

CLOUD AND SERVERLESS COMPUTING

PROJECT REPORT

**TITLE: BUILD AND TRAIN A MACHINE LEARNING MODEL
USING AWS SAGEMAKER**

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SECTION-S14

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What is Amazon SageMaker?

Amazon SageMaker is a managed service in the Amazon Web Services ([AWS](#)) public cloud. It provides the tools to build, train and deploy machine learning ([ML](#)) models for predictive analytics applications. The platform automates the tedious work of building a production-ready artificial intelligence (AI) pipeline. Machine learning has a range of uses and benefits. Among them are advanced analytics for customer data and back-end security threat detection.

Deploying ML models is challenging, even for experienced application developers. Amazon SageMaker aims to simplify the process. It uses common algorithms and other tools to accelerate the machine learning process

INTRODUCTION:

Artificial Intelligence & Machine learning is the most exciting and disruptive area in the current era. AI/ML has become an integral part of research and innovations. The main objective of the AI system is to solve real-world problems where businesses are concerned about profitability, sustainability, brand image, cost structure, customers, to compliance. AI products & solutions will be easily adopted by businesses and organizations when this is integrated with the broader enterprise system and consumer-facing organizations can use it to enrich their customers' experiences. This article will mainly deal with the Machine Learning model in AWS Sagemaker. A serverless architecture exposes the runtime inference endpoint available to client software running on consumer devices. REST is a well-architected web-friendly approach and is used to integrate the inference endpoint with the broader enterprise application.

I will describe the step-by-step code and set up to deploy an ML model in AWS Sagemaker using serverless architecture.

The journey from POC to Production is complex and time-consuming. AWS Sagemaker is an advanced Machine Learning platform which is offering a broad range of capabilities to manage large volumes of data to train the model, choose the best algorithm for training it, manage the scalability, capacity of infrastructure while training it, and then deploy & monitor the model into a production environment.

PROBLEM:

When we are talking about the data-driven value of Machine Learning systems, customer analysis is always drawing everyone's interest. There is huge churn in the mobile industry due to the large volume of customers and various service providers.

Mobile operators have historical records on which customers ultimately ended up churning and which continued using the service. Leaving customers effect the business revenue badly. Hence mobile company wants to predict such customers in advance so that they can be retained with additional discount/offers etc.

SOLUTION:

I am going to build a model with the customer dataset. The data is the historical data. The dataset has the customer behaviour, transaction details which will help the algorithm to detect the pattern and predict the customer churn. The operation team will use this data along with other CRM data and try to retain the customer whose churn prediction is high.

APPROACH:

The solution will be implemented using AWS Sagemaker Container from the Notebook instance. Then the endpoint will be invoked by the Lambda function. AWS API Gateway will call the Lambda when the client system will send a POST request with test data to detect the churn or not churn.

ARCHITECTURE:

Client-side applications host a client script that calls an Amazon API Gateway API action and passes parameter values.

API Gateway is a layer that provides the API to the client. API also invokes the backend system with AWS Lambda in a protected private network (VPC).

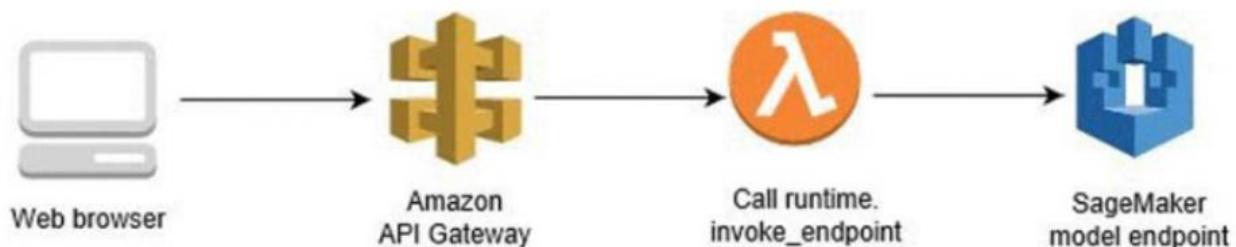
API Gateway passes the parameter values to the Lambda function.

The Lambda function parses the value and sends it to the SageMaker model endpoint.

The model performs the prediction and returns the predicted value to Lambda.

The Lambda function parses the returned value and sends it back to API Gateway. API Gateway responds to the client with that value.

