Lecture 21: Operations - Automation - Review



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Outline

- 1. Motivation
- 2. Automation

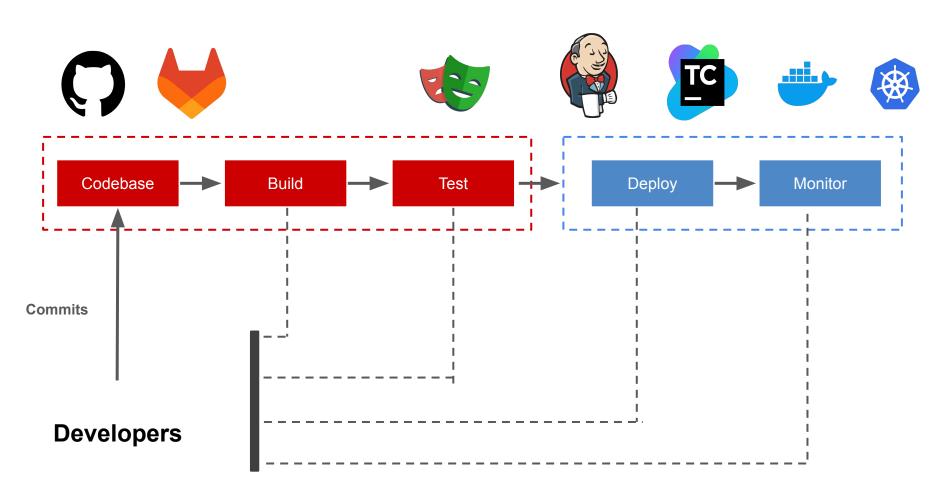
Outline

1. Motivation

2. Automation

We want to automate the following operations:

- Running data pipeline jobs
- Model training & model deployment
- Frontend deployment
- Backend deployment
- Building & Pushing Docker containers



Notifications

Continuous Integration (CI) automates a series of scripts to run whenever changes are pushed, in order to:

- Continuously integrate changes into the production branch
- Run automated tests
- Enforce coding standards
- Perform static code analysis

Note: Coding standards are about style and consistency.

Static code analysis is about identifying potential defects and vulnerabilities.

Continuous Integration (CI) allows multiple developers to contribute to a shared repository by **automating** quality checks to ensure the code remains functional.

This is achieved through:

- Running automated tests to catch bugs early
- Enforcing coding standards to maintain consistency
- Performing static code analysis

Note: Coding standards are about style and consistency.

Static code analysis is about identifying potential defects and vulnerabilities without executing the code.

Continuous Deployment (CD): Is to take automation further by **deploying** code changes to production automatically, as soon as the new features are **integrated** into the main codebase.

Continuous Delivery (CD): extension of CI to ensure software can be reliably released at any time.

Note: Delivery ensures changes are always ready to be deployed, but keeping a person in the loop. Deployment ensures changes are automatically passed into production.

CI in AI/ML

Regularly integrates **ML code**, **data**, and **models** into a shared repository while ensuring consistency through automated checks.

Automatically tests these integrations to ensure consistency.

Enables quick detection of issues in:

- Data quality
- Code correctness
- Model performance

Challenges in AI/ML CI

Balancing **frequent integrations** with:

- High computational demands of training models.
- Long-running tests for large datasets and complex pipelines.

Ensuring compatibility between evolving data, features, and model versions.

CI in AI/ML

Regularly integrates **ML code**, **data**, and **models** into a shared repository while ensuring consistency through automated checks.

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Challenges in AI/ML CI

Balancing frequent integrations with:

- High computational demands of training models.
- Long-running tests for large datasets and complex pipelines.

Ensuring compatibility between evolving data, features, and model versions.

Automates the release of **ML models, data, and pipelines** to production environments, ensuring that updates are deployed as soon as possible, maintaining reliability after passing all validation steps.

