

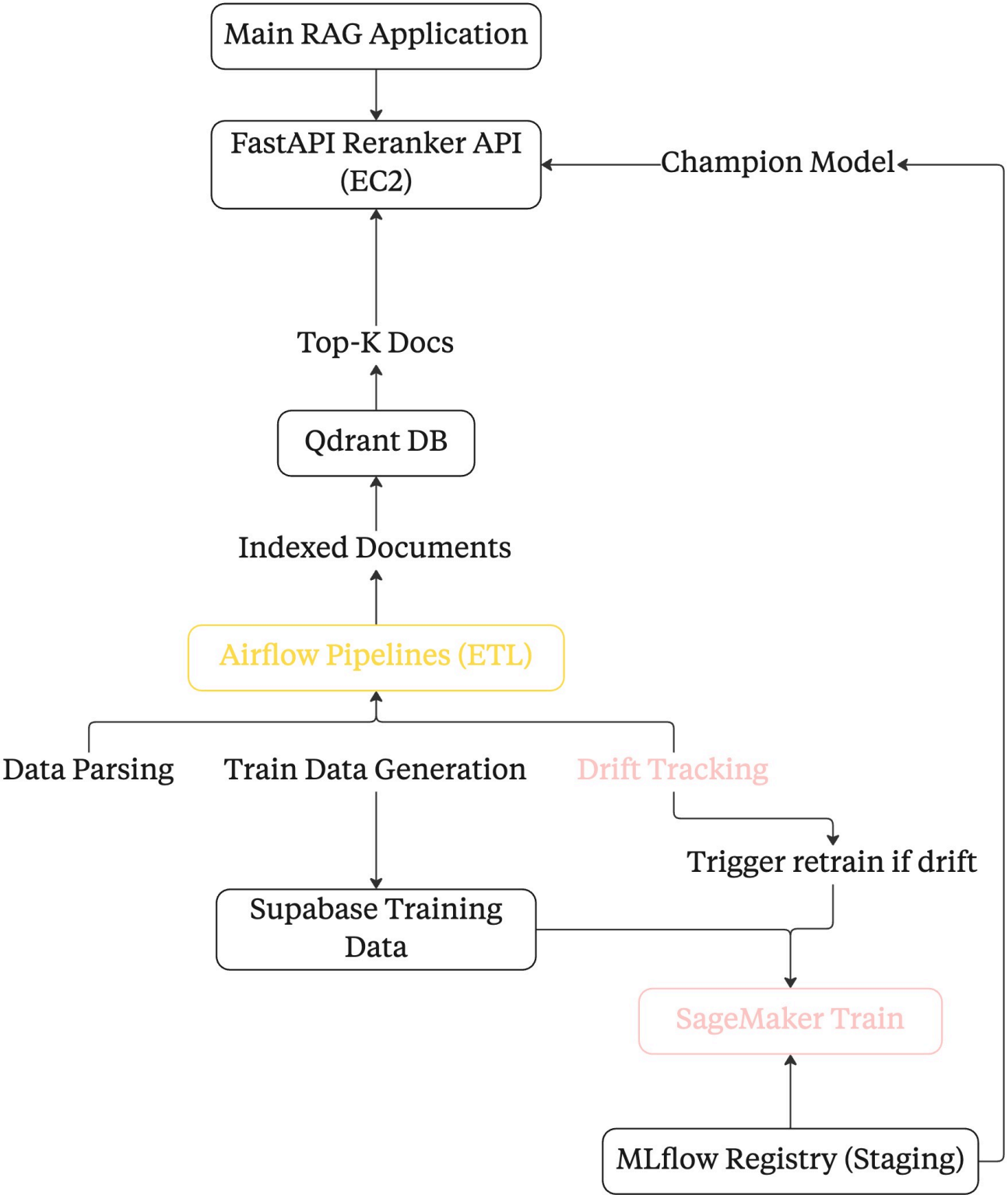
MLOps Lab4 Assignment

Task Summary

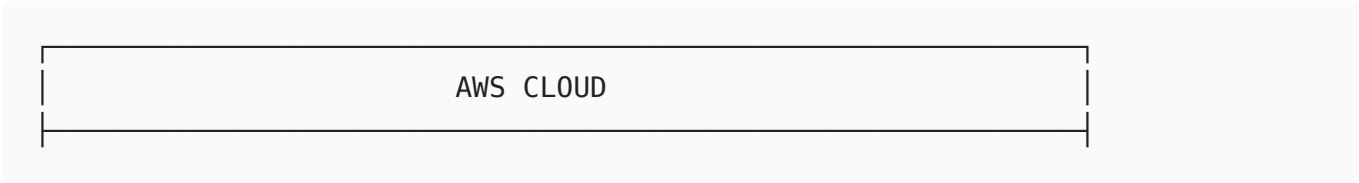
In this task, you will need to finish the full pipeline for the project from assignment 1. Your pipeline should include the following: gather the full pipeline of the model on the sagemaker, databricks or Kubeflow, and combine with the previous tasks using you favourite tools for everything (or tools that make the most sense) and prepare a short presentation/demo. The pipeline should