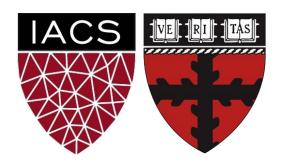
Lecture 11&12: ML Workflow Management



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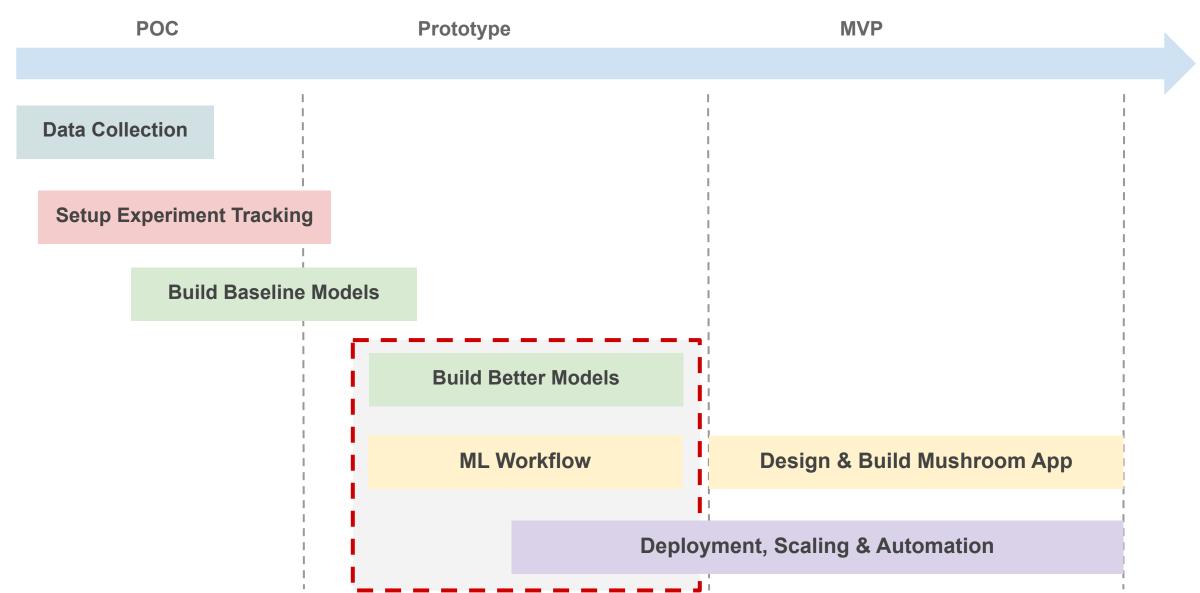


- 1. Recap
- 2. Serverless: Cloud Functions
- 3. Serverless: Cloud Run
- 4. Serverless: Model Deployment
- 5. ML Workflow Management
- 6. Vertex Al Pipelines

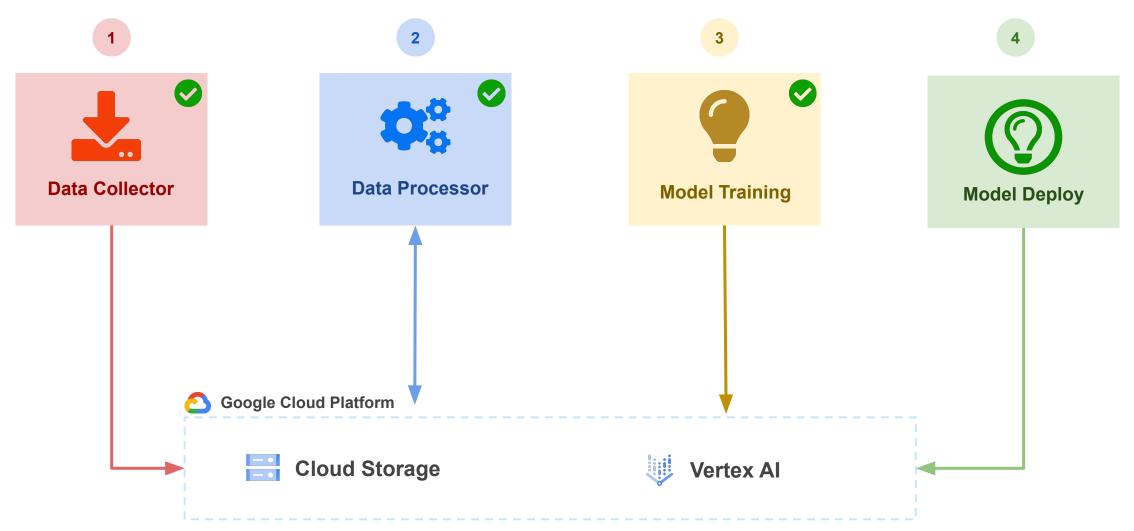
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Recap: Mushroom App Status



Mushroom App Development



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Serverless

What is serverless?

