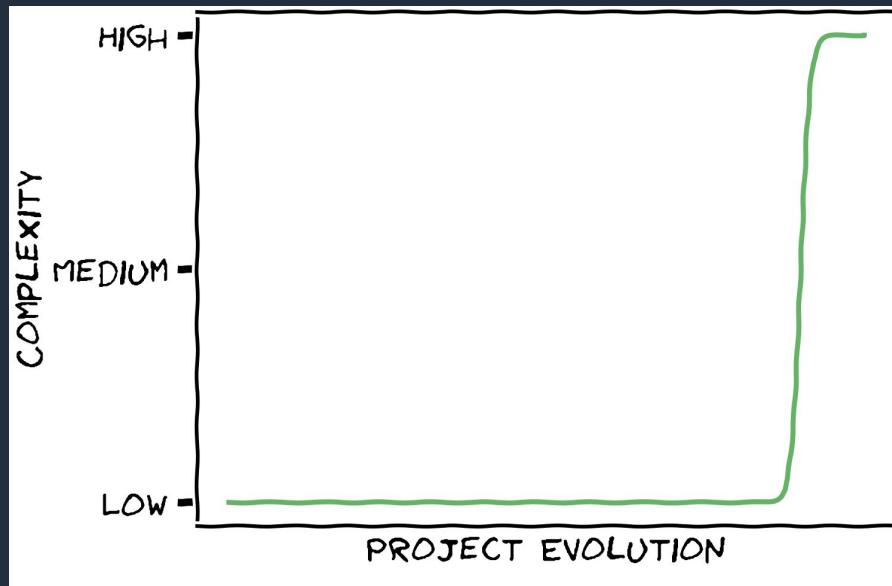
Major Knowledge Gap

Lack of intermediate resources makes learning much harder than necessary



Resulting Difficulty Cliff

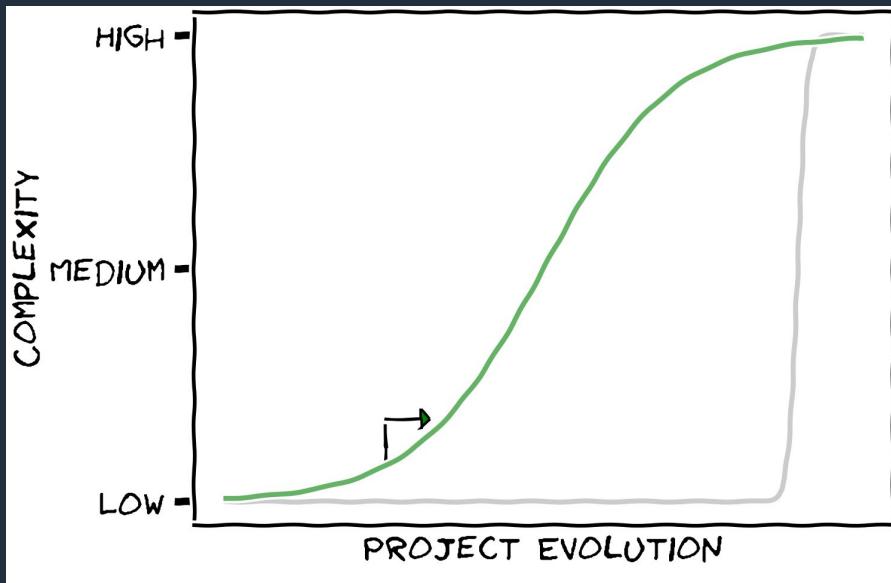
Very hard to transition out of the beginner phase of a project without enough educational resources



Flattening the Difficulty Cliff

Today's theme: finding atomic, tractable improvements to allow for meaningful iteration

Along the way, identifying pitfalls to avoid



Building Philosophy

Machine learning solutions are costly to properly build and maintain

Lots of models end up not working as intended

Goal: avoid sinking time in untested ideas while avoiding getting crushed by technical debt if we want to scale our solution

Think “scalable proof of concept”

It's ok to have “bad” system designs as long as they can be easily improved

Where to Build

Prototype locally, deploy to AWS



How to Build

1. Fact Finding

2. Bake-off

3. Microservice Translation

4. Cloud Essentials Migration

5. Full Cloud Migration

Machine Learning

Software Engineering

Solutions Architecture

Specific Application: 3D Pose Estimation

What is possible with open-source models?