include the following steps: data gathering (using airflow or other orchestrator, from prev. assignment), data processing step, model training step, parameters tuning step (this one is optional, if it makes sense), registering the model in model registry, logging all needed metrics, evaluating to detect if it is better then previous model, serving the best model.

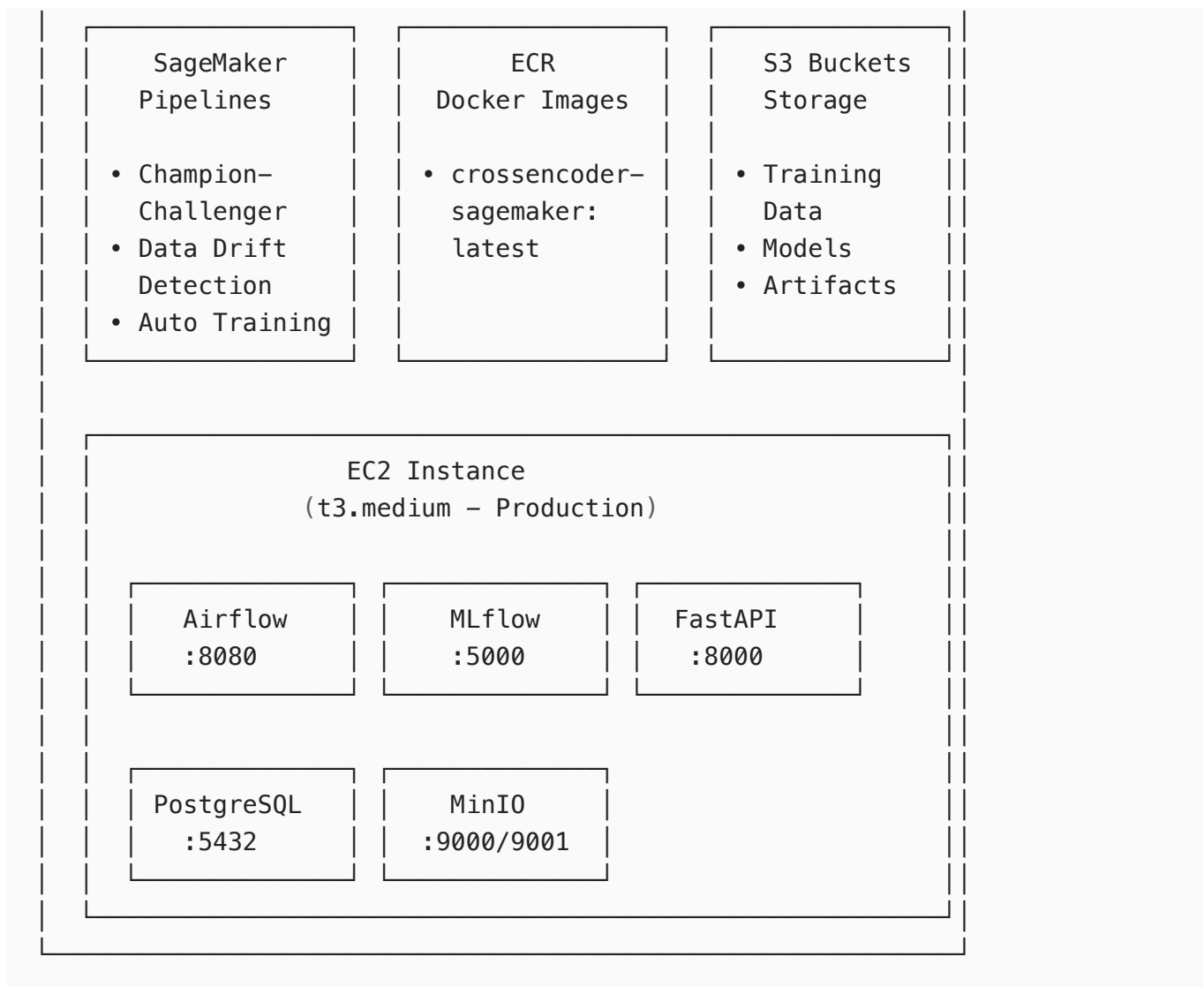
Note: all artefacts for lab results demo are located in the `ML0ps/reranker_cloud_demo` folder