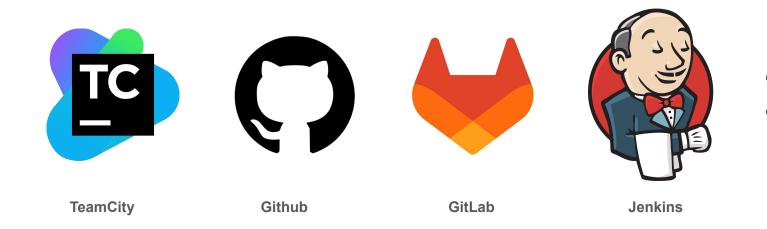
Benefits:

- All the CI benefits
- Performance validation: Ensures metrics are within tolerance
- Scalability: Leverages cloud infrastructure

Challenges:

- Data drift
- Validation of complex models such as LLMs
- Cost

Some of the common CI CD provider and tools:





CircleCl

TravisCl

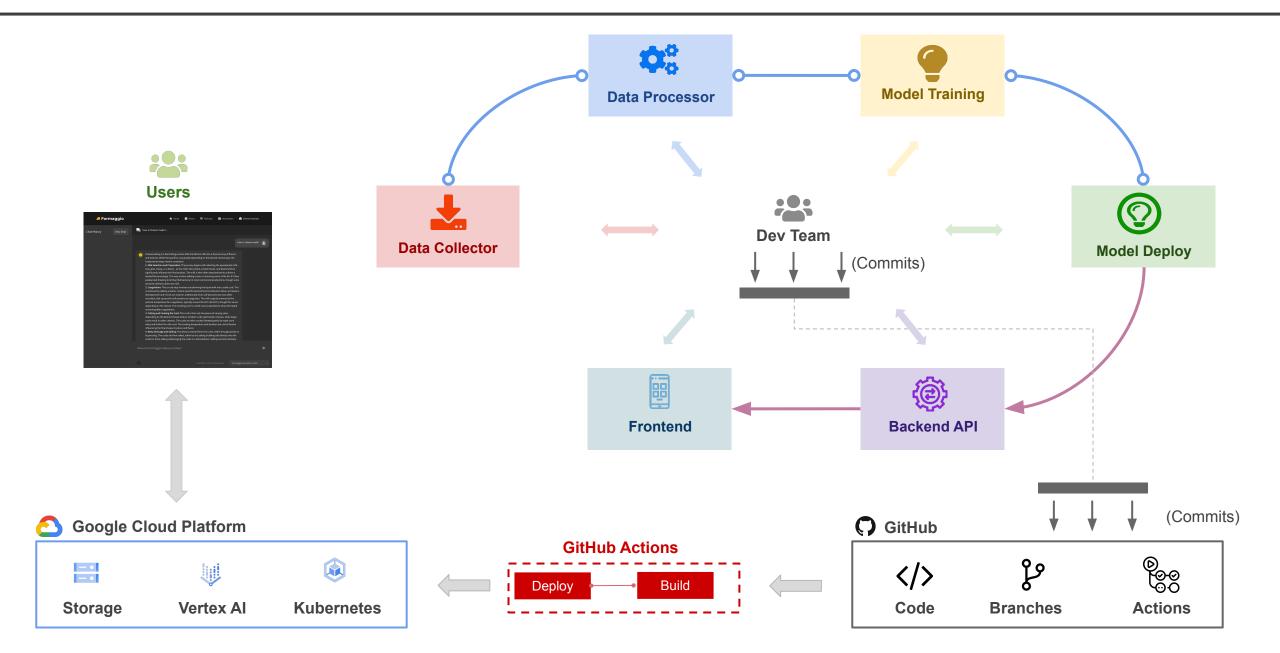
We already have a deployment container that can:

- Build Docker images.
- Run Vertex AI pipeline jobs.
- Deploy app to K8s cluster.

We can automate using GitHub Actions by:

- Monitoring code commits.
- Build & run deployment container.
- Invoke CLI in deployment container to:
 - Build & push docker images for release
 - Run Vertex AI jobs using new newly build images
 - Deploy newly build images of app to K8s cluster

Cheese App: CI CD



Tutorial: Continuous Integration, Continuous Deployment

Steps to apply CI / CD on the cheese app components:

• Create a Github / Workflow file defining deployment steps

https://github.com/dlops-io/cheese-app-v4/tree/main/.github/workflows

• Commit your code using /deploy-app to the commit message

Basic format
git commit -m "your message /deploy-app"

