

AWS SageMaker

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Agenda

- Everything-as-a-Service for Machine Learning
- Tutorial
- Conclusion

Everything-as-a-Service for Machine Learning

- AWS SageMaker started in 2017 [1]
- Started as a platform for analytics, optimization, and machine learning
- Renamed to AWS SageMaker AI in 2024 [2]
- **Focus on the machine learning workflow**

Data

- Fully integrated into AWS data management services, e.g. EMR [3]
- Data labeling and model validation AWS Ground Truth [4] and Mechanical Turk [5]
- Cloud services for human workforce

Everything-as-a-Service for Machine Learning

Development

- Jupyter notebooks for preprocessing and development
- SageMaker has a widely used Python SDK [6]
- And a IDE called SageMaker Studio [7]

Everything-as-a-Service for Machine Learning

Deployment

- Run pre-built or self-developed models on accelerated EC2 instances
- The more responsive, the more expensive [8]
- E.g. real-time inference with auto-scaling vs. serverless vs. batch transform overnight

Tutorial

- Predict shared bike availability
- Load data into the cloud
- Define and train the model
- Provision it securely via an endpoint

IAM Configuration

- ↑ Permissions for SageMaker resources are handled through IAM
- We need to grant the user access to SM-related resources
- ↓ We recommend doing it in a real (paid) AWS account to make sure all resources are available

```
"Statement": [  
  {  
    "Effect": "Allow",  
    "Principal": {  
      "Service": "sagemaker.amazonaws.com"  
    },  
    "Action": "sts:AssumeRole"  
  }  
]
```

Figure: Create a new IAM Role in **IAM** → **Access Management** → **Roles** called AmazonSageMaker-TrainingExecutionRole

S3 Configuration

- ↑ SM can get data from all kinds of services and formats
- ↑ We settled for S3 because it is the simplest solution for testing
- ↓ Other persistent resources (e.g. RDS) can be expensive, even if we turn them off

Create new **S3 bucket**

- Set name to ccbda-research-sagemaker
- Leave all other settings as default

Create Jupyter Notebook for development

- Jupyter notebooks are great for exploring and plotting data
- ↑ Run seamless on AWS
- ↑ Easy to collaborate with others
- ↓ Run on remote accelerated EC2 instances which can become expensive
- ↑ Users can download and execute notebooks step-by-step, e.g. our tutorial in `sagemaker_ml.ipynb`

Handling the dataset

We are gonna use the **Seoul Bike Sharing Demand** dataset [9].

↑ Each cell in the notebook corresponds to:

- Loading the dataset
- Cleaning the data
- Splitting into train and test set
- Uploading it to our S3 bucket

Training Job Configuration

```
xgboost_image_uri = image_uris.retrieve("xgboost",  
    ↪ region=region, version="1.5-1")  
estimator = Estimator(  
    image_uri=xgboost_image_uri,  
    role=role,  
    instance_count=1,  
    instance_type="ml.m5.large",  
    volume_size=5,  
    ...  
)  
estimator.set_hyperparameters(  
    objective="reg:squarederror",  
    num_round=100,  
    max_depth=5,  
    subsample=0.8,  
    ...  
)
```

Training Job goes to the Queue

Training jobs [Info](#)



Actions ▼

Create training job

 Search training jobs

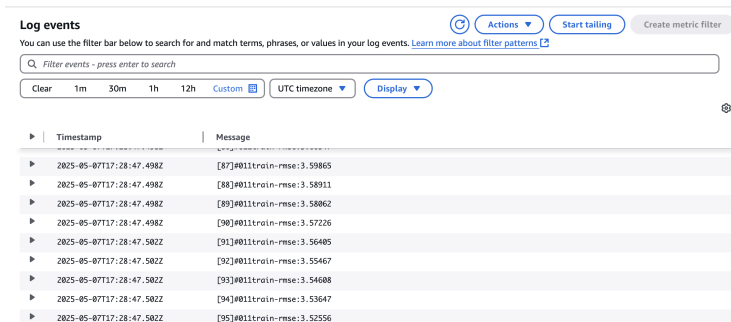
< 1 >

| | Name | Creation time | Duration | Job status | Warm pool status | Time left |
|--|---|----------------------|-----------|------------|------------------|-----------|
| | sagemaker-xgboost-2025-05-07-17-26-25-385 | 5/7/2025, 7:26:25 PM | 3 minutes | Completed | - | - |

