Lecture 9: Advanced Training Workflows

AI-5

Productionizing AI (MLOps)

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- 1. Recap
- 2. Experiment Tracking
- 3. Tutorial: Experiment Tracking
- 4. Vertex AI, Serverless Training
- 5. Tutorial: Serverless Training
- 6. Multi GPU Training

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Recap: Motivation

The 3 components for better Deep Learning



- Extraction
- Transformation
- Labeling
- Versioning
- Storage
- Processing
- Input to Training

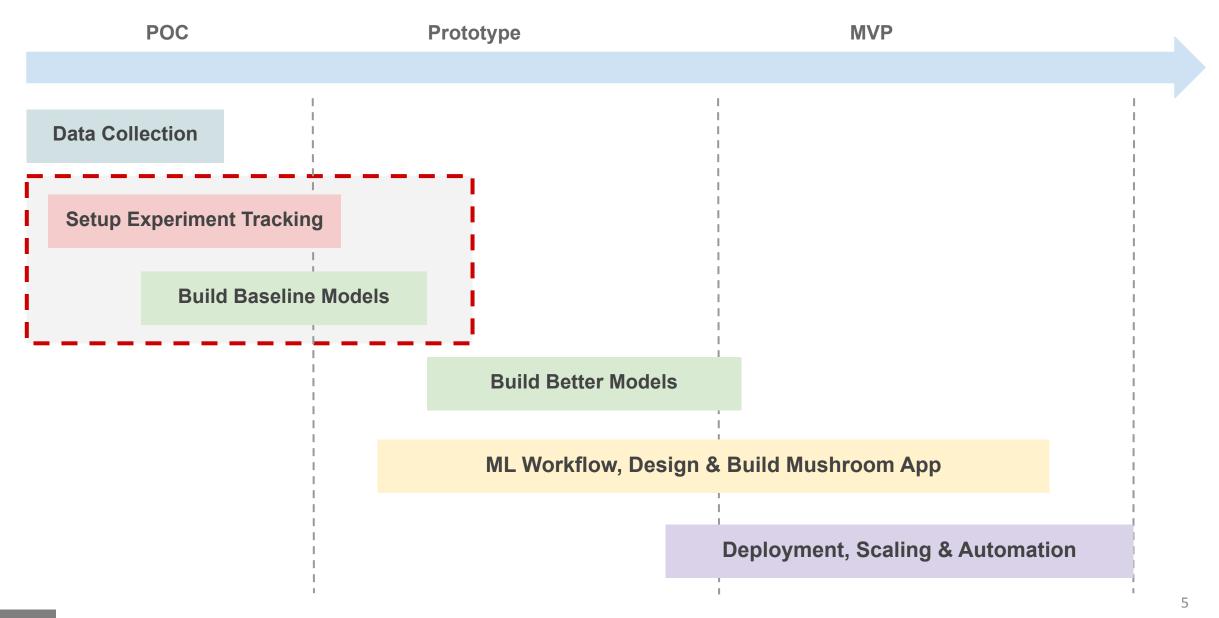


- SOTA Models
- Transfer Learning
- Distillation
- Compression



- Scaling data processing
- GPU, TPU
- Multi GPU Server Training

Recap: Project Workflow



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Experiment Tracking

Why

- Organize your work (data collection/model training)
- Reproducibility
- Logging

Experiment Tracking

What

- Environments
- Scripts (code)
- Data (version, train/validate/test split)
- Data pre-processing logic
- Model hyper parameters / configurations
- Evaluation metrics
- Model weights
- Performance results
- Sample predictions

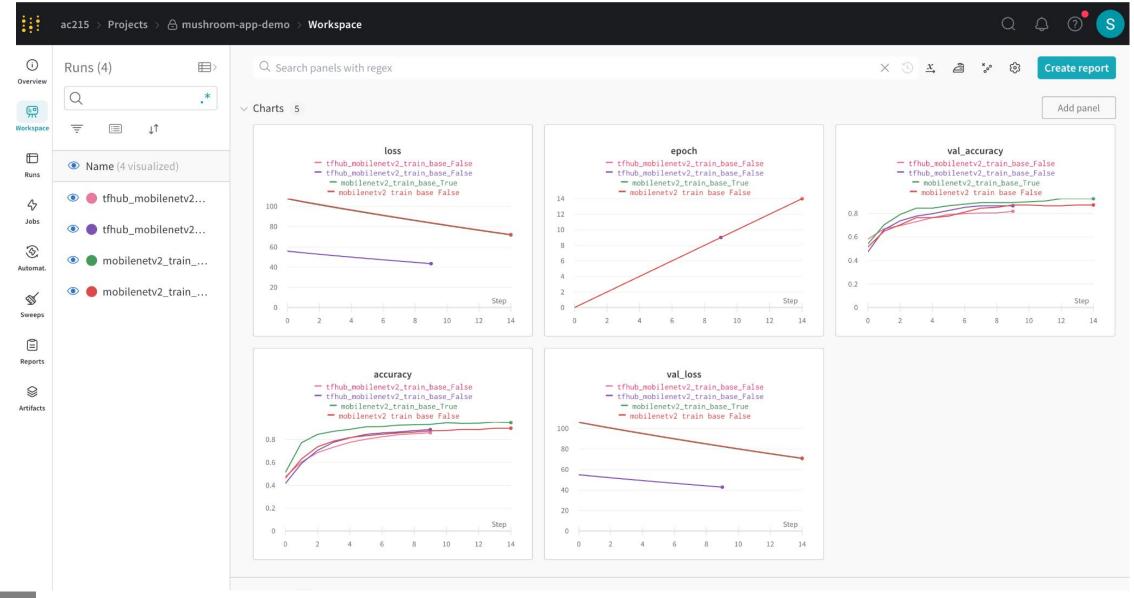
Training Code

```
# Training Params
learning_rate = 0.001
batch_size = 32
epochs = 10
# Data
train_data, validation_data = get_dataset(...)
# Model
model = build_model(...)
# Train
training_results = model.fit(...)
```