Some benchmarks for general tasks

Nothing for our specific use case

**Studio grade motion capture,
anywhere, for only \$1,995**



Step #1: Fact Finding

Double-pronged investigation through literature and repos

Goals:

1. Learn something about the classes of models
2. Make a list of repos with public code/weights

Fact Finding

Two major classes of approaches: top-down vs. bottom-up

Lots of potential projects to try

Benchmark results are only a starting point

Benchmarks

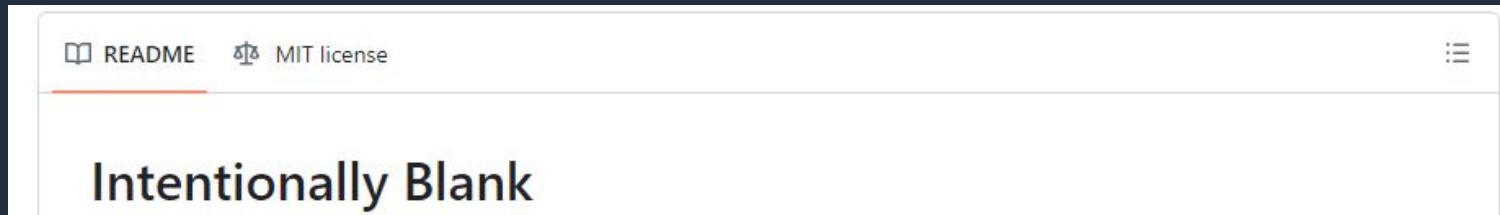
These leaderboards are used to track progress in Pose Estimation

Trend	Dataset	Best Model	Paper	Code	Compare
	MPII Human Pose	PCT (swin-l, test set)	Paper	Code	See all
	COCO test-dev	ViTPose (ViTAE-G, ensemble)	Paper	Code	See all
	Leeds Sports Poses	OmniPose	Paper	Code	See all
	OCHuman	ViTPose (ViTAE-G, GT bounding boxes)	Paper	Code	See all
	CrowdPose	BUCTD-W48 (w/cond. input from PETR, and generative sampling)	Paper	Code	See all
	MS COCO	OmniPose (WASPv2)	Paper	Code	See all
	AIC	Hulk(Finetune, ViT-L)	Paper	Code	See all

[Add a Result](#)

Step #2: Bake Off

“Simple”: clone the repos, read the READMEs, and run the example scripts



Two main obstacles:

1. Managing CUDA versions
2. Bit rot

Managing CUDA Versions: Wrong Way

Main obstacle: CUDA

Managing multiple CUDA versions using environment modules in Ubuntu

[Steps_multiple_cuda_environments.md](#) [Raw](#)

Steps to manage multiple CUDA environments

Latest Update: May 19th, 2024

This gist contains all the steps required to:

- Install multiple CUDA versions (e.g., `CUDA 11.8` and `CUDA 12.1`)
- Manage multiple CUDA environments on Ubuntu using the utility called [environment modules](#).
- Use this approach to avoid CUDA environment conflicts.



Solution: Docker

Every project gets a Dockerfile

Install a recent version of CUDA on your machine
Pin it until you have a good reason to upgrade

Every ML project gets its own container isolated from your system



Bit Rot

Many academic repositories are not maintained



Build Type	Linux	MacOS	Windows
Build Status	 CI failing	 CI failing	 build failing

[OpenPose](#) has represented the **first real-time multi-person system to jointly detect human body, hand, facial, and foot keypoints (in total 135 keypoints)** on single images.

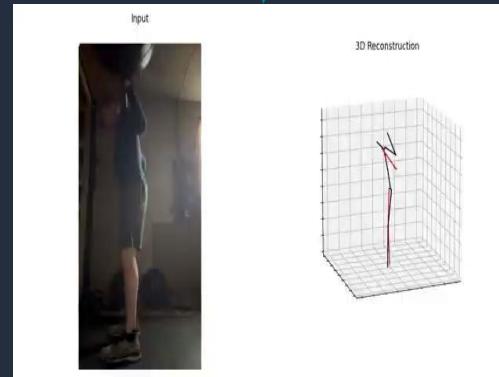
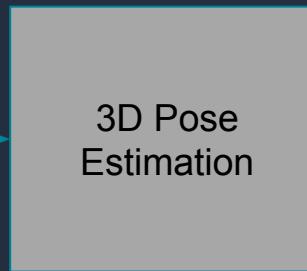
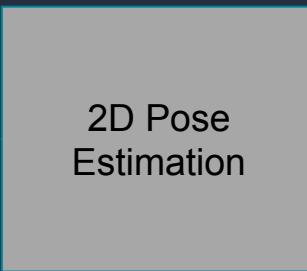
No Keypoints Detected After Running Python Examples #2280

