

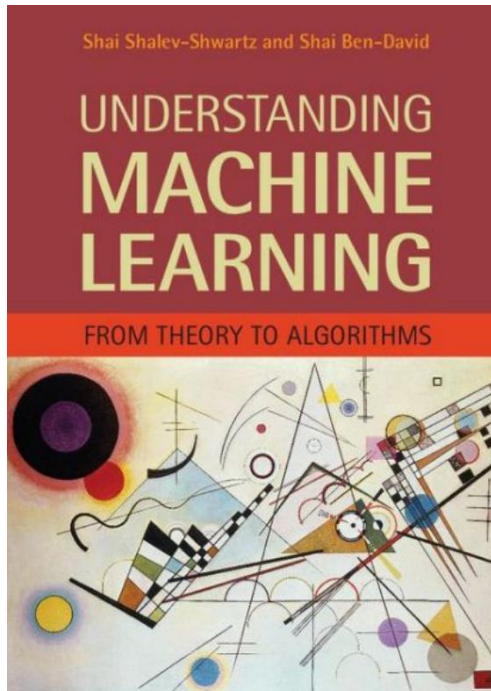
Cloud Computing Linear Regression-AWS sagemaker

Farid Afzali, Ph.D., P.Eng.

Exploratory Data Analysis

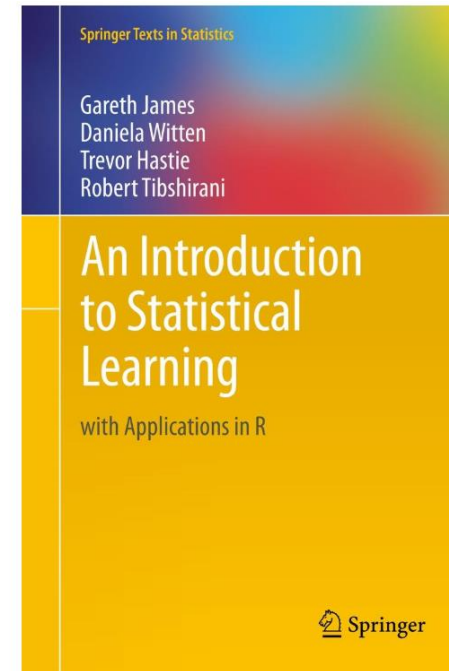
Additional Resources, Page #123:

<http://www.cs.huji.ac.il/~shais/UnderstandingMachineLearning/understanding-machine-learning-theory-algorithms.pdf>



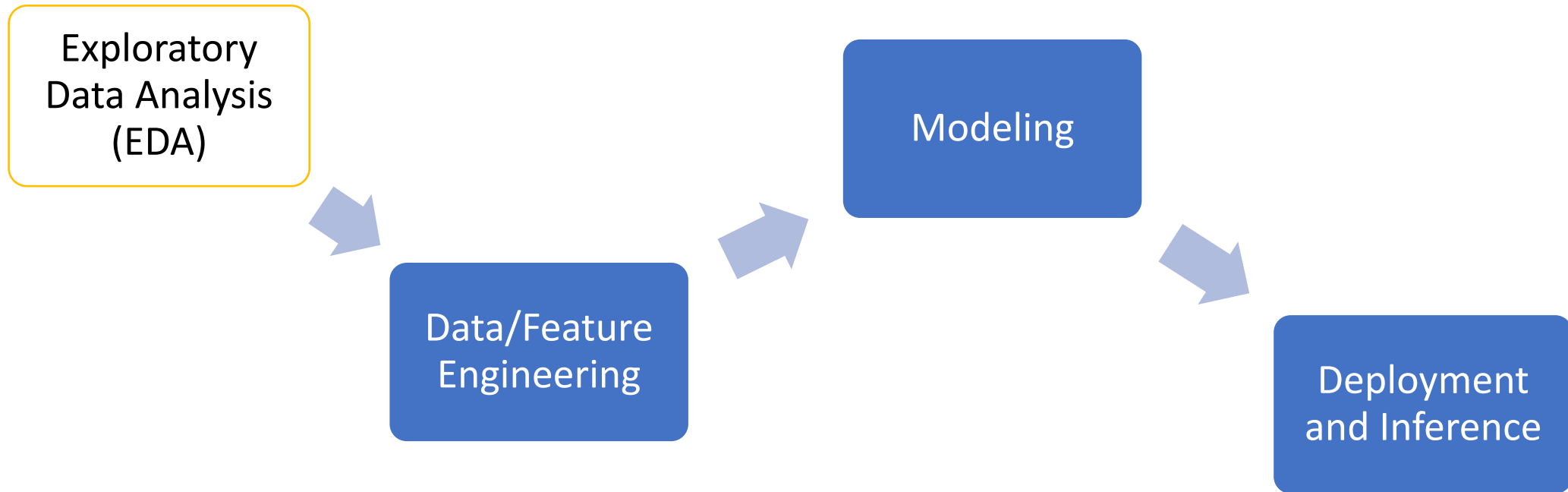
Additional Resources, Page #61:

<http://www-bcf.usc.edu/~gareth/ISL/ISLR%20Seventh%20Printing.pdf>



Dr. Ryan Ahmed

EDA



University admission project



- ❑ **Goal:** The project's objective is to build, train, test, and deploy a machine learning model that can predict the likelihood of university admission based on a student's profile.
- ❑ **Tool:** AWS SageMaker is mentioned as the tool for launching a training job from the Management Console, which is a part of AWS that allows you to build, train, and deploy machine learning models at scale.
- ❑ **Practical Real-World Application:** The machine learning model is intended to be used by university admissions departments to identify the most qualified students.
- ❑ **Data:**
 - ❑ **Inputs (Features):** These are the parameters the model will use to make its predictions. They include GRE scores, TOEFL scores, University Rating, Statement of Purpose (SOP) quality, Letter of Recommendation (LOR) strength, Undergraduate GPA, and Research Experience.
 - ❑ **Outputs:** The model will output the probability of admission, which will range from 0 to 1 (with 1 likely indicating certain admission and 0 indicating no chance of admission).
- ❑ The **Data Source** for the machine learning model is provided as a [URL](#), which is a Kaggle link. Kaggle is a popular platform for data science competitions and datasets.

University admission project



- **Inputs:** These are the variables or features the model will use to make its predictions. They include:
 - ☒ GRE Score
 - ☒ TOEFL Score
 - ☒ University Rating
 - ☒ Statement of Purpose (SOP)
 - ☒ Letter of Recommendation (LOR)
 - ☒ Cumulative Grade Point Average (CGPA)
 - ☒ Research (likely a binary indicator of research experience)
- **Machine Learning Model:** This is the core computational component that processes the inputs to make a prediction. This component represents the algorithms and statistical methods that learn from the input data to make predictions or decisions without being explicitly programmed to perform the task.
- **Output:** The result of the machine learning model's processing of the inputs, which in this case is the "Chance of Admission." This is likely a probability score between 0 and 1 where a higher score indicates a higher likelihood of admission.

The impact of Machine Learning (ML) and Artificial Intelligence (AI) on various facets of higher education

[Link](#)

- ML and AI are presented as transformative forces in higher education, impacting several areas.
- The specific areas affected include:
 - **Marketing and Recruiting:** Using ML and AI to better target potential students and understand the effectiveness of marketing strategies.
 - **Students Admission and Enrollment:** Applying ML models, like the previously discussed predictive models, to streamline the admissions process and predict enrollment numbers.
 - **Curriculum and Resources Planning and Forecasting:** Utilizing AI for optimizing curriculum design based on trends and predicting resource requirements.
 - **Pedagogy and Personalized Students Learning Experience:** Implementing AI to customize the learning experience for students, adapting to their individual learning styles and needs.
 - **Students Support (AI-Powered Counseling):** Providing student support services using AI, such as virtual advisors for academic counseling.

More Resources

- Please read the 2 articles below and answer the following quiz.
 - <https://edtechmagazine.com/higher/article/2020/02/5-year-vision-artificial-intelligence-higher-ed>
 - <https://er.educause.edu/articles/2019/8/artificial-intelligence-in-higher-education-applications-promise-and-perils-and-ethical-questions>

Artificial Intelligence in Higher Education: Applications, Promise and Perils, and Ethical Questions

Elana Zeide Monday, August 26, 2019 In Print PDF

16 min read


What are the benefits and challenges of using artificial intelligence to promote student success, improve retention, streamline enrollment, and better manage resources in higher education?