BLOCK DIAGRAM:



OUTPUTS:

Create a Sagemaker Notebook instance

Amazon SageMaker Home cscawsproject - Jupyter Notebook

us-east-1.console.aws.amazon.com/sagemaker/home?region=us-east-1#/notebook-instances/create

AWS Services Search [Alt+S]

N. Virginia LIKHITH KUMAR

Create notebook instance

Amazon SageMaker provides pre-built fully managed notebook instances that run Jupyter notebooks. The notebook instances include example code for common model training and hosting exercises. [Learn more](#)

Notebook instance settings

Notebook instance name: cscproject

Notebook instance type: ml.t2.medium

Elastic Inference: none

Platform identifier: Amazon Linux 2, Jupyter Lab

▶ Additional configuration

Permissions and encryption

IAM role: AmazonSageMakerFullAccess

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Amazon SageMaker Home cscawsproject - Jupyter Notebook

us-east-1.console.aws.amazon.com/sagemaker/home?region=us-east-1#/notebook-instances/create

AWS Services Search [Alt+S]

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▶ Additional configuration

Permissions and encryption

IAM role: AmazonSageMakerFullAccess

Create role using the role creation wizard

Root access: Enable - Give users root access to the notebook
 Disable - Don't give users root access to the notebook

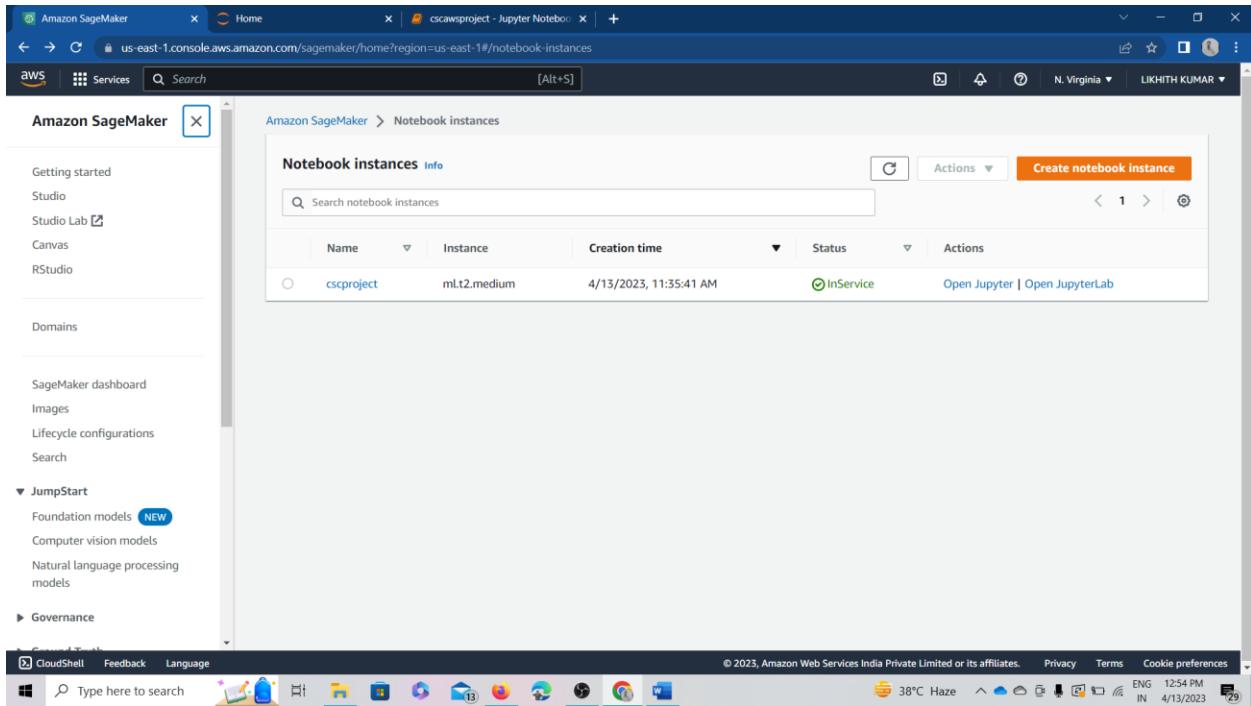
Encryption key: No Custom Encryption

▶ Network - optional

▶ Git repositories - optional

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- NOW OPEN JUPYTER LAB IN THE MANAGEMENT CONSOLE
- CREATE A NEW CONDA_PYTHON3 FILE IN JUPYTER LAB
- THE CURRENT JUPYTERKERNEL DOES NOT HAVE THE SHARP LIBRARY
- INSTALL THE SHARP LIBRARY IN THE CURRENT JUPYTER LAB BY THE FOLLOWING COMMAND: %conda install -c conda-forge shap
- AFTER INSTALLING YOU CAN DOWNLOAD AND EXPLORE THE DATA SETS

The image shows a Windows desktop environment with two browser windows open, both displaying Jupyter Notebooks on an Amazon SageMaker instance.

