Movie Recommendations

Group Project Presentation
[Machine Learning in Production]



Team: K-Avengers

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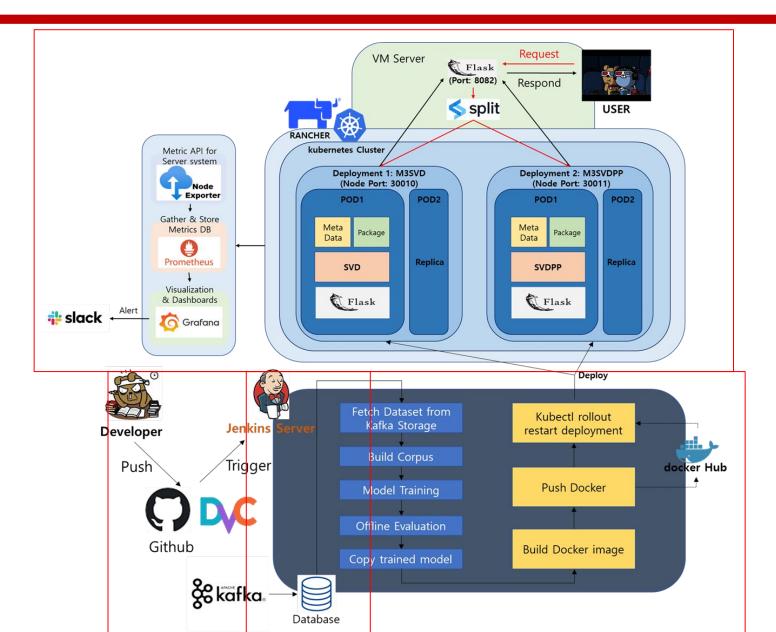
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- Conclusion(20s)

Overall Architecture





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• Overall CI pipeline for movie recommendation



Data storage

- Apache Kafka is a distributed event store and stream-processing platform.
- Collect Kafka log data
 - Data (movies watched by user) \rightarrow for (re)training model and for online evaluation
 - Rate (rating by user) \rightarrow for (re)training model and for online evaluation
 - Request → for online evaluation
- This pipeline, once run, continues to run until it is intentionally stopped.
- After online evaluation, expired data is automatically deleted.



• Overall CI pipeline for movie recommendation

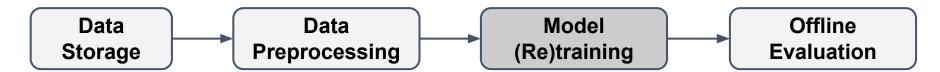


Data preprocessing

- Pre-process the stored raw data
- Generate a compressed sparse row (CSR) matrix
- Split it into train/validation sets



Overall CI pipeline for movie recommendation



- Model (re)training
 - Matrix Factorization (MF)
 - SVD

$$\hat{r}_{ui} = \mu + b_u + b_i + q_i^T p_u$$

■ SVD++

$$\hat{r}_{ui} = \mu + b_u + b_i + q_i^T \left(p_u + \left| I_u
ight|^{-rac{1}{2}} \sum_{j \in I_u} y_j
ight)$$



• Overall CI pipeline for movie recommendation



- Offline evaluation
 - We use 'RMSE' as metric for offline evaluation.
 - Root mean square error (RMSE)

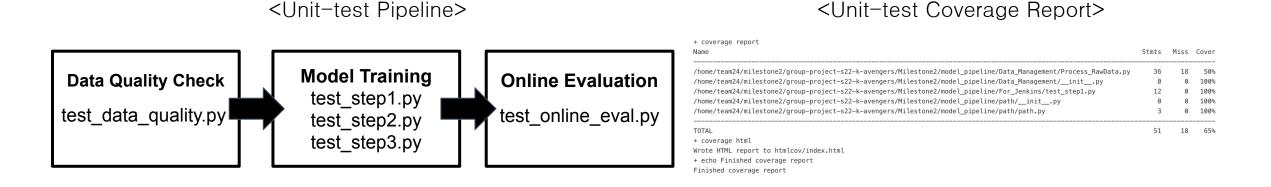
$$RMSE = \sqrt{\frac{\sum\limits_{(i,j)\in E}(r_{ij}-\widehat{r_{ij}})^{2}}{|E|}}, where E = nonzero \ entries \ in \ the \ test set.$$

Code Integrity Check with Unit-test



Unit-test

- Unittest
 - Python's built in library.
- The process is integrated on Jenkins pipeline, which runs automatically.
- The result can be identified in a coverage report format on Jenkins.