SHARE



EDUCAUSE

On the hunt for a



A 5-Year Vision for Artificial Intelligence in Higher Ed

From machine learning algorithms to AI-driven assistants, higher ed will see more highly developed AI applications in the coming years.

by Micah Castelo

Micah Castelo is a web editor for EdTech Focus on K-12. Her experience includes education and community news coverage for the Syracuse Post-Standard and international news reporting for the Pulitzer Center on Crisis Reporting.


LISTEN 05:15

Artificial intelligence is all around us today, rapidly transforming and improving the way industries operate.

That includes the field of education, which is poised to discover new and evolving applications for AI, said Lukman Ramsey, a global solution manager at Google. Ramsey touched on some of those

Latest Articles

- How To Ace Your Windows 11 Migration
- Higher Education Institutions Are Using Firewall As A Service To Enhance Security
- What Software Asset Management Programs Do For Higher Education
- 3 Ways To Supercharge Higher Ed Device Management With An E-Procurement Solution



JumpStart

Foundation models

Computer vision models

Natural language processing models

Governance

Ground Truth

Notebook

Processing

Training

Algorithms

Training jobs

Hyperparameter tuning jobs

Inference

Augmented AI

AWS Marketplace

Amazon SageMaker > Domains

Domains

Info

A domain includes an associated Amazon Elastic File System (EFS) volume; a list of authorized users; and a variety of security, application, policy, and Amazon Virtual Private Cloud (VPC) configurations. Each user in a domain receives a personal and private home directory within the EFS for notebooks, Git repositories, and data files.

Domains (1)

Info



View

Edit

Create domain

Find domain name

< 1 > ⚙️

	Name	Id	Status	Created on	Modified on
<input type="radio"/>	QuickSetupDomain-20240221T155745	d-n0d8ljkkcebf	✓ InService	Feb 21, 2024 20:57 UTC	Feb 21, 2024 21:00 UTC



Create training job

When you create a training job, Amazon SageMaker sets up the distributed compute cluster, performs the training, and deletes the cluster when training has completed. The resulting model artifacts are stored in the location you specified when you created the training job. [Learn more](#)

Job settings

Job name

Maximum of 63 alphanumeric characters. Can include hyphens (-), but not spaces. Must be unique within your account in an AWS Region.

IAM role

Amazon SageMaker requires permissions to call other services on your behalf. Choose a role or let us create a role that has the [AmazonSageMakerFullAccess](#) IAM policy attached.

AmazonSageMakerServiceCatalogProductsUseRole ▼

[Create role using the role creation wizard](#)

Algorithm options

Use an Amazon SageMaker built-in algorithm, your own algorithm, or a third-party algorithm from AWS Marketplace.

▼ Algorithm source

☒ Amazon SageMaker built-in algorithm [Learn more](#)

✓

Successfully created bucket "admissioninuniversity"
To upload files and folders, or to configure additional bucket settings, choose **View details**.

View details

✕

Amazon S3 > Buckets

▶ Account snapshot

Storage lens provides visibility into storage usage and activity trends. [Learn more](#)

View Storage Lens dashboard

General purpose buckets | Directory buckets

General purpose buckets (3) [Info](#)

Copy ARN

Empty




Delete

Create bucket

Find buckets by name

< 1 > ⚙

	Name ▲	AWS Region ▼	Access ▼	Creation date ▼
<input type="radio"/>	admissioninuniversity	Canada (Central) ca-central-1	Bucket and objects not public	April 2, 2024, 16:16:36 (UTC-04:00)
<input type="radio"/>	sagemaker-studio-730335673928-4wmpfk8qzaa	US East (N. Virginia) us-east-1	Bucket and objects not public	March 27, 2024, 19:08:32 (UTC-04:00)
<input type="radio"/>	sagemaker-us-east-1-730335673928	US East (N. Virginia) us-east-1	Bucket and objects not public	March 27, 2024, 19:08:34 (UTC-04:00)

  Successfully created folder "model". [Amazon S3](#) > [Buckets](#) > admissioninuniversity

admissioninuniversity [Info](#)

[Objects](#)[Properties](#)[Permissions](#)[Metrics](#)[Management](#)[Access Points](#)

Objects (3) [Info](#)



Copy S3 URI

Copy URL

Download

Open

[Delete](#)[Actions](#) ▼[Create folder](#)