Top Browser Window:

- Title bar: "Amazon SageMaker" - "Home" - "cscawsproject - Jupyter Notebook" - "+"
- Address bar: "https://cscproject.notebook.us-east-1.sagemaker.aws/notebooks/cscawsproject.ipynb"
- Content area:
 - Jupyter logo and "cscawsproject" text.
 - Last Checkpoint: "a minute ago (unsaved changes)"
 - File menu: File, Edit, View, Insert, Cell, Kernel, Widgets, Help.
 - Toolbar: Back, Forward, Home, Run, Cell, Code, etc.
 - Code cell: In []: `print('hello')`

Bottom Browser Window:

- Title bar: "Amazon SageMaker" - "Home" - "cscawsproject - Jupyter Notebook" - "+"
- Address bar: "https://cscproject.notebook.us-east-1.sagemaker.aws/notebooks/cscawsproject.ipynb"
- Content area:
 - Jupyter logo and "cscawsproject" text.
 - Last Checkpoint: "a minute ago (unsaved changes)"
 - File menu: File, Edit, View, Insert, Cell, Kernel, Widgets, Help.
 - Toolbar: Back, Forward, Home, Run, Cell, Code, etc.
 - Code cell: In [1]: `print('hello')`
Output: hello
 - Code cell: In []:
 - Code cell: In []:

The desktop taskbar at the bottom shows several pinned icons, including File Explorer, Task View, Edge, File History, Mail, Photos, and Google Chrome. The system tray indicates the date and time as 4/13/2023, 11:45 AM, with a battery level of 15%.

The image shows a Windows desktop environment with two Jupyter Notebook windows open side-by-side. Both windows have the title "jupyter cscawsproject Last Checkpoint: a minute ago (unsaved changes)".

Top Window (Left):

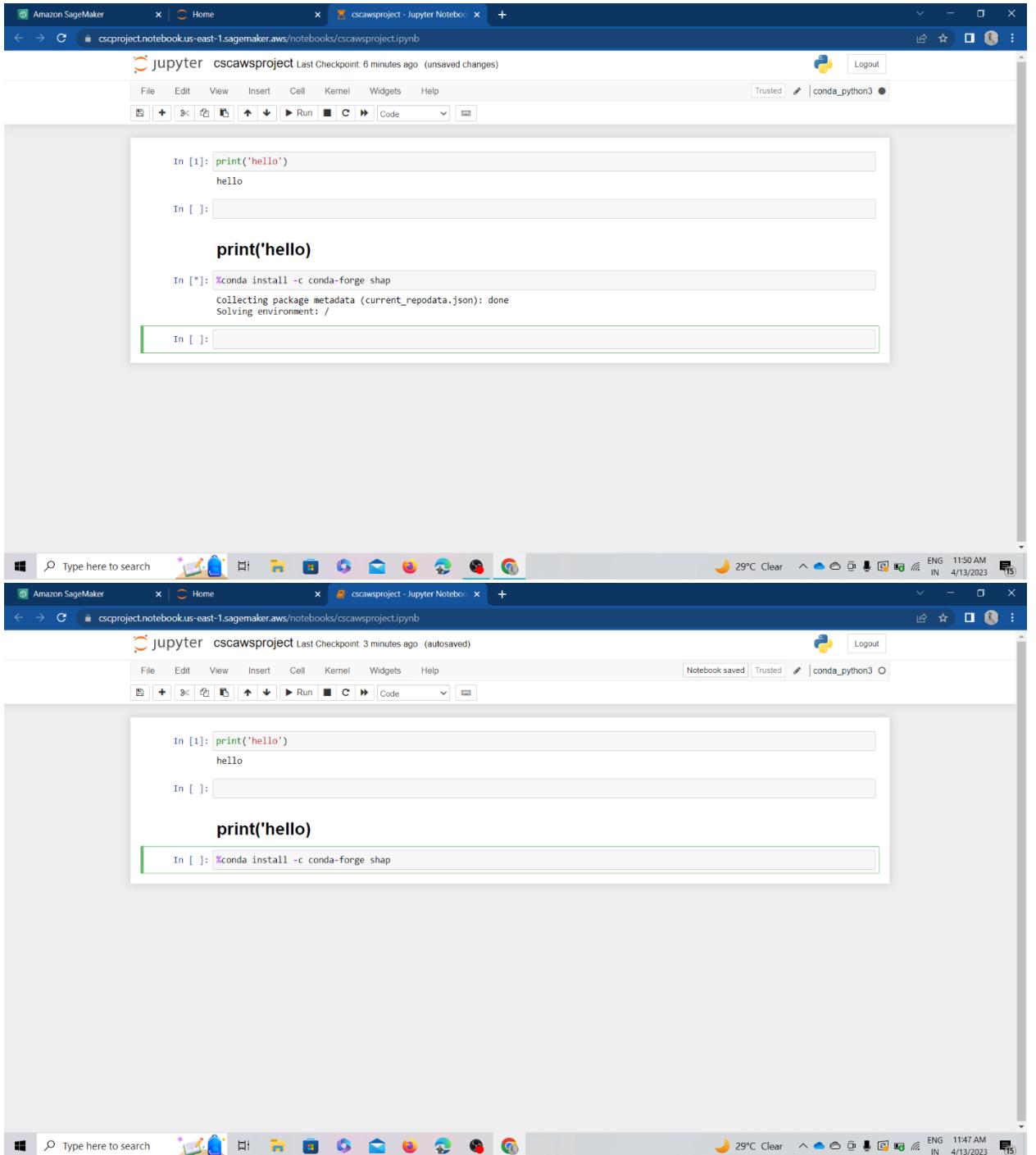
- Cell 1 (In [1]): `print('hello')` - Output: hello
- Cell 2 (In []): `# print('hello')`