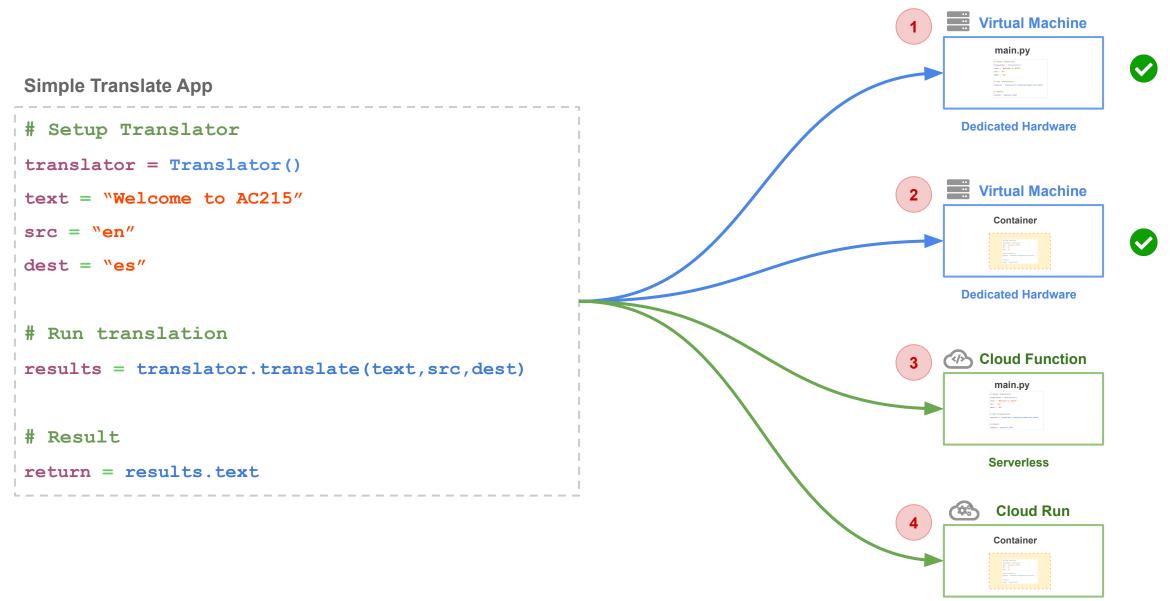
- Execute code on an as-need basis
- No setup of servers required
- Access GPU hardware only for the "training" step in a pipeline
- Brings down code execution cost

Serverless

Types of serverless:

- Cloud Function
- Cloud Run
- Training Job (Vertex AI)
- Model Deployment (Vertex AI)
- Pipeline (Vertex AI)

Deployment Options



Serverless

Cloud Function

What is a could function?

- Run your code in GCP with no servers or containers.
- Pay only for function execution time.
- Scale out easily

Tutorial: Cloud Function

Steps to deploy an app as a Cloud Function

- Go to https://console.cloud.google.com/functions.
- o Enable GCP APIs.
- Create a python code file.
- Deploy code as Cloud Function.
- For detailed instructions, please refer to the following link
 - Running App as Cloud Function. (https://github.com/dlops-io/serverless-deployment#running-app-as-cloud-function)

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Cloud Run

What is cloud run?