Training Code using WandB

```
# Training Params
learning rate = 0.001
batch size = 32
                                                        Initialize wandb run
epochs = 10
# Data
train_data, validation_data = get_dataset(...)
# Model
model = build model(...)
                                                    Add a callback to monitor model
# Initialize a W&B run
wandb.init(...)
# Train
training results = model.fit(...,callbacks=[WandbCallback()])
# Close the W&B run
                                               Let wandb know to finish the run
wandb.run.finish()
```

Experiment Tracking using WandB



Tutorial: Experiment Tracking

Goals of the tutorial are

- Explore models for mushroom classifications
- https://colab.research.google.com/drive/1H7KuMWpzSm8f_a_8 sA1JAylTKISTEqpz?usp=sharinq
- Experiment Tracking using Weights & Bias
 https://colab.research.google.com/drive/1tY5CcNP24kSnaac40
 pvcc2CPBoz0G6GF?usp=sharing

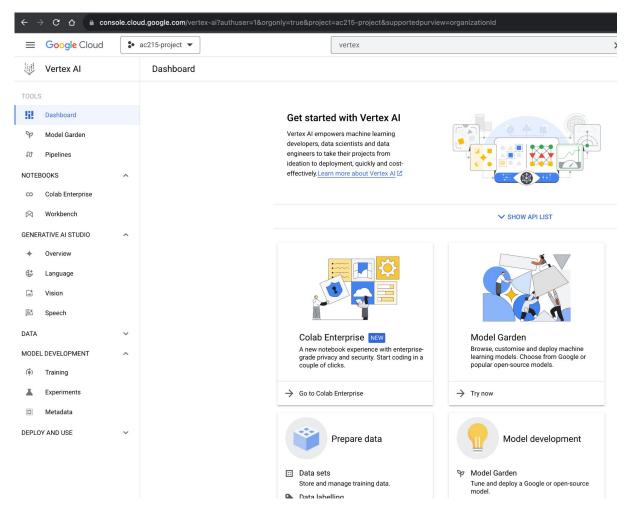
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Vertex AI

- Vertex AI is machine learning platform offered by Google in GCP.
- Vertex AI combines data engineering, data science, and ML engineering workflows, enabling your teams to collaborate using a common toolset and scale your applications using the benefits of Google Cloud.

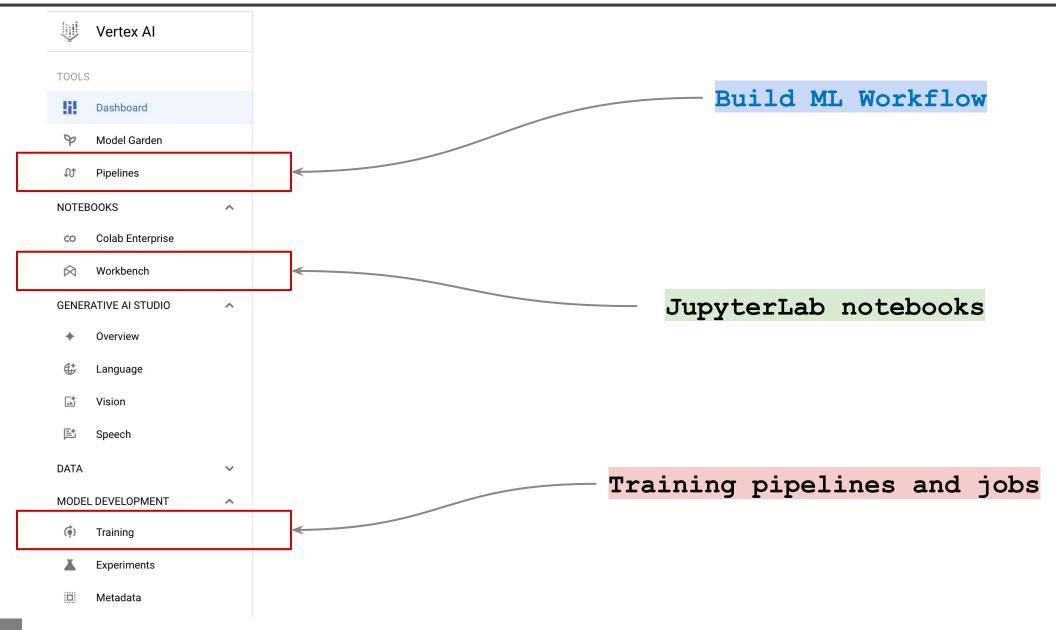
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Vertex AI: https://console.cloud.google.com/vertex-ai



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Vertex AI



Serverless Training

What is serverless training?