Bottom Window (Right):

- Cell 1 (In [1]): `print('hello')` - Output: hello
- Cell 2 (In []): `print('hello')`

The desktop taskbar at the bottom shows various pinned icons and the system tray on the right indicating the date and time as 4/13/2023.

➤ INSTALLING THE SHARP LIBRARIES



The screenshot shows a Jupyter Notebook interface running on Amazon SageMaker. The browser tab is titled "cscawsproject - Jupyter Notebook". The notebook contains the following code:

```
In [1]: print('hello')
hello

In [ ]: print('hello')

In [*]: %conda install -c conda-forge shar
          collecting package metadata (current_repotdata.json): done
          Solving environment: /
```

The command "%conda install -c conda-forge shar" is partially typed in the input cell In []. The status bar at the bottom of the browser window shows the date and time as 4/13/2023.

```
In [ ]: print('hello')

In [*]: conda install -c conda-forge shap

Collecting package metadata (current_repodata.json): done
Solving environment: /
The environment is inconsistent, please check the package plan carefully
The following packages are causing the inconsistency:

- conda-forge/noarch::bleach==5.0.1-pyhd8ed1ab_0
- conda-forge/noarch::pytest==7.2.0-pyhd8ed1ab_2
- conda-forge/noarch::python-lsp-jsonrpc==1.0.0-pyhd8ed1ab_0
- conda-forge/noarch::qtippy==2.3.0-pyhd8ed1ab_0
- conda-forge/linux-64::sip==6.7.5-py310h0d8f1fbef0_0
- conda-forge/noarch::tqdm==4.64.1-pyhd8ed1ab_0
- conda-forge/linux-64::watchdog==2.2.1-py310hff52083_0
- conda-forge/noarch::dask-core==2022.11.0-pyhd8ed1ab_0
- conda-forge/noarch::flask==2.2.2-pyhd8ed1ab_0
- conda-forge/noarch::importlib_metadata==6.0.0-hd8ed1ab_0
- conda-forge/noarch::nltk==3.8.3-pyhd8ed1ab_0
- conda-forge/linux-64::pyqt5-sip==12.11.0-py310hdfbf1fbef0_2
- conda-forge/noarch::python-lsp-server-base==1.7.0-pyhd8ed1ab_0
- conda-forge/noarch::pytoolconfig==1.2.4-pyhd8ed1ab_1
- conda-forge/noarch::qdarkstyle==3.0.3-pyhd8ed1ab_0
- conda-forge/noarch::qtawesome==1.2.2-pyhd8ed1ab_0
- conda-forge/noarch::flask-cors==3.0.10-pyhd8ed1ab_0
- conda-forge/noarch::rope==1.6.0-pyhd8ed1ab_0
- conda-forge/noarch::pylint==2.15.10-pyhd8ed1ab_0
- conda-forge/linux-64::keyring==23.1.1-py310hff52083_0
- conda-forge/noarch::nbformat==5.7.1-pyhd8ed1ab_0
- conda-forge/noarch::python-lsp-server==1.7.0-hd8ed1ab_0
```