 Open

YogeshNeoS opened this issue on Feb 5 · 0 comments

Finalizing Architecture

End goal: settle on the model pipeline



Step #3: Microservice Translation

General procedure:

1. Wrap model inference in APIs using Flask/fastAPI
2. Create a web server using gunicorn/uWSGI
3. Run NGINX reverse-proxy



Microservice Translation

Main obstacle: Monolith with tight coupling



Can someone use yolov8 instead of yolov3 in detector? #1212

Open

Ism101010 opened this issue 27 days ago · 1 comment



Ism101010 commented 27 days ago

...

Can someone use yolov8 instead of yolov3 in detector?



Madou-AI commented last week

...

i want to use the yolov9 but i don't know how

Docker Compose

Other containers supporting the ML microservices
Database, utilities, front end, etc.



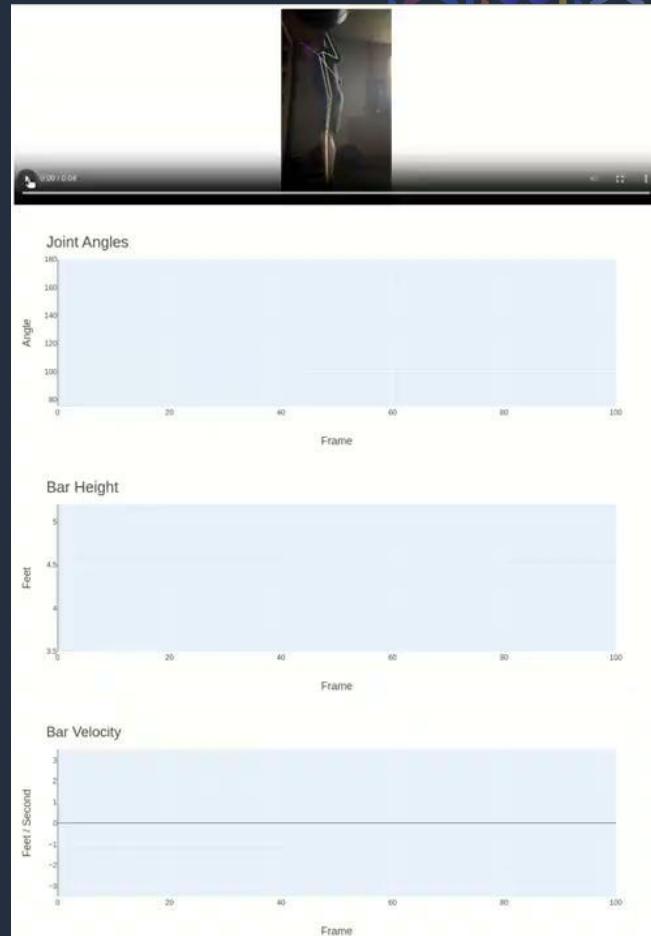
Docker Compose allows running multi-container applications

```
git docker-compose.yml
You, 9 months ago | 1 author (You)
1 services:
2   postgres:
3     build: ./database/
4     env_file: database.conf
5     volumes:
6       - /home/max/database_data:/var/lib/postgresql/data/form_judge_data
7     expose:
8       - "5432"
9   nginx:
10    build: ./nginx/
11    ports:
12      - "80:80"
13    depends_on:
14      - front-end
15   front-end:
16     build: ./front_end/
```

Microservices End State

Local containerized service running end-to-end

```
(base) max@max-desktop:~$ docker ps
CONTAINER ID   IMAGE          COMMAND
ec197702217c   form_judge-postgres   "docker-entrypoint.s...
77f53792e931   form_judge-nginx    "/docker-entrypoint...."
f3e2dda3c7a0   form_judge-front-end "gunicorn app:app -b...
faa939155c31   form_judge-detection "gunicorn app:app -b...
382433448e32   form_judge-pose-2d   "/tmp/post_init.sh"
ffb6a0ee7540   form_judge-pose-3d   "gunicorn app:app -b...
d5374d97ae13   form_judge-video-renderer "gunicorn app:app -b..."
```