Final MLOps Pipeline is as follows:



The AWS cloud infrastructure is as follows:





Side Preparations

Firstly we have to create a new EC2 instance with custom security group to open all necessary ports for future services.

Note: base for this EC2 instance is Amazon Linux (random choice, but it will affect the syntax of the following commands)

EC2 > Security Groups > sg-0c3e04f64503410a6 - launch-wizard-2

EC2 < **sg-0c3e04f64503410a6 - launch-wizard-2** Actions

Details

Security group name launch-wizard-2	Security group ID sg-0c3e04f64503410a6	Description launch-wizard-2 created 2025-07-27T21:04:48.791Z	VPC ID vpc-0d5fab91ece93233d
Owner 416607071613	Inbound rules count 6 Permission entries	Outbound rules count 1 Permission entry	

Inbound rules | Outbound rules | Sharing - new | VPC associations - new | Tags

Inbound rules (6) Manage tags Edit inbound rules

<input type="checkbox"/>	Name	Security group rule ID	IP version	Type	Protocol	Port range	Source
<input type="checkbox"/>	-	sgr-04df4d43eebf020de	IPv4	Custom TCP	TCP	5000	0.0.0.0/0
<input type="checkbox"/>	-	sgr-0b08529094d173d96	IPv4	Custom TCP	TCP	8000	0.0.0.0/0
<input type="checkbox"/>	-	sgr-0ef013fdae335bfbe	IPv4	HTTP5	TCP	443	0.0.0.0/0
<input type="checkbox"/>	-	sgr-0d3e7e980f9124591	IPv4	Custom TCP	TCP	8080	0.0.0.0/0
<input type="checkbox"/>	-	sgr-04994385504c902b6	IPv4	HTTP	TCP	80	0.0.0.0/0
<input type="checkbox"/>	-	sgr-043dde328c2076e3e	IPv4	SSH	TCP	22	0.0.0.0/0

Then standart connection procedure:

```
chmod 400 cloud-demo-key.pem
ssh -i "cloud-demo-key.pem" ec2-user@ec2-16-170-226-195.eu-north-1.compute.amazonaws.com
```

Now we can upgrade the environment and install docker to build main services for next steps:

```
sudo yum update -y

sudo yum install -y docker
sudo systemctl start docker
sudo systemctl enable docker
sudo usermod -aG docker ec2-user

sudo curl -L
"https://github.com/docker/compose/releases/download/v2.20.2/docker-compose-
linux-x86_64" -o /usr/local/bin/docker-compose sudo chmod +x
/usr/local/bin/docker-compose

sudo yum install -y git vim curl wget htop
```

```

      #_
    ~\  #####_      Amazon Linux 2023
  ~ ~  \_#####\
  ~ ~   \###|
  ~ ~    \#/  ---
  ~ ~      V~'  '->
  ~ ~
  ~ ~
  ~ ~. _ .
    _/  _/
    _/m/'