➤ CREATING AND EXPLORING THE DATA SET

```
In [2]: conda install -c conda-forge shap

Collecting package metadata (current_repodata.json): done
Solving environment: /
The environment is inconsistent, please check the package plan carefully
The following packages are causing the inconsistency:

- conda-forge/noarch::bleach==5.0.1-pyhd8ed1ab_0
- conda-forge/noarch::pytest==7.2.0-pyhd8ed1ab_2
- conda-forge/noarch::python-lsp-jsonrpc==1.0.0-pyhd8ed1ab_0
- conda-forge/noarch::qtippy==2.3.0-pyhd8ed1ab_0
- conda-forge/linux-64::sip==6.7.5-py310h0d8f1fbef0_0
- conda-forge/noarch::tqdm==4.64.1-pyhd8ed1ab_0
- conda-forge/linux-64::watchdog==2.2.1-py310hff52083_0
- conda-forge/noarch::dask-core==2022.11.0-pyhd8ed1ab_0
- conda-forge/noarch::flask==2.2.2-pyhd8ed1ab_0
- conda-forge/noarch::importlib_metadata==6.0.0-hd8ed1ab_0
- conda-forge/noarch::nltk==3.8.1-pyhd8ed1ab_0
- conda-forge/linux-64::pyqt5-sip==12.11.0-py310hdfbf1fbef0_2
- conda-forge/noarch::python-lsp-server-base==1.7.0-pyhd8ed1ab_0
- conda-forge/noarch::pytoolconfig==1.2.4-pyhd8ed1ab_1

In [3]: import shap

In [4]: import shap

In [5]: import shap
X, y = shap.datasets.adult()
X_display, y_display = shap.datasets.adult(display=True)
feature_names = list(X.columns)
feature_names
```

```
File Edit View Insert Cell Kernel Widgets Help Trusted | conda_python3 ○
```

```
In [3]: import shap
```

```
In [4]: import shap
```

```
In [5]: import shap
X, y = shap.datasets.adult()
X_display, y_display = shap.datasets.adult(display=True)
feature_names = list(X.columns)
feature_names
```

```
Out[5]: ['Age',
 'Workclass',
 'Education-Num',
 'Marital Status',
 'Occupation',
 'Relationship',
 'Race',
 'Sex',
 'Capital Gain',
 'Capital Loss',
 'Hours per week',
 'Country']
```

```
In [ ]:
```

```
In [3]: import shap
```

```
In [4]: import shap
```

```
In [5]: import shap
X, y = shap.datasets.adult()
X_display, y_display = shap.datasets.adult(display=True)
feature_names = list(X.columns)
feature_names
```

```
Out[5]: ['Age',
 'Workclass',
 'Education-Num',
 'Marital Status',
 'Occupation',
 'Relationship',
 'Race',
 'Sex',
 'Capital Gain',
 'Capital Loss',
 'Hours per week',
 'Country']
```

```
In [ ]: display(X.describe())
hist = X.hist(bins=30, sharey=True, figsize=(20, 10))
```

Amazon SageMaker Home cscawsproject - Jupyter Notebook

jupyter cscawsproject Last Checkpoint: 13 minutes ago (unsaved changes)

File Edit View Insert Cell Kernel Widgets Help Trusted conda_python3

```
In [6]: display(X.describe())
hist = X.hist(bins=30, sharey=True, figsize=(20, 10))
```

	Age	Workclass	Education-Num	Marital Status	Occupation	Relationship	Race	Sex	Capital Gain	Capital Loss	Hours per week
count	32561.000000	32561.000000	32561.000000	32561.000000	32561.000000	32561.000000	32561.000000	32561.000000	32561.000000	32561.000000	32561.000000
mean	38.581646	3.668902	10.080679	2.611836	6.572740	2.494518	3.665858	0.669205	1077.648804	87.303833	40.43
std	13.640442	1.455960	2.572562	1.505222	4.228857	1.758232	0.848806	0.470506	7385.911621	403.014771	12.34
min	17.000000	0.000000	1.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	1.00
25%	28.000000	4.000000	9.000000	2.000000	3.000000	0.000000	4.000000	0.000000	0.000000	0.000000	40.00
50%	37.000000	4.000000	10.000000	2.000000	7.000000	3.000000	4.000000	1.000000	0.000000	0.000000	40.00
75%	48.000000	4.000000	12.000000	4.000000	10.000000	4.000000	4.000000	1.000000	0.000000	0.000000	45.00
max	90.000000	8.000000	16.000000	6.000000	14.000000	5.000000	4.000000	1.000000	99999.000000	4356.000000	99.00

Amazon SageMaker Home cscawsproject - Jupyter Notebook

jupyter cscawsproject Last Checkpoint: 13 minutes ago (unsaved changes)

File Edit View Insert Cell Kernel Widgets Help Trusted conda_python3

```
In [ ]:
```

A screenshot of a Jupyter Notebook interface running on Amazon SageMaker. The notebook has three cells:

- In [7]:** Shows a `ValueError` from the `train_test_split` function. The error message indicates that there are too many values to unpack (expected 3).

```
ValueError
Cell In[7], line 2
  1 from sklearn.model_selection import train_test_split
  ----> 2 X_train, X_test, y_test = train_test_split(X, y, test_size=0.2, random_state=1)
      3 X_train_display = X_display.loc[X_train.index]
ValueError: too many values to unpack (expected 3)
```
- In [8]:** Shows the correct import statement for `train_test_split`.

```
In [8]: from sklearn.model_selection import train_test_split
```
- In [9]:** Shows the correct code for splitting the dataset into training, validation, and test sets.

```
In [9]: X_train, X_val, y_train, y_val = train_test_split(X_train, y_train, test_size=0.25, random_state=1)
X_train_display = X_display.loc[X_train.index]
X_val_display = X_display.loc[X_val.index]
```

The code in the current cell (In []:) is `train`.