Figure: Created training job

CloudWatch Integration

- ↑ Training and deployment logs are integrated in CloudWatch



The screenshot displays the AWS CloudWatch 'Log events' console. At the top, there are buttons for 'Actions', 'Start tailing', and 'Create metric filter'. Below these is a search bar with the placeholder text 'Filter events - press enter to search'. Under the search bar, there are filters for 'Clear', '1m', '30m', '1h', '12h', 'Custom', 'UTC timezone', and 'Display'. The main area shows a table of log events with two columns: 'Timestamp' and 'Message'. The table contains 10 rows of log data, each with a timestamp and a message starting with '[87]#011train-rmse:3.59865'.

| Timestamp | Message |
|--------------------------|----------------------------|
| 2025-05-07T17:28:47.498Z | [87]#011train-rmse:3.59865 |
| 2025-05-07T17:28:47.498Z | [88]#011train-rmse:3.58911 |
| 2025-05-07T17:28:47.498Z | [89]#011train-rmse:3.58062 |
| 2025-05-07T17:28:47.498Z | [90]#011train-rmse:3.57226 |
| 2025-05-07T17:28:47.502Z | [91]#011train-rmse:3.56405 |
| 2025-05-07T17:28:47.502Z | [92]#011train-rmse:3.55467 |
| 2025-05-07T17:28:47.502Z | [93]#011train-rmse:3.54608 |
| 2025-05-07T17:28:47.502Z | [94]#011train-rmse:3.53647 |
| 2025-05-07T17:28:47.502Z | [95]#011train-rmse:3.52556 |

Figure: Training Job Logs

Deploying Endpoint from the Notebook

- ↑ With a single line, we can deploy the trained model
- ↑ We have full control from here, e.g. about instance size, scaling policy

```
predictor = estimator.deploy(initial_instance_count=1,  
    ↪ instance_type="ml.m5.large")
```

The Endpoint is deployed & scaled automatically

Endpoints

Update endpoint

Actions

Create endpoint

Search endpoints

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>

| | Name | ARN | Creation time | Status | Last updated |
|-------------|---|--|----------------------|---------------------------------|----------------------|
| <div></div> | sagemaker-xgboost-2025-05-07-17-31-35-496 | arn:aws:sagemaker:eu-north-1:940819259195:endpoint/sagemaker-xgboost-2025-05-07-17-31-35-496 | 5/7/2025, 7:31:36 PM | <div><div></div>InService</div> | 5/7/2025, 7:36:01 PM |

Figure: Published endpoint

Conclusion & Opinion

- ↑ AWS SageMaker AI is feature complete for all aspects of ML
- ↑ We got started very fast thanks to the notebooks
- ↑ We have fine grain control over expected performance and QoS
- ↑↓ AWS provides adequate tooling, e.g. Python SDK
- ↑ Features industry-standard libraries, e.g. TensorFlow

References & Questions I

- [1] Ron Miller. *AWS releases SageMaker to make it easier to build and deploy machine learning models*. 8 May 2025. 2017. URL: <https://techcrunch.com/2017/11/29/aws-releases-sagemaker-to-make-it-easier-to-build-and-deploy-machine-learning-models/>.
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<https://docs.aws.amazon.com/sagemaker/latest/dg/sms-workforce-management-public.html>. Accessed: 2025-05-08.
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References & Questions III

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