```

Last login: Sun Jul 27 21:23:07 2025 from 176.122.119.210

[ec2-user@ip-172-31-33-146 ~]\$ docker --version

Docker version 25.0.8, build 0bab007

[ec2-user@ip-172-31-33-146 ~]\$ docker-compose --version

Docker Compose version v2.20.2

Create project folder on EC2:

```
mkdir -p ~/reranker_cloud_demo
```

Transfer the necessary files to the EC2 instance using SCP:

```

scp -i ~/Downloads/cloud-demo-key.pem docker-compose.yml ec2-
user@16.170.226.195:~/reranker_cloud_demo/
scp -i ~/Downloads/cloud-demo-key.pem Dockerfile.airflow ec2-
user@16.170.226.195:~/reranker_cloud_demo/
scp -i ~/Downloads/cloud-demo-key.pem Dockerfile.api ec2-
user@16.170.226.195:~/reranker_cloud_demo/
scp -i ~/Downloads/cloud-demo-key.pem Dockerfile.mlflow ec2-
user@16.170.226.195:~/reranker_cloud_demo/
scp -i ~/Downloads/cloud-demo-key.pem requirements.txt ec2-
user@16.170.226.195:~/reranker_cloud_demo/
scp -i ~/Downloads/cloud-demo-key.pem reranker_api_service.py ec2-
user@16.170.226.195:~/reranker_cloud_demo/
scp -r -i ~/Downloads/cloud-demo-key.pem dags/ ec2-
user@16.170.226.195:~/reranker_cloud_demo/

```

Check all files are transfer correctly:

```
-----
[[ec2-user@ip-172-31-33-146 ~]$ cd ~/reranker_cloud_demo
[[ec2-user@ip-172-31-33-146 reranker_cloud_demo]$ ls -la
total 28
drwxr-xr-x. 3 ec2-user ec2-user 172 Jul 27 21:41 .
drwx-----. 4 ec2-user ec2-user 122 Jul 27 21:35 ..
-rw-r--r--. 1 ec2-user ec2-user 759 Jul 27 21:41 Dockerfile.airflow
-rw-r--r--. 1 ec2-user ec2-user 591 Jul 27 21:35 Dockerfile.api
-rw-r--r--. 1 ec2-user ec2-user 990 Jul 27 21:41 Dockerfile.mlflow
drwxr-xr-x. 3 ec2-user ec2-user 129 Jul 27 21:41 dags
-rw-r--r--. 1 ec2-user ec2-user 5112 Jul 27 21:35 docker-compose.yml
-rw-r--r--. 1 ec2-user ec2-user 294 Jul 27 21:35 requirements.txt
-rw-r--r--. 1 ec2-user ec2-user 3376 Jul 27 21:35 reranker_api_service.py
[[ec2-user@ip-172-31-33-146 reranker_cloud_demo]$ ls -la dags/
total 32
drwxr-xr-x. 3 ec2-user ec2-user 129 Jul 27 21:41 .
drwxr-xr-x. 3 ec2-user ec2-user 172 Jul 27 21:41 ..
drwxr-xr-x. 2 ec2-user ec2-user 121 Jul 27 21:41 clients
-rw-r--r--. 1 ec2-user ec2-user 2340 Jul 27 21:41 data_pipeline_dag.py
-rw-r--r--. 1 ec2-user ec2-user 4181 Jul 27 21:41 generate_training_data.py
-rw-r--r--. 1 ec2-user ec2-user 10479 Jul 27 21:41 parse_raw_batch_data.py
-rw-r--r--. 1 ec2-user ec2-user 4779 Jul 27 21:41 utils.py
```

Finally, magic is going to happen:

```
docker-compose up --build -d
```

SageMaker Pipeline

Each stage of the pipeline—data preparation, training, drift tracking, and deployment—is encapsulated in a separate, reusable component, and then defined as a single step in the final pipeline.