A screenshot of a Jupyter Notebook interface running on Amazon SageMaker. The notebook has three cells:

- In [7]:** Shows a `ValueError` from the `train_test_split` function. The error message indicates that there are too many values to unpack (expected 3).

```
ValueError
Cell In[7], line 2
  1 from sklearn.model_selection import train_test_split
  ----> 2 X_train, X_test, y_test = train_test_split(X, y, test_size=0.2, random_state=1)
      3 X_train_display = X_display.loc[X_train.index]
ValueError: too many values to unpack (expected 3)
```
- In [8]:** Shows the correct import statement for `train_test_split`.

```
In [8]: from sklearn.model_selection import train_test_split
```
- In [9]:** Shows the correct code for splitting the dataset into training, validation, and test sets.

```
In [9]: X_train, X_val, y_train, y_val = train_test_split(X_train, y_train, test_size=0.25, random_state=1)
X_train_display = X_display.loc[X_train.index]
X_val_display = X_display.loc[X_val.index]
```

The code in the current cell (In []:) is `train`.

Amazon SageMaker

cscawsproject - Jupyter Notebook

jupyter cscawsproject Last Checkpoint: 25 minutes ago (unsaved changes)

In [11]: train

Out[11]:

	Income>50K	Age	Workclass	Education-Num	Marital Status	Occupation	Relationship	Race	Sex	Capital Gain	Capital Loss	Hours per week	Country
10911	1	47.0	4	9.0	2	3	4	4	1	0.0	0.0	40.0	39
17852	0	31.0	4	13.0	2	7	4	3	1	0.0	0.0	36.0	26
29165	1	32.0	4	10.0	2	13	5	4	0	0.0	0.0	32.0	39
30287	0	58.0	4	9.0	2	3	4	2	1	0.0	0.0	40.0	39
24019	0	17.0	4	6.0	4	6	3	4	1	0.0	0.0	20.0	39
...
21168	0	43.0	4	8.0	2	14	4	4	1	0.0	0.0	40.0	39
6452	0	26.0	4	9.0	4	7	0	4	1	0.0	0.0	52.0	39
31352	0	32.0	7	14.0	2	10	4	4	1	0.0	0.0	50.0	39
6575	0	45.0	4	9.0	4	6	0	4	1	0.0	0.0	40.0	39
23608	0	23.0	4	9.0	4	1	1	4	0	0.0	0.0	40.0	39

19536 rows × 13 columns

In []:

Amazon SageMaker

cscawsproject - Jupyter Notebook

jupyter cscawsproject Last Checkpoint: 25 minutes ago (unsaved changes)

In [11]: train

Out[11]:

	Income>50K	Age	Workclass	Education-Num	Marital Status	Occupation	Relationship	Race	Sex	Capital Gain	Capital Loss	Hours per week	Country
10911	1	47.0	4	9.0	2	3	4	4	1	0.0	0.0	40.0	39
17852	0	31.0	4	13.0	2	7	4	3	1	0.0	0.0	36.0	26
29165	1	32.0	4	10.0	2	13	5	4	0	0.0	0.0	32.0	39
30287	0	58.0	4	9.0	2	3	4	2	1	0.0	0.0	40.0	39
24019	0	17.0	4	6.0	4	6	3	4	1	0.0	0.0	20.0	39
...
21168	0	43.0	4	8.0	2	14	4	4	1	0.0	0.0	40.0	39
6452	0	26.0	4	9.0	4	7	0	4	1	0.0	0.0	52.0	39
31352	0	32.0	7	14.0	2	10	4	4	1	0.0	0.0	50.0	39
6575	0	45.0	4	9.0	4	6	0	4	1	0.0	0.0	40.0	39
23608	0	23.0	4	9.0	4	1	1	4	0	0.0	0.0	40.0	39

19536 rows × 13 columns

In []: validation

Amazon SageMaker Home cscawsproject - Jupyter Notebook

jupyter cscawsproject Last Checkpoint: 25 minutes ago (unsaved changes)

In [12]: validation

Out[12]:

	Income>50K	Age	Workclass	Education-Num	Marital Status	Occupation	Relationship	Race	Sex	Capital Gain	Capital Loss	Hours per week	Country
16530	0	25.0	4	4.0	2	6	4	4	1	0.0	0.0	40.0	26
26723	0	41.0	6	9.0	2	5	5	4	0	0.0	0.0	40.0	39
3338	0	79.0	0	9.0	6	0	0	2	0	0.0	0.0	30.0	39
19367	1	43.0	2	15.0	2	10	4	4	1	15024.0	0.0	45.0	39
30274	0	51.0	5	9.0	4	12	2	4	1	0.0	0.0	40.0	0
...
1604	0	46.0	7	9.0	2	13	4	4	1	0.0	0.0	40.0	39
6937	1	71.0	4	10.0	6	12	0	4	1	0.0	0.0	35.0	39
11034	0	36.0	4	9.0	5	14	2	4	1	0.0	0.0	60.0	26
2819	0	31.0	4	9.0	4	8	0	4	0	0.0	0.0	40.0	39
14152	1	37.0	4	10.0	2	12	4	4	1	0.0	0.0	50.0	11