Upload

Objects are the fundamental entities stored in Amazon S3. You can use [Amazon S3 inventory](#) to get a list of all objects in your bucket. For others to access your objects, you'll need to explicitly grant them permissions. [Learn more](#)

< 1 >

<input type="checkbox"/>	Name ▲	Type ▼	Last modified ▼	Size ▼	Storage class ▼
<input type="checkbox"/>	model/	Folder	-	-	-
<input type="checkbox"/>	test/	Folder	-	-	-
<input type="checkbox"/>	train/	Folder	-	-	-

Resource configuration

Instance type	Instance count	Additional storage volume per instance (GB)
ml.m4.xlarge ▼	1	1

Keep alive period
Use [SageMaker Training Managed Warm Pools](#)

Maximum runtime is 1 hour (60 minutes or 3600 seconds).

Encryption key - optional
Encrypt your data. Choose an existing KMS key or enter a key's ARN.


Stopping condition




Maximum runtime

hours ▼

EDA

Hyperparameters

You can use hyperparameters to finely control training. We've set default hyperparameters for the algorithm you've chosen. [Learn more](#) 

Key	Value
feature_dim	
mini_batch_size	1000
epochs	15
predictor_type	
use_bias	true 
binary_classifier_model_selection_criteri.	
target_recall	
target_precision	
num_models	auto

Amazon SageMaker



Getting started

Studio

Studio Lab

Canvas

RStudio

Admin configurations

Domains

Role manager

Images

Lifecycle configurations

SageMaker dashboard

Search

JumpStart

✔ Success! You created a training job.
To track the status of the job, view details.

[View details](#)[Amazon SageMaker](#) > Training jobsTraining jobs [Info](#)

Actions ▾

[Create training job](#)

< 1 >



	Name ▾	Creation time ▾	Duration	Job status ▾	Warm pool status	Time left
<input type="radio"/>	Admission-training-in-University	4/2/2024, 4:37:26 PM	-	InProgress	-	-
<input type="radio"/>	Canvas1708554239450-t1-1-7331dcc27deb47a392bba1b77db440384184af	2/21/2024, 5:24:02 PM	9 minutes	Completed	-	-

Monitor

Access logs for debugging and progress reporting. [Learn more](#)

[View algorithm metrics](#)

[View instance metrics](#)

[View logs](#)

[Search logs](#)

2024-0... > 2024-0...

UTC timezone



CPUUtilization

Percent

230

229

229

20:45

20:51

CPUUtilization

MemoryUtilization

Percent

1

0.755

0.509

20:45

20:51

MemoryUtilization

DiskUtilization

Percent

0.05

0.025

0

20:45

20:51

DiskUtilization

test:objective_loss

No unit

test:mse

No unit

train:objective_loss:final

No unit

EDA

aws

Services

Search

[Alt+S]

Central

Farid_winter2024

CloudWatch

Favorites and recents

Dashboards

Alarms 0 0 0

Logs

Log groups

Log Anomalies

Live Tail

Logs Insights

Metrics

X-Ray traces

Events

Application Signals

Network monitoring

Insights

Settings

[04/02/2024 20:50:08 INFO 139731244078912] #train_score (algo-1) : ('mae', 0.03931092866638355)

▶ 2024-04-02T16:50:08.388-04:00 [04/02/2024 20:50:08 INFO 139731244078912] #quality_metric: host=algo-1, train mse_objective <loss>=0.0030...

▶ 2024-04-02T16:50:08.388-04:00 [04/02/2024 20:50:08 INFO 139731244078912] #quality_metric: host=algo-1, train mse <loss>=0.00300123833556...

▶ 2024-04-02T16:50:08.388-04:00 [04/02/2024 20:50:08 INFO 139731244078912] #quality_metric: host=algo-1, train absolute_loss <loss>=0.0393...

▶ 2024-04-02T16:50:08.388-04:00 [04/02/2024 20:50:08 INFO 139731244078912] #quality_metric: host=algo-1, train rmse <loss>=0.0547835589895...

▶ 2024-04-02T16:50:08.388-04:00 [04/02/2024 20:50:08 INFO 139731244078912] #quality_metric: host=algo-1, train r2 <loss>=0.8398472628684945

▶ 2024-04-02T16:50:08.388-04:00 [04/02/2024 20:50:08 INFO 139731244078912] #quality_metric: host=algo-1, train mae <loss>=0.03931092866638...