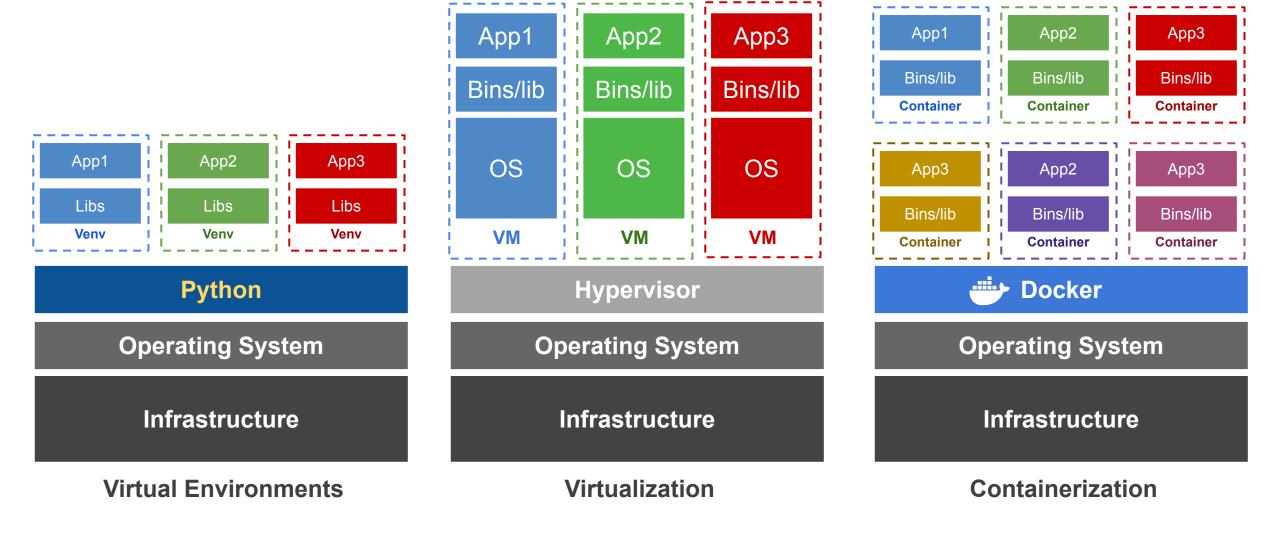
For detailed instructions, please refer to the following link

- <u>Cheese App CICD</u>.
 (<u>https://github.com/dlops-io/cheese-app-v4#cheese-app---automatic</u>
- <u>Cheese App GitHub Actions</u>.
 (<u>https://github.com/dlops-io/cheese-app-v4/actions</u>)

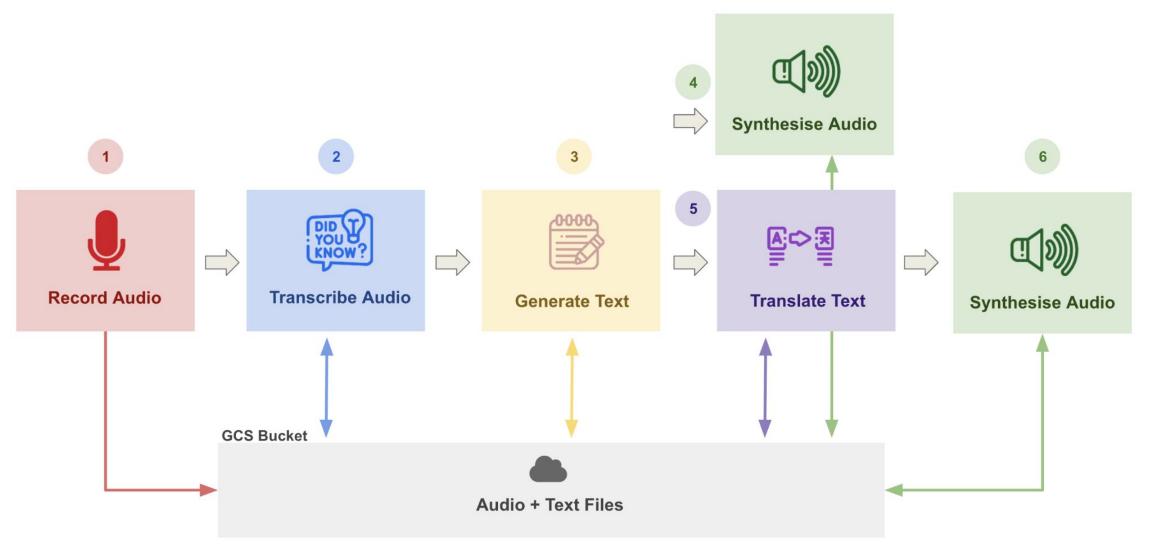


Review

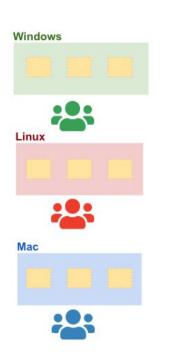
Lecture 2-4: Environments vs Virtualization vs Containerization