- Run your containerized apps with no servers.
- Run containers as service or job.
- Only pay when your code is running
- Scale out easily

Tutorial: Cloud Run

Steps to deploy an app in Cloud Run

- Go to https://console.cloud.google.com/run.
- Enable GCP APIs.
- Deploy Docker Image in Cloud Run.
- For detailed instructions, please refer to the following link
 - Running App in Cloud Run. (https://github.com/dlops-io/serverless-deployment#running-app-in-cloud-run)

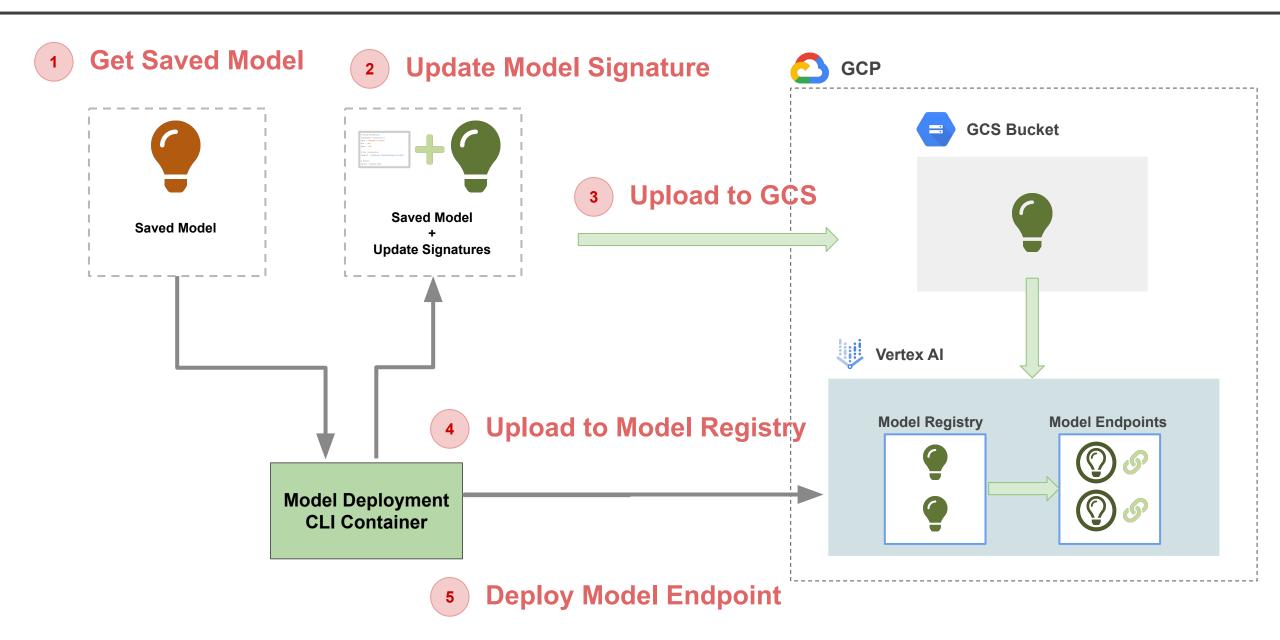
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Serverless Model Deployment

What is serverless model deployment?

- Deploy our trained model for predictions with no servers.
- Setup online or batch prediction modes
- For online predictions there is an ongoing cost
- Access GPU or CPU hardware for inference
- Scale out easily
- Alert: Continuous cost to keep endpoint up

Serverless Model Deployment



Serverless Model Deployment: Model Signature

Why do we need to update the model signature?

- Make model input to accept a raw image
- Perform data preprocessing steps prior to model inference
- Combine data preprocessing & model inference in one endpoint

Serverless Model Deployment: Update Model Signature

```
Preprocess Image
                                                                 Define preprocessing function
def preprocess image(bytes input):
   resized = ...
   return resized
# Define tf functions
@tf.function(input signature=[tf.TensorSpec([None], tf.string)])
                                                                            Define @tf.function for new model
def preprocess function(bytes inputs):
   decoded images = tf.map fn(
                                                                            signature
      preprocess image, bytes inputs, dtype=tf.float32, back prop=False
   return {"model input": decoded images}
@tf.function(input signature=[tf.TensorSpec([None], tf.string)])
def serving function(bytes inputs):
   images = preprocess function(bytes inputs)
   results = model call(**images)
   return results
                                                            Save Model with the new model signature
# Update model signature and save
tf.saved model.save(
   prediction model,...,
   signatures={"serving default": serving function},
                                                                                                                               19
```

Tutorial: Serverless Model Deployment

Steps to perform Serverless Model Deployment on mushroom classification model:

- Create a GCS bucket to store saved model.
- Update Model Serving Signature
- Upload Model to Vertex Al Model Registry.
- Deploy Model as an Endpoint.
- For detailed instructions, please refer to the following link
 - Serverless Model Deployment. (https://github.com/dlops-io/model-deployment)
 - View Model Endpoints. (https://console.cloud.google.com/vertex-ai/online-prediction/endpoints)
 - View Model Registry. (https://console.cloud.google.com/vertex-ai/models)

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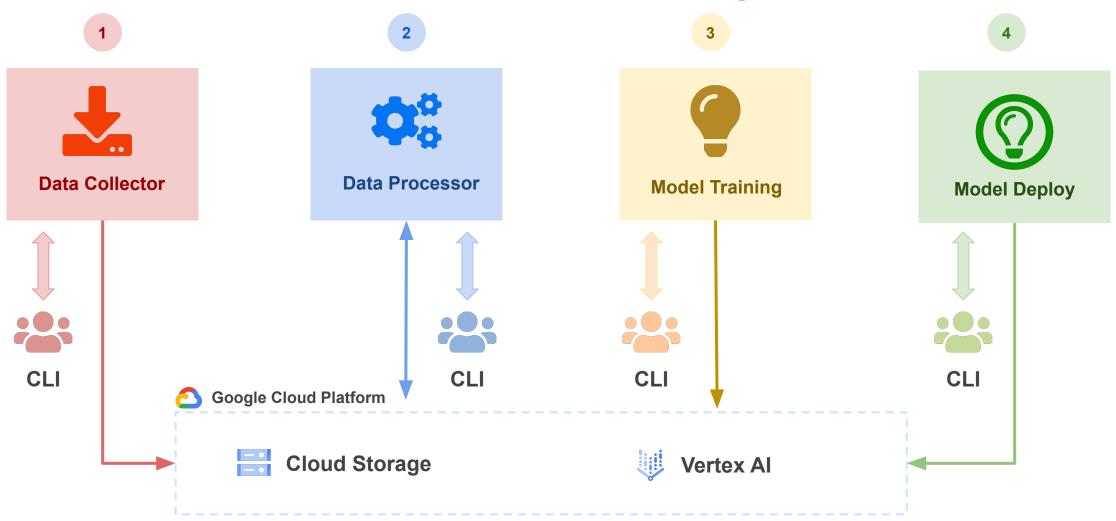
ML Workflow Management

What is ML workflow management?

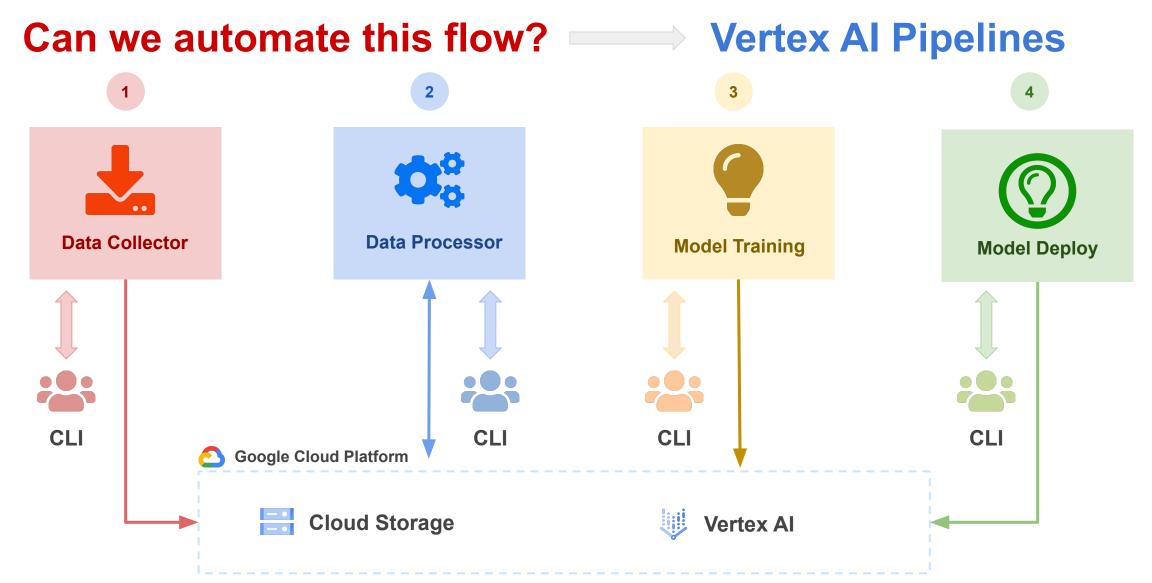
- Helps us efficiently manage end-to-end ML tasks from data collection to model deployment
- Helps orchestrate various and automated pipeline execution
- Manages collaboration, integration, and scalability