6512 rows × 13 columns

In []:

Amazon SageMaker Home cscawsproject - Jupyter Notebook

jupyter cscawsproject Last Checkpoint: 25 minutes ago (unsaved changes)

In [12]: validation

Out[12]:

	Income>50K	Age	Workclass	Education-Num	Marital Status	Occupation	Relationship	Race	Sex	Capital Gain	Capital Loss	Hours per week	Country
16530	0	25.0	4	4.0	2	6	4	4	1	0.0	0.0	40.0	26
26723	0	41.0	6	9.0	2	5	5	4	0	0.0	0.0	40.0	39
3338	0	79.0	0	9.0	6	0	0	2	0	0.0	0.0	30.0	39
19367	1	43.0	2	15.0	2	10	4	4	1	15024.0	0.0	45.0	39
30274	0	51.0	5	9.0	4	12	2	4	1	0.0	0.0	40.0	0
...
1604	0	46.0	7	9.0	2	13	4	4	1	0.0	0.0	40.0	39
6937	1	71.0	4	10.0	6	12	0	4	1	0.0	0.0	35.0	39
11034	0	36.0	4	9.0	5	14	2	4	1	0.0	0.0	60.0	26
2819	0	31.0	4	9.0	4	8	0	4	0	0.0	0.0	40.0	39
14152	1	37.0	4	10.0	2	12	4	4	1	0.0	0.0	50.0	11

6512 rows × 13 columns

In []: test

Amazon SageMaker

cscawsproject - Jupyter Notebook

jupyter cscawsproject Last Checkpoint: 25 minutes ago (unsaved changes)

In [13]: test

Out[13]:

	Income>50K	Age	Workclass	Education-Num	Marital Status	Occupation	Relationship	Race	Sex	Capital Gain	Capital Loss	Hours per week	Country
9646	0	62.0	6	4.0	6	8	0	4	0	0.0	0.0	66.0	39
709	0	18.0	4	7.0	4	8	2	4	1	0.0	0.0	25.0	39
7385	1	25.0	4	13.0	4	5	3	4	1	27828.0	0.0	50.0	39
16671	0	33.0	4	9.0	2	10	4	4	1	0.0	0.0	40.0	39
21932	0	36.0	4	7.0	4	7	1	4	0	0.0	0.0	40.0	39
...
5889	1	38.0	4	13.0	2	10	5	4	0	0.0	0.0	20.0	39
28723	0	17.0	4	6.0	4	12	3	4	0	0.0	0.0	20.0	39
29514	0	35.0	4	9.0	4	14	3	4	1	0.0	0.0	40.0	39
1600	0	30.0	4	7.0	2	3	4	4	1	0.0	0.0	45.0	39
639	1	52.0	6	16.0	2	10	4	4	1	0.0	0.0	60.0	39

6513 rows x 13 columns

In []:

Type here to search

Amazon SageMaker

cscawsproject - Jupyter Notebook

jupyter cscawsproject Last Checkpoint: 35 minutes ago (unsaved changes)

In [14]: traini.to_csv ('train.csv', index=False, header=False)
validation.to_csv('validation.csv', index=False, header=False)

```
NameError: name 'traini' is not defined
Cell In[14], line 1
----> 1 traini.to_csv ('train.csv', index=False, header=False)
      2 validation.to_csv('validation.csv', index=False, header=False)

NameError: name 'traini' is not defined
```

In [15]: train.to_csv ('train.csv', index=False, header=False)
validation.to_csv('validation.csv', index=False, header=False)

In [16]: import sagemaker, boto3, os
bucket = sagemaker.Session().default_bucket()
prefix = "demo-sagemaker-xgboost-adult-income-prediction"

boto3.Session().resource('s3').Bucket(bucket).Object(
 os.path.join(prefix, "data/train.csv")).upload_file('train.csv')
boto3.Session().resource('s3').Bucket(bucket).Object(
 os.path.join(prefix, "data/validation.csv")).upload_file('validation.csv')

In []: ! aws s3 ls {bucket}/{prefix}/data --recursive

The screenshot shows a Jupyter Notebook interface running on Amazon SageMaker. The notebook has three cells:

- In [14]:** `traini.to_csv('train.csv', index=False, header=False)`
A `NameError` occurs because 'traini' is not defined.
- In [15]:** `train.to_csv('train.csv', index=False, header=False)`
A `validation.to_csv('validation.csv', index=False, header=False)` command is present but not run.
- In [16]:** `import sagemaker, boto3, os`
The code uses `boto3.Session().resource('s3').Bucket(bucket).Object(os.path.join(prefix, 'data/train.csv')).upload_file('train.csv')` to upload 'train.csv' to an S3 bucket.