Step 3/5: Microservices

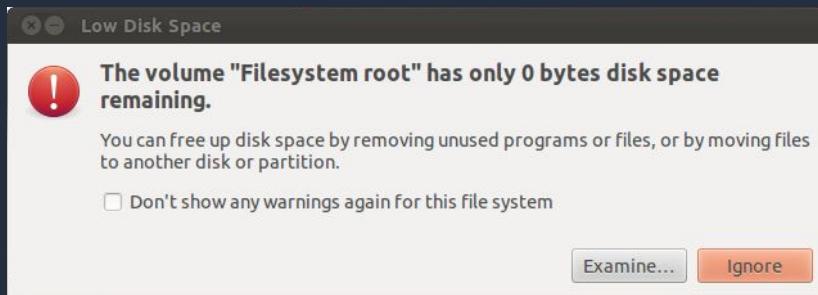
Step #4: Cloud Essentials

Still major design decisions before choosing a cloud architecture

Some common elements to all routes: database and object storage

These are the first things we don't want to manage

```
postgres-1  PostgreSQL Database directory appears to contain a database; Skipping initialization
postgres-1  2024-06-02 04:29:28.311 UTC [1] FATAL:  database files are incompatible with server
postgres-1  2024-06-02 04:29:28.311 UTC [1] DETAIL:  The data directory was initialized by PostgreSQL version 15, which is not compatible with this version 16.3 (Debian 16.3-1.pgdg120+1).
```



Database Choice

Hobby projects really benefit from a scale-to-zero database

Usage	Go to billing
Storage <small>?</small>	30.55 MiB
Compute time since May 31 <small>?</small>	0.3 h
Branches	1
Data transfer	11.88 KiB
<small>Metrics may be delayed by up to one hour. Read more about metrics.</small>	

Minimum monthly cost of Aurora serverless: \$43

Minimum monthly cost of Aurora on db.t4g.medium: \$53



Storage

S3 always a good starting point

Later migrate to something EFS or similar



Step #5: Full Cloud Microservice Deployments

Two routes here:

1. Run the microservices on Elastic Container Service
2. Run the microservices on Elastic Kubernetes Service

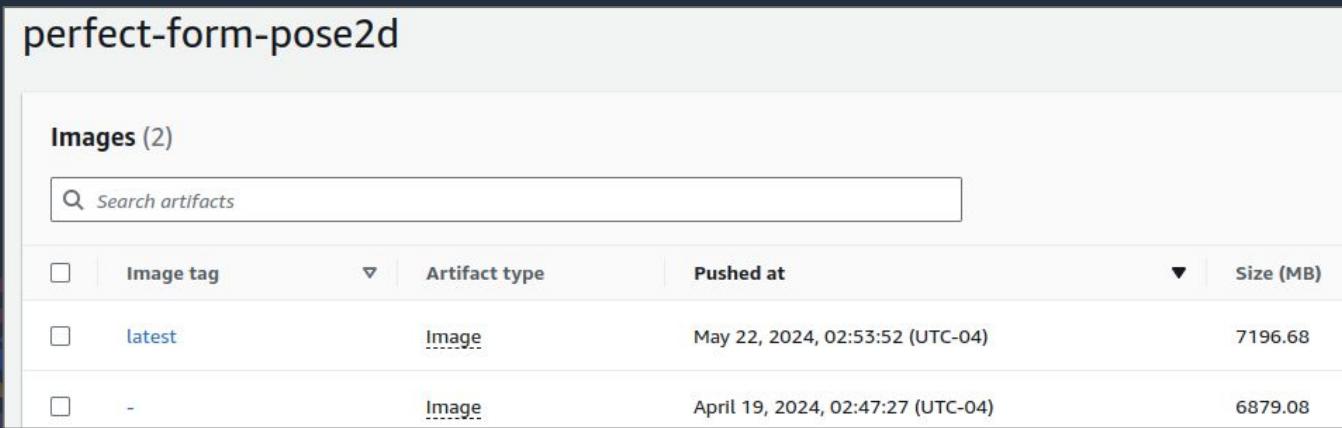
But first, let's get ready for running containers on AWS hardware