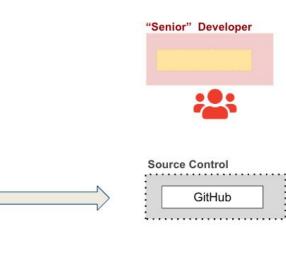


Lecture 2-4: Environments vs Virtualization vs Containerization



Workflow with Docker: Scenario 2 (later stages of development)

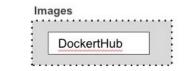




• Developers submit a pull request to GitHub for their code.

• In some cases, they may also include changes to the Dockerfile or Pipfiles in the pull request.

The senior developer reviews and merges all pull requests, then consolidates the updates into a new Dockerfile and Pipfiles and builds new images which are pushed to DockerHub.

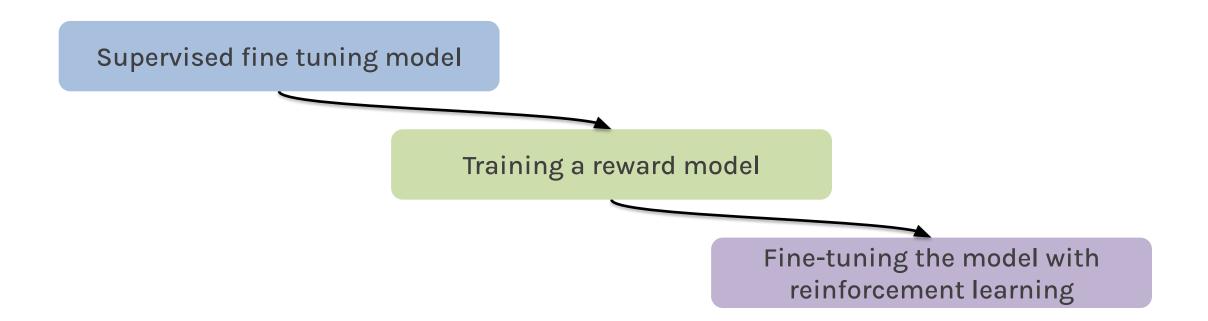


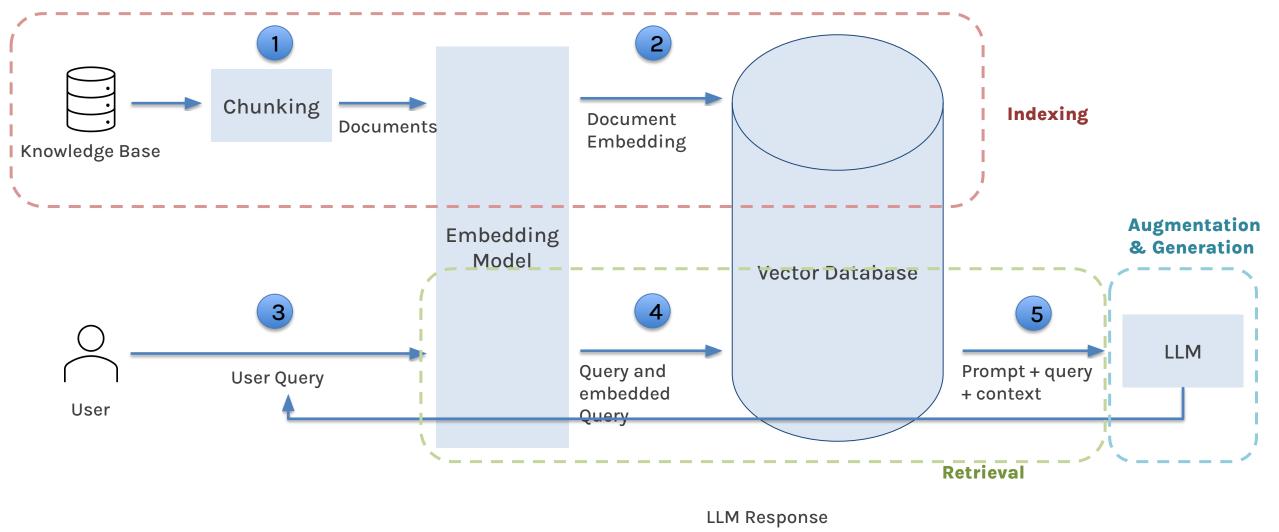
Lecture 6: Data Labeling and Data Versioning

| Unlabeled | How do we do this? | Labeled brie gouda | 4 Colab {training} |
|----------------------------|---|--|-----------------------|
| | | parmigiano <table-cell-columns> 💽 gruyere 📰 🜷</table-cell-columns> | |
| GCS Bucket | cheese-app-data-demo | | |
| 2 | | | |
| Data Labeling Container | 3 = Label Studio Signature Studio | | |