Automatic integration pipeline with Jenkins



Continuous integration

- Jenkins
 - Pipeline
 - Unit test 1 to 5 → model management & offline evaluation(model)
 → online evaluation
 - Using Blue Ocean plugin
 - A more visualized dashboard than ever before
 - Commit occurs in master branch of github → Autorun the entire pipeline
 - Save after pipeline build → Jenkinsfile for pipeline is committed to master branch on github
 - Using freestyle project
 - Automatically run once in a specific period of time
 - Setting the "build periodically" option

```
pipeline {
         stage('unit-1') {
           steps {
     pip3 install matplotlib
     pip3 install numpy
     echo \'unit 1\'
     coverage run --omit=\'/usr/lib*\' /home/team24/group-project-s
13
     coverage report
     echo \'Finished coverage report\'''
18
19
20
         stage('unit-2') {
21
22
             sh '''echo \'unit 2\'
23
24
     PYTHONPATH=/home/team24/group-project-s22-k-avengers/Milestone
28
     coverage report
```

model

online evaluation





unit-1

unit-2

unit-3

unit-4

unit-5







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Continuous Deployment

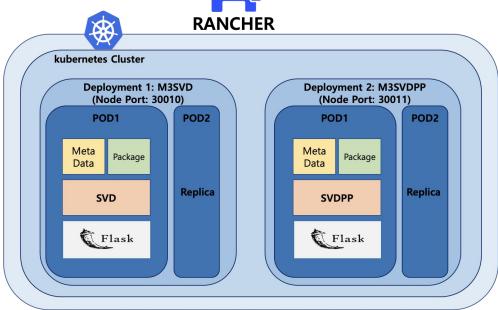
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Containerization with Rancher



Containerization

- Rancher
 - A complete container management platform
 that includes everything necessary for container management
 during the production process
- Deployment components
 - Our system manages two recommendation models as different deployments in one cluster
 - Each deployment consists of two pods, one replica
 of the other, which distributes and processes tasks

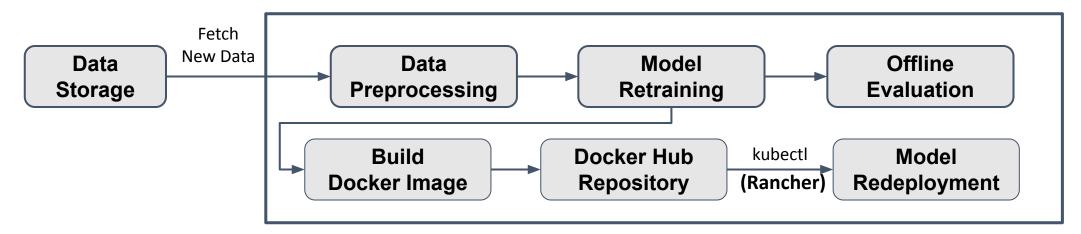


Overview of Our Deployments

Automatic Continuous Deployment with Jenkins



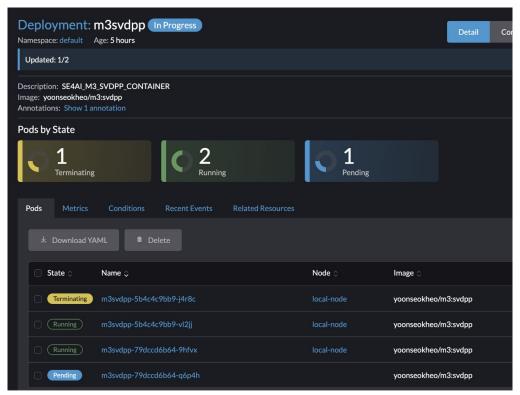
- Contribution: Automatic Continuous Deployment with Jenkins
 - Extending our integration pipeline to model deployment
 - We leverage jenkins to transmit the deployment signal to the Rancher
 - Whenever committed to Github, the pipeline is executed:
 - Continuous Integration: Data fetching, Data preprocessing, Model retraining
 - Continuous Deployment: Build docker images, Push images to docker repo
 - Model deployment: Pull docker images for retrained models and redeploy it through Rancher (* command: kubetcl rollout restart deployment)