Elastic Container Registry

Use ECR to store Docker container images

Large models produce large images



Images (2)				
<input type="text"/> Search artifacts				
<input type="checkbox"/>	Image tag	Artifact type	Pushed at	Size (MB)
<input type="checkbox"/>	latest	Image	May 22, 2024, 02:53:52 (UTC-04)	7196.68
<input type="checkbox"/>	-	Image	April 19, 2024, 02:47:27 (UTC-04)	6879.08

Optimizing Dockerfiles for ECR



Pushing to ECR is slow: think about layers

```
app > 📄 Dockerfile > ...
1  # Use a base image with Ubuntu and cuda
2  FROM nvidia/cuda:11.5.2-base-ubuntu20.04
3
4  # Set the working directory in the container
5  WORKDIR /app
6
7  # Copy the Flask application files to the container
8  COPY app/ ./
9
10 # Copy the Flask application files to the container
11 COPY app/requirements.txt ./
12
13 # Install the required Python packages
14 RUN pip install -r requirements.txt
15
16 # Expose the Flask application port
17 EXPOSE 5000
18
19 CMD ["python3", "-m", "flask", "run", "--host=0.0.0.0", "--port=5000"]
```

```
app > 📄 Dockerfile > ...
1  # Use a base image with Ubuntu and cuda
2  FROM nvidia/cuda:11.5.2-base-ubuntu20.04
3
4  # Set the working directory in the container
5  WORKDIR /app
6
7  # Copy the Flask application files to the container
8  COPY app/src/ ./
9  COPY app/weights/ ./
10 COPY app/app.py ./
11
12 # Copy the Flask application files to the container
13 COPY app/requirements.txt ./
14
15 # Install the required Python packages
16 RUN pip install -r requirements.txt
17
18 # Expose the Flask application port
19 EXPOSE 5000
20
21 CMD ["python3", "-m", "flask", "run", "--host=0.0.0.0", "--port=5000"]
```

Precursor: EC2 with Docker

Spin up an EC2 instance with a GPU (p3.2xlarge)



Recreate your local Docker Compose app pulling from ECR

CPU Dockerized applications behave better than GPU applications

If we have to debug something, let's do it in easy mode

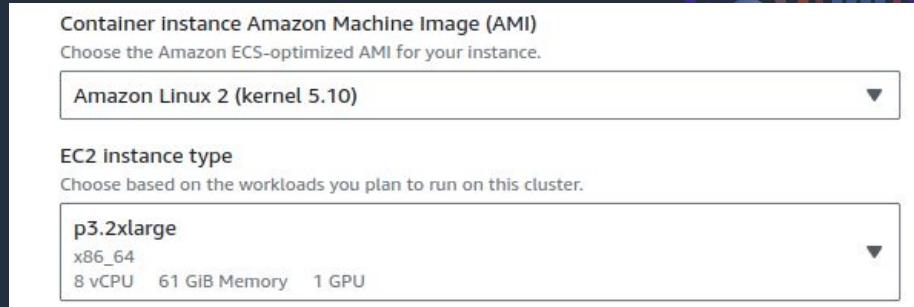
```
ValueError: Unknown CUDA arch (8.6) or GPU not supported
```

```
from judge-pose-3d-1           | RuntimeError: The NVIDIA driver on your system is too old (found version 11060). Please
update your GPU driver by downloading and installing a new version from the URL: http://www.nvidia.com/Download/index
.aspx Alternatively, go to: https://pytorch.org to install a PyTorch version that has been compiled with your version
of the CUDA driver.
```

Choosing Cloud Direction

Elastic Container Service
Less complexity
Scale to zero

Elastic Kubernetes Service
Best practice for full control
Easier local development
Multi-cloud solution



Container instance Amazon Machine Image (AMI)
Choose the Amazon ECS-optimized AMI for your instance.

Amazon Linux 2 (kernel 5.10)

EC2 instance type
Choose based on the workloads you plan to run on this cluster.

p3.2xlarge
x86_64
8 vCPU 61 GiB Memory 1 GPU

Managing EKS Clusters at Scale using Blueprints and Infrastructure as Code

Track 3 – 2024

10:00 am - 10:30 am

Presenter: Julia Furst Morgado

Elastic Container Service

True container orchestration: scalable

Translate our Docker Compose YAML
to a Task Definition

▼ Infrastructure requirements

Specify the infrastructure requirements for the task definition.