Lecture 7: Instruction based GPT => ChatGPT

We will break it down into 3 steps:

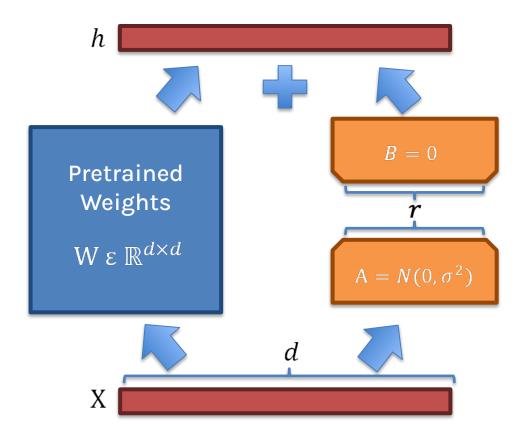








Lecture 9-10: Fine Tuning LLMs (LoRA, qLoRA)

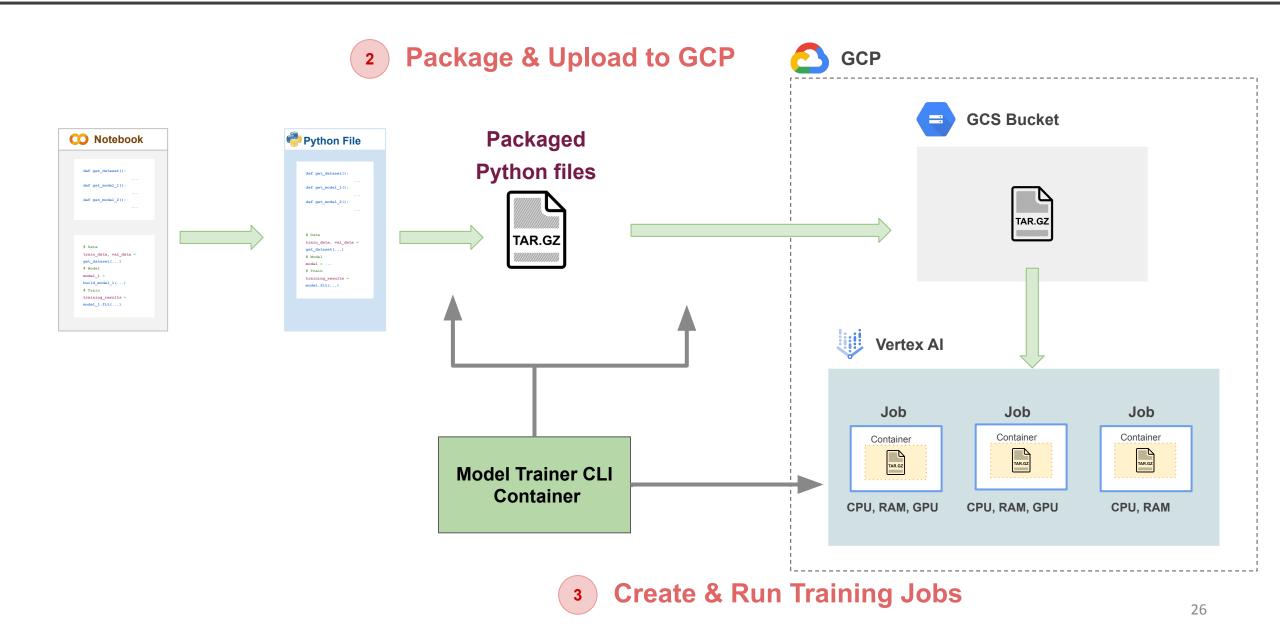


Notice how the reparameterization (LoRA) runs parallel to the original model.