Additionally some preparation files are present in main directory to setup the environment for execution.

```
sagemaker_pipeline/
├── scripts/
│   ├── deploy.py
│   ├── evaluate.py
│   ├── load_data.py
│   ├── prepare.py
│   ├── supabase_client_wrapper.py
│   ├── track_shift.py
│   └── train.py
├── build_and_push.sh           # Shell script to build and push the
                                Docker image to ECR
├── Dockerfile                 # Dockerfile used to create a container
                                for pipeline steps
```

```
└─ pipeline.py # Main orchestration script defining the SageMaker pipeline
└─ requirements-sagemaker.txt # Python dependencies needed in the SageMaker environment
```

Preparations

1. Create S3 bucket with folder structure as a preparing step in advance:

The screenshot shows the Amazon S3 console for the bucket 'crossencoder-pipeline-data'. The 'Objects' tab is selected, displaying a table of objects. The table has columns for Name, Type, Last modified, Size, and Storage class. The objects listed are folders: 'metrics/', 'model-artifacts/', 'processed/', and 'raw/'.

Name	Type	Last modified	Size	Storage class
metrics/	Folder	-	-	-
model-artifacts/	Folder	-	-	-
processed/	Folder	-	-	-
raw/	Folder	-	-	-

2. Create S3 Full Access Permission for the SageMaker Executor Role to have access to these crossencoder-pipeline-data bucket.

The screenshot shows the AWS IAM console for the role 'AmazonSageMaker-ExecutionRole-20250712T045048'. The 'Permissions' tab is selected, displaying a table of permissions policies. The table has columns for Policy name, Type, and Attached entities. The policies listed are: 'AmazonSageMaker-ExecutionPolicy-20250712T045048' (Customer managed, 1 attached entity), 'AmazonSageMakerCanvasAIServicesAccess' (AWS managed, 1 attached entity), 'AmazonSageMakerCanvasDataPrepFullAccess' (AWS managed, 1 attached entity), 'AmazonSageMakerCanvasFullAccess' (AWS managed, 1 attached entity), 'AmazonSageMakerFullAccess' (AWS managed, 2 attached entities), and 'AWS-S3_FullAccess' (Customer inline, 0 attached entities).

Policy name	Type	Attached entities
AmazonSageMaker-ExecutionPolicy-20250712T045048	Customer managed	1
AmazonSageMakerCanvasAIServicesAccess	AWS managed	1
AmazonSageMakerCanvasDataPrepFullAccess	AWS managed	1
AmazonSageMakerCanvasFullAccess	AWS managed	1
AmazonSageMakerFullAccess	AWS managed	2
AWS-S3_FullAccess	Customer inline	0

3. Create custom docker image based in it to have a suitable environment with all necessary dependencies for future pipeline execution.
 1. Prepare requirements-sagemaker.txt for data shift/drift tracking, training, evaluation and deployment task.
 2. Prepare Dockerfile based on requirements-sagemaker.txt .

3. Prepare `build_and_push.sh` for Dockerfile to become an accessible image.

```
sagemaker-crossencoder $ chmod +x build_and_push.sh
sagemaker-crossencoder $ ./build_and_push.sh
Building image: 416607071613.dkr.ecr.eu-north-1.amazonaws.com/crossencoder-sagemaker:latest
```

4. Just copy the `scripts` directory and run `pipeline.py` in the SageMaker Studio.

Final SageMaker Pipeline Architecture

SageMaker Studio

Pipelines

View

CrossEncoderChampionChallengerPipeline

Executions

Cf7xo0755oxi

Graph

Provide feedback

Whats new

Applications (3)

JupyterLab

Code Editor

MLflow

Partner AI Apps

New

Home

Running instances

Compute

Data

Auto ML

Experiments

Jobs

Pipelines

Models

JumpStart

Deployments

Collapse Menu

PIPELINE EXECUTION

execution-1753260565991

Succeeded

Execute latest version

Scheduler

Visual editor

GraphParametersInformation

LoadShiftData

Process data

TrackDataDrift

Process data

CheckShiftTrigger

Condition

true

LoadTrainData

Process data

PrepareDatasets

Process data

TrainModel

Process data

EvaluateNewModel

Process data

EvaluateProductionModel

Process data

CompareModels

Condition

true

DeployNewModel

Process data

Select step

Select step

75%

© 2025, Amazon Web Services, Inc. or its affiliates. All rights reserved.