ML Workflow: Mushroom App

How do we execute these steps?



ML Workflow: Mushroom App



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Vertex Al Pipelines

What is Vertex Al Pipelines?

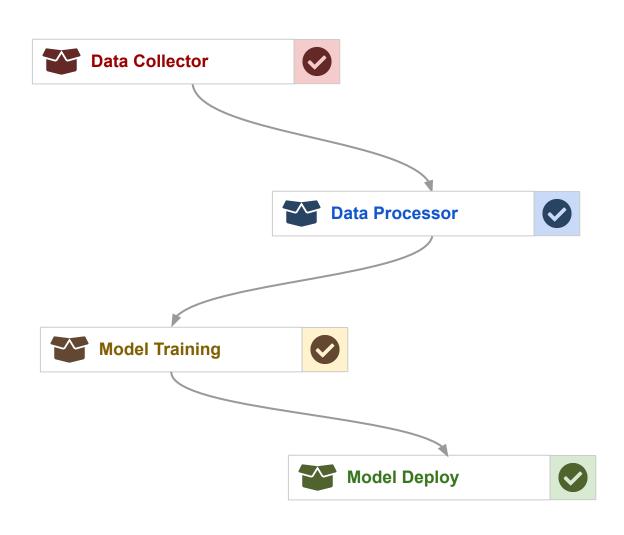
- Vertex AI is machine learning platform offered by Google in GCP.
- Vertex Al Pipelines helps you to automate, monitor, and govern your ML components by orchestrating your ML workflow in a serverless manner

Building Vertex Al Pipelines

Steps to build pipelines?

- Make your containers callable
- Build & Push Container Images to a Container Registry
- Define a sequence of steps using a directed acyclic graph (DAG)

Building Vertex Al Pipelines



- 1. Download images
- 2. Uploads to GCP

- 1. Verify images
- 2. Check for duplicates
- 3. Convert to TF Records

- 1. Train model
- 2. Save model

- 1. Upload to Registry
- 2. Deploy model Endpoint

Making Container Callable

Dockerfile

```
# Use the official Debian-hosted ...
FROM python: 3.9-slim-buster

.

# Add the rest of the source code.
RUN --chown=app:app . /app
# Entry point
ENTRYPOINT ["pipenv", "shell"]
```

Dockerfile

```
# Use the official Debian-hosted ...
FROM python: 3.9-slim-buster
# Add the rest of the source code.
RUN --chown=app:app . /app
# Entry point
ENTRYPOINT ["/bin/bash","./docker-entrypoint.sh"]
```

Change entrypoint to a shell file

Making Container Callable

```
Development mode:
docker-entrypoint.sh
                                           Authenticated to GCP
                                           pipenv shell to test cli inside container
#!/bin/bash
args = "$@"
if [[ -z ${args} ]];
then
   # Authenticate gcloud using service account
   gcloud auth activate-service-account --key-file $GOOGLE APPLICATION CREDENTIALS
   # Set GCP Project Details
   gcloud config set project $GCP PROJECT
   pipenv shell
else
  fi
                   Production mode:
                   Run container using "docker run ... cli.py -search"
```

Tutorial: Vertex Al Pipelines

Steps to build **Vertex Al Pipelines** on the mushroom app ML workflow components:

- Make Containers Callable.
- Build & Push Image.
- Build ML Pipeline.
- Run Pipeline in Vertex Al
- For detailed instructions, please refer to the following link
 - Mushroom App Workflows. (https://github.com/dlops-io/ml-workflow#mushroom-app-ml-workflow-management)
 - View Vertex Al Pipelines. (https://console.cloud.google.com/vertex-ai/pipelines/runs)

THANK YOU