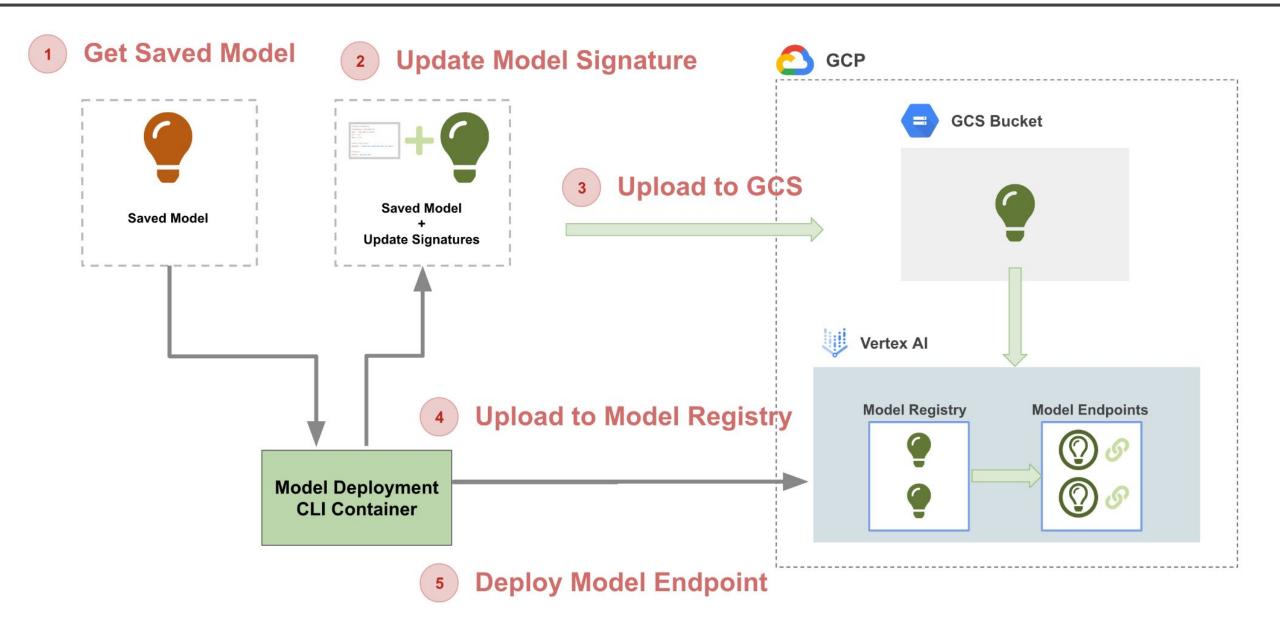
Lecture 11: Distillation

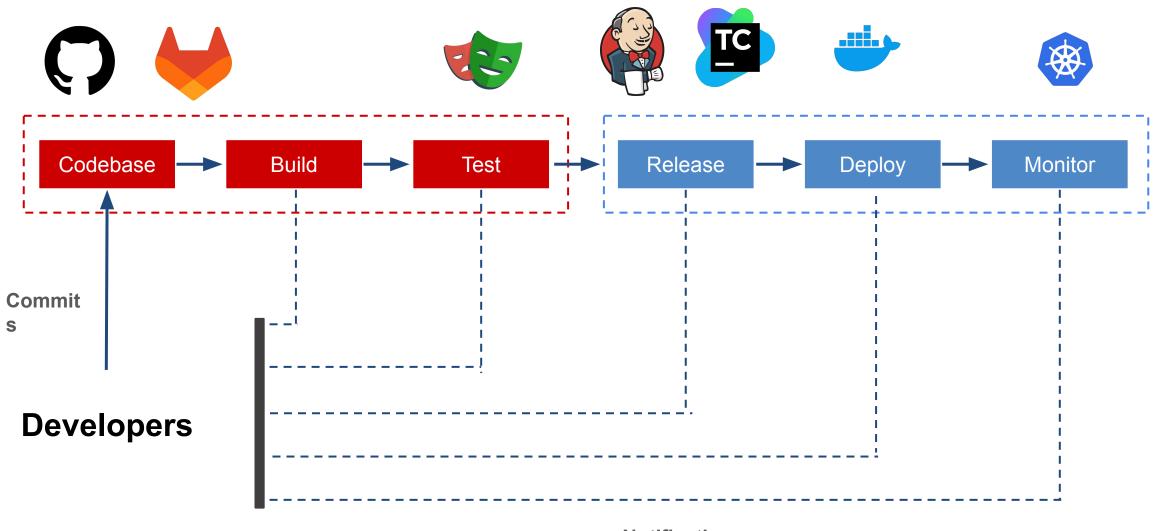


Lecture 12-13: Weight and Biases/VertexAI serveless training