Model Re-deployment without Downtime



- Contribution: ZERO downtime for model redeployment
 - The new redeployment also has 2 pods with replica
 - After one new pod is deployed, one existing pod is terminated
 - After a new pod is deployed again,
 the remaining existing pod is also terminated
 - ⇒ **ZERO downtime** in the process of deploying the re-trained models
- All these processes are stably controlled under the Rancher platform, and even now, the process of retraining and redeployment is being stably operated periodically

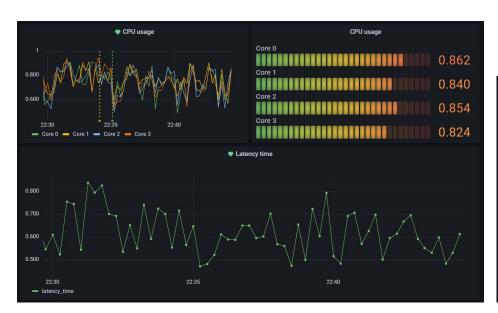


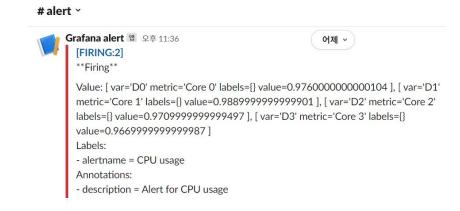
Monitoring

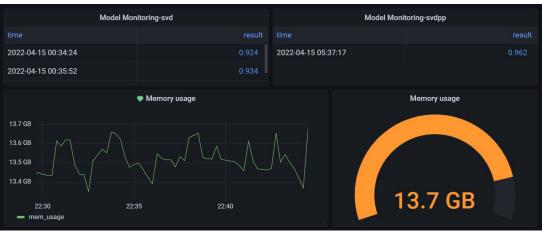


Monitoring infrastructure

- Prometheus, Grafana and Node Exporter to monitor our infrastructure
 - Memory usage
 - CPU usage
 - Latency time in flask
 - Model quality
- Sending alerts to our slack #alert channel







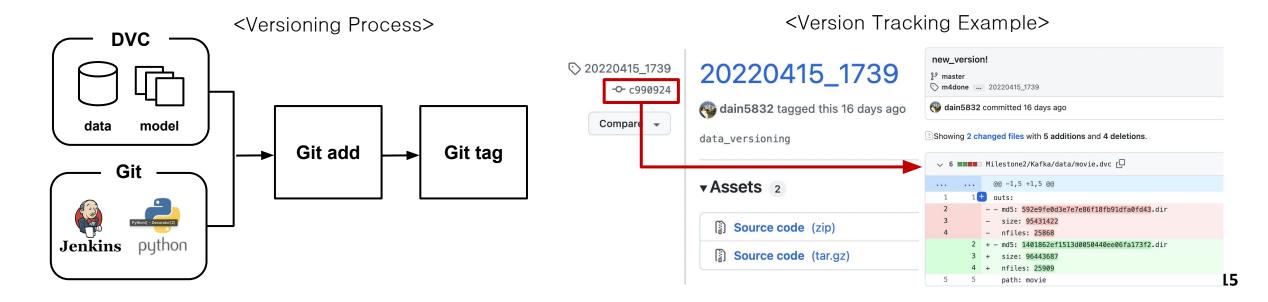
Versioning and Tracking Provenance



Provenance

1

- o DVC
 - an open-source version control system.
 - DVC stores the information of dataset and the model in .dvc format.
- Process
 - lacktriangle track modification ightarrow add changes to git ightarrow push git tag



Reflection



Things went very well



- CI & CD
 - One pipeline using Jenkins platform
- Zero-downtime
 - Rolling update with replica in Rancher platform
- A/B test
 - Continuous A/B test & Load balancing
- Things went less well
 - Less well used Git branch
 - Misallocated storage
 - Less efficiently written code
 - Server maintenance issue



Conclusion



We have learned how to:

- Collect data from multiple sources and engineer features for learning
- Deploy and measure a model inference service
- Build and operate infrastructures
 - a continuous integration infrastructure for evaluate a model in production
 - a monitoring infrastructure for system health and model quality
 - a continuous deployment infrastructure for automatic periodic retraining and versioning
- Design and implement a monitoring strategy to detect possible issues in machine-learning systems.