Cell In [17] contains the command `aws s3 ls {bucket}/{prefix}/data --recursive`, which lists files in the S3 bucket:

```
2023-04-13 06:49:15    786285 demo-sagemaker-xgboost-adult-income-prediction/data/train.csv  
2023-04-13 06:49:15    262122 demo-sagemaker-xgboost-adult-income-prediction/data/validation.csv
```

The status bar at the bottom shows the date and time as 4/13/2023 and 12:20 PM, and the system temperature as 36°C Haze.

➤ TRAIN THE MODEL

The screenshot shows a Jupyter Notebook interface running on Amazon SageMaker. The notebook has four cells:

- In [17]:** `aws s3 ls {bucket}/{prefix}/data --recursive`
This cell lists the same files as the previous screenshot.
- In [18]:** `TRAINING THE MODEL`
A `SyntaxError: invalid syntax` occurs because the code is just text and not valid Python.
- In [19]:** `import sagemaker4`
A `ModuleNotFoundError` occurs because there is no module named 'sagemaker4'.
- In []:** `import sagemaker`
The code defines `region = sagemaker.Session().boto_region_name` and prints the AWS Region.
It also defines `role = sagemaker.Session().boto_role_name` and prints the Role ARN.

The status bar at the bottom shows the date and time as 4/13/2023 and 12:23 PM, and the system temperature as 36°C Haze.

In [21]:

```
import sagemaker
region = sagemaker.Session().boto_region_name
print("AWS Region: {}".format(region))

role = sagemaker.Session().get_execution_role()
print("Role:Arn {}".format(role))
```

AWS Region: us-east-1

AttributeError: Traceback (most recent call last)
Cell In[21], line 6
3 region = sagemaker.Session().boto_region_name
4 print("AWS Region: {}".format(region))
--> 6 role = sagemaker.Session().get_execution_role()
7 print("Role:Arn {}".format(role))

AttributeError: 'Session' object has no attribute 'get_execution_role'

In [21]:

```
import sagemaker
region = sagemaker.Session().boto_region_name
print("AWS Region: {}".format(region))

role = sagemaker.Session().get_execution_role()
print("Role:Arn {}".format(role))
```

AWS Region: us-east-1

AttributeError: Traceback (most recent call last)
Cell In[21], line 6
3 region = sagemaker.Session().boto_region_name
4 print("AWS Region: {}".format(region))
--> 6 role = sagemaker.Session().get_execution_role()
7 print("Role:Arn {}".format(role))

AttributeError: 'Session' object has no attribute 'get_execution_role'

In [22]:

```
import sagemaker
region = sagemaker.Session().boto_region_name
print("AWS Region: {}".format(region))

role = sagemaker.get_execution_role()
print("Role:Arn {}".format(role))
```

AWS Region: us-east-1
Role:Arn arn:aws:iam::008321711134:role/service-role/AmazonSageMaker-ExecutionRole-20230413T113412

```
region = sagemaker.Session().boto_region_name
print("AWS Region: {}".format(region))

role = sagemaker.Session().get_execution_role()
print("Role:Arn {}".format(role))

AWS Region: us-east-1

AttributeError: 'Session' object has no attribute 'get_execution_role'

In [22]: import sagemaker

region = sagemaker.Session().boto_region_name
print("AWS Region: {}".format(region))

role = sagemaker.Session().get_execution_role()
print("Role:Arn {}".format(role))

AWS Region: us-east-1
Role:Arn arn:aws:iam::00832171134:role/service-role/AmazonSageMaker-ExecutionRole-20230413T113412
```

In []: sagemaker.__version__

```
AWS Region: us-east-1

AttributeError: 'Session' object has no attribute 'get_execution_role'

In [22]: import sagemaker

region = sagemaker.Session().boto_region_name
print("AWS Region: {}".format(region))

role = sagemaker.Session().get_execution_role()
print("Role:Arn {}".format(role))

AWS Region: us-east-1
Role:Arn arn:aws:iam::00832171134:role/service-role/AmazonSageMaker-ExecutionRole-20230413T113412

In [23]: sagemaker.__version__

Out[23]: '2.132.0'

In [ ]:
```

In []:

YOUTUBE LINK: <https://youtu.be/BDnE4w0o2yI>

PPT LINK:

https://docs.google.com/presentation/d/17kAHWF_8hFOqOz8gazSSS9PygHryOHvb/edit?usp=sharing&oid=109261200163574191129&rtpof=true&sd=true

LINKED IN: <https://www.linkedin.com/pulse/build-train-machine-learning-model-amazon-sage-maker-likhith-kumar>