Launch type [Info](#)

Selection of the launch type will change task definition parameters.

AWS Fargate

Serverless compute for containers.

Amazon EC2 Instances

Self-managed infrastructure using Amazon EC2 instances.

Port mappings [Info](#)

Add port mappings to allow the container to access ports on the host to send or receive traffic. For port name, a default will be assigned if left blank.

Container port

Protocol

Port name

App protocol

5000

TCP

5000

HTTP

[Remove](#)

To start, group all GPU containers into a single Task

Working with ECS

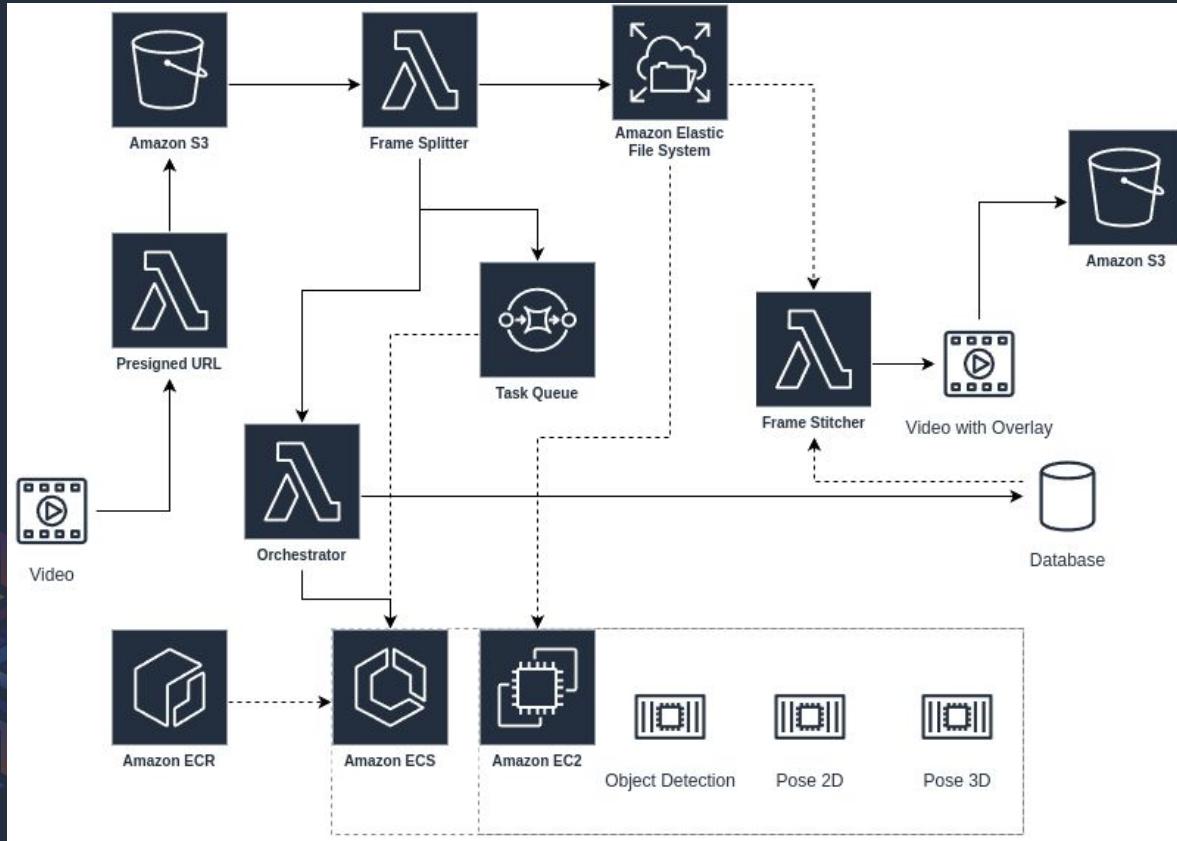
Pay attention to the Network Mode in Task Definitions

Can scale to 0 with some effort

Expose endpoints using Service Discovery/Service Connect

Don't be afraid to re-architect system to better utilize AWS services
API Gateway has 30 second timeout

Final Architecture



Recap

We started with a problem we wanted to solve

1. Found many potentially relevant repos with model weights
2. Determined the best model(s) for our use case
3. Created local microservice with Docker Compose
4. Moved storage to cloud
5. Migrated full microservice to AWS