▼ 2024-04-02T16:50:08.388-04:00 [04/02/2024 20:50:08 INFO 139731244078912] Best model found for hyperparameters: {"optimizer": "adam", "le...

[04/02/2024 20:50:08 INFO 139731244078912] Best model found for hyperparameters:

{
 "optimizer": "adam",
 "learning_rate": 0.005,
 "l1": 0,
 "wd": 0.0001,
 "lr_scheduler_step": 10,
 "lr_scheduler_factor": 0.99,
 "lr_scheduler_minimum_lr": 0.00001
}

▶ 2024-04-02T16:50:08.388-04:00 [04/02/2024 20:50:08 INFO 139731244078912] Saved checkpoint to "/tmp/tmp89w55wb7/mx-mod-0000.params"

▶ 2024-04-02T16:50:08.388-04:00 #metrics {"StartTime": 1712091008.252668, "EndTime": 1712091008.280849, "Dimensions": {"Algorithm": "XGBoost", "Host": "algo-1", "Operation": "Train", "Status": "Completed"}}

Back to top



Create model

To deploy a model to Amazon SageMaker, first create the model by providing the location of the model artifacts and inference code. See [Deploying a Model on Amazon SageMaker Hosting Services](#) [Learn more about the API](#)

Model settings

Model name

Maximum of 63 alphanumeric characters. Can include hyphens (-), but not spaces. Must be unique within your account in an AWS Region.

IAM role

Amazon SageMaker requires permissions to call other services on your behalf. Choose a role or let us create a role that has the [AmazonSageMakerFullAccess](#) IAM policy attached.

AmazonSageMakerServiceCatalogProductsUseRole ▾

[Create role using the role creation wizard](#)


Container definition 1

▼ Container input options

Amazon SageMaker ×

Getting started

Studio

Studio Lab 

Canvas

RStudio

▼ Admin configurations

Domains

Role manager

Images

Lifecycle configurations

SageMaker dashboard

Search

▼ JumpStart


Foundation models ▾

✔ Success! You created a model.
To use this model, choose Create endpoint.

Create endpoint ×

Amazon SageMaker > Models

Models




Create endpoint

Create endpoint configuration

Actions ▾

Create model

< 1 >



	Name ▾	ARN	Creation time ▾
<input type="radio"/>	admissionmodel	arn:aws:sagemaker:ca-central-1:730335673928:model/admissionmodel	4/2/2024, 5:00:59 PM

Foundation models

Computer vision models

Natural language processing models

▶ Governance

▶ Ground Truth

▶ Notebook

▶ Processing

▶ Training

▶ Inference

▶ Augmented AI

▶ AWS Marketplace

Tutorials

Documentation

admissionmodel

Actions

Create batch transform job

Create endpoint

Model settings

Name	ARN	Creation time	IAM role ARN
admissionmodel	arn:aws:sagemaker:ca-central-1:730335673928:model/admissionmodel	4/2/2024, 5:00:59 PM	arn:aws:iam::730335673928:role/service-role/AmazonSageMaker-ExecutionRole-20240402T162812


Container 1

Container Name	Model data location
Container 1	s3://admissioninuniversity/model/Admission-training-in-University-copy-04-02/output/model.tar.gz
Image	Mode
469771592824.dkr.ecr.ca-central-1.amazonaws.com/linear-learner:1	Single model

Amazon SageMaker ×

Getting started

Studio

Studio Lab 

Canvas

RStudio

▼ Admin configurations

Domains

Role manager

Images

Lifecycle configurations


SageMaker dashboard

Search


▼ JumpStart

Foundation models

✓ Success! You created an endpoint.
To track the status of the endpoint, view details.

View details × 

Amazon SageMaker > Endpoints


Endpoints 

Update endpoint

Actions ▾

Create endpoint

< 1 >

	Name ▾	ARN ▾	Creation time ▾	Status ▾	Last updated ▾
<input type="radio"/>	admissionuniversity	arn:aws:sagemaker:ca-central-1:730335673928:endpoint/admissionuniversity	4/2/2024, 5:05:13 PM	 Creating	4/2/2024, 5:05:14 PM