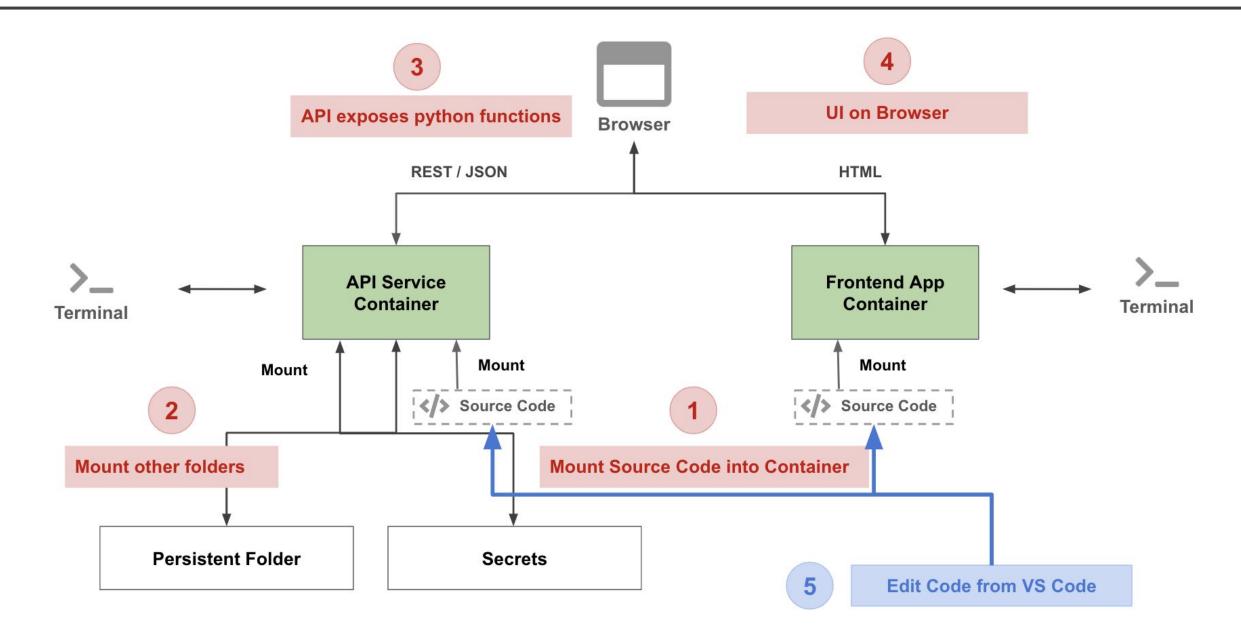
Lecture 15: Model Deployment with VertexAl

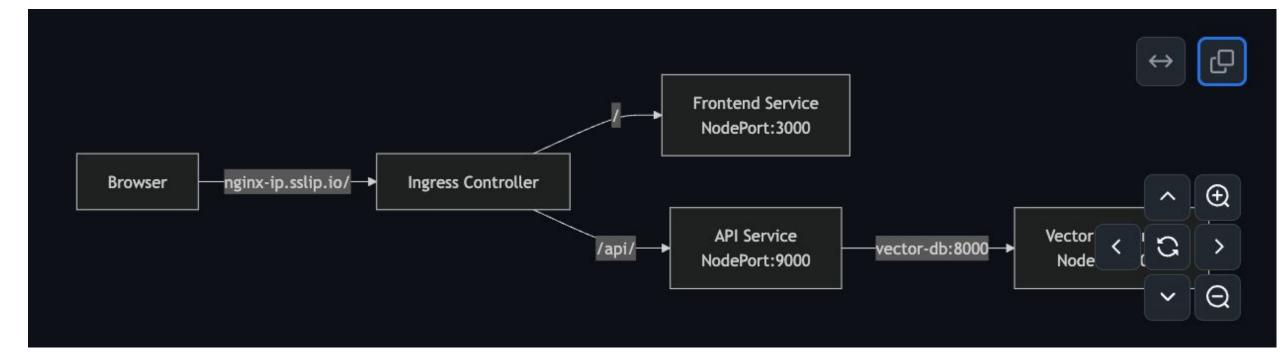




Notifications

Lecture 17-18: Frontend, Fast API





THANK YOU