- Execute training on an as-need basis
- Access GPU hardware only for the "training" step in a pipeline
- No setup of servers required
- Brings down training cost

Serverless Training

Move Code to Python File

CO Notebook

```
def get dataset():
def get model 1():
def get model 2():
# Data
train_data, val_data = get_dataset(...)
# Model
model_1 = build_model_1(...)
# Train
training results = model 1.fit(...)
```

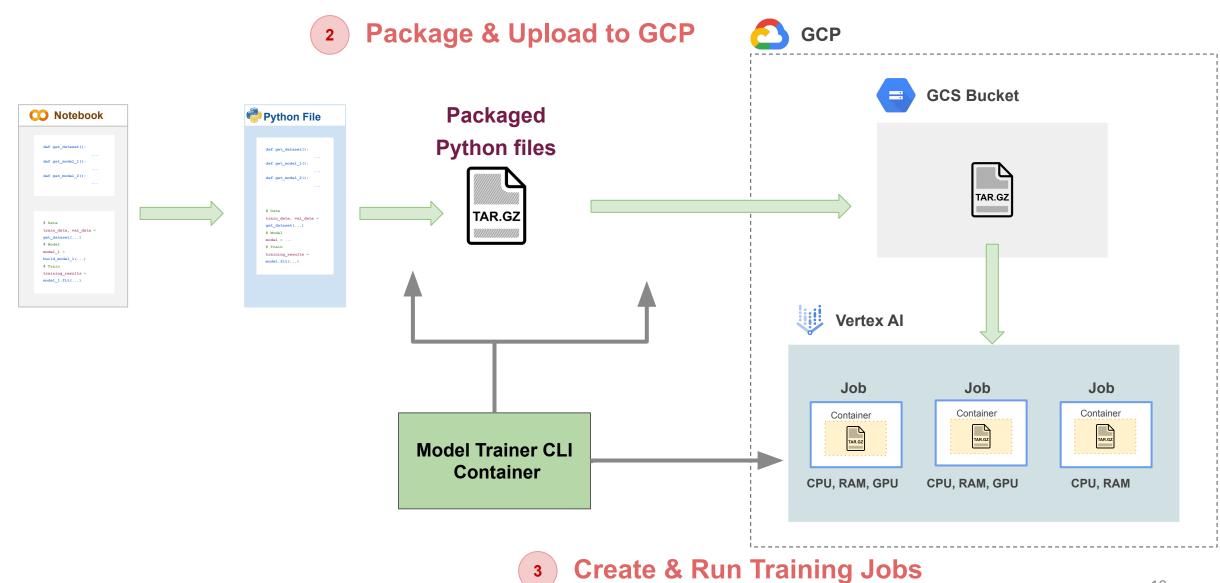


Python File

```
def get dataset():
def get model 1():
def get model 2():
# Data
train_data, val_data = get_dataset(...)
# Model
model = \dots
# Train
training results = model.fit(...)
```

Pass Arguments:

Serverless Training



Tutorial: Serverless Training

Steps to perform Serverless Training on mushroom classification models:

- Create a GCS bucket to store packaged python training code.
- Get Weights & Bias API Key for experiment tracking.
- Package & Upload python code.
- Create Jobs in Vertex AI.
- For detailed instructions, please refer to the following link
 - Serverless Training. (https://github.com/dlops-io/model-training)
 - View Training Jobs. (https://console.cloud.google.com/vertex-ai/training/custom-jobs)
 - View Experiment Metrics. (https://wandb.ai/home)

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Multi GPU Training

How do we perform distributed training?

- What type of distribution:
 - Single Machine, Single GPU [One Device Strategy]
 - Single Machine, Multiple GPUs [Mirrored Strategy]
 - Multiple Machine, Multiple GPUs [Multi Worker Mirrored Strategy]
- Organize code to apply the appropriate Strategy
- Train as usual

Training Code

```
# Training Params
learning_rate = 0.001
batch_size = 32
epochs = 10
# Data
train_data, validation_data = get_dataset(...)
# Model
model = build_model(...)
# Train
training_results = model.fit(...)
```

Training Code for Multi GPU

```
# Training Params
                                                       Create distribution Strategy
batch size = 32
. . .
# Create distribution strategy
strategy = tf.distribute.MirroredStrategy()
                                                   Adjust dataset batch size
num workers = strategy.num replicas in sync
# Data
train_data, validation_data = get_dataset(...,batch_size=batch_size * num_workers)
# Wrap model creation and compilation within scope of strategy
with strategy.scope():
   # Model
   model = build model(...)
                                        Create & Compile model in strategy scope
# Train
training results = model.fit(...)
```

