Deploying ML Models on AWS platform

Agenda

- Train a model locally. (Local environment should be ready).
- Upload the model to AWS S3 bucket.
- Create a lambda function.
- Integrate with API gateway.
- Optional (discussion on layers)
- Optional (discussion on how to productionize DNN models on Lambda)

INTRODUCTIONS

A Request

- This will be a hands-on session
- Please do not ask code for copy paste. This exercise to make you implement a ml model, end to end, copy paste will not serve the purpose.
- Please remain actively engaged during workshop and follow facilitator’s guidance·
- Please keep your video on for the entire session duration
- Feedback link available below must be filled. That will be also the record of your participation in this session.
- Do ask for breaks.

Features of AWS Lambda

- For hosting small models
- Serverless
- Suitable for Low-cost scenario, experimentation. When you need to have a end to end demo.
- No GPU
- Deep Neural Networks of reasonable size can be productionized.
- Suitable for domains like Manufacturing.
- No heavy infrastructure, less maintenance.
- ML Pipeline can be established (retraining) – Can address some of the ML lifecycle challenges.

Lambda Limits

Resource	Quota
Function memory allocation	128 MB to 10,240 MB, in 1-MB increments.
Function timeout	900 seconds (15 minutes)
Function environment variables	4 KB
Function resource-based policy	20 KB
Function layers	five layers
Function burst concurrency	500 - 3000 (varies per Region)
Invocation payload (request and response)	6 MB (synchronous) 256 KB (asynchronous)
Deployment package (.zip file archive) size	50 MB (zipped, for direct upload)

Resource	Quota
	256 KB (asynchronous)
Deployment package (.zip file archive) size	50 MB (zipped, for direct upload) 250 MB (unzipped, including layers) 3 MB (console editor), 512 KB maximum for an individual file
Container image code package size	10 GB
Test events (console editor)	10
/tmp directory storage	512 MB
File descriptors	1,024
Execution processes/threads	1,024

Part – A - Train the model (if we do not have)

- Use Wine dataset for training a model.
- Save the model to disk.
- Load the model from the disk.
- Do prediction on sample data.

Train the model

```
from sklearn.datasets import load_wine
import pandas as pd
import numpy as np
import pickle

data = load_wine() # import dataset
df = pd.DataFrame(data['data'], columns=data['feature_names']) # build dataframe
df['target'] = data['target'] # add dependent variable
df = df.sample(frac=1) # randomize the data
df.head(3)

train_df = df[:150]
test_df = df[150:]

feature_cols = ['alcohol', 'malic_acid']
X = train_df[feature_cols] # Features
y = train_df.target # Target variable

from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.25,random_state=0)
```

Train the model

```
from sklearn.linear_model import LogisticRegression

# instantiate the model (using the default parameters)
logreg = LogisticRegression()

# fit the model with data
logreg.fit(X_train,y_train)
y_pred=logreg.predict(X_test)
```

Testing (Predicting) for one observation:

```
data = {
    "alcohol": 11.1, "malic_acid": 5
}
df = pd.DataFrame([data])
result = logreg.predict(df)
result[0]
```

Creating Model file

```
import pickle
pickle.dump( logreg, open( "wine-model-masterclass-2-input-params.p", "wb" ))

# change the name so that it is unique to you <model-wine-ritesh.p>
```

Testing the model from disk at Training time:

- This is an important step.
- This is setting up the code which can be used at the time of setting this up.
- if you have done this step well, rest of things will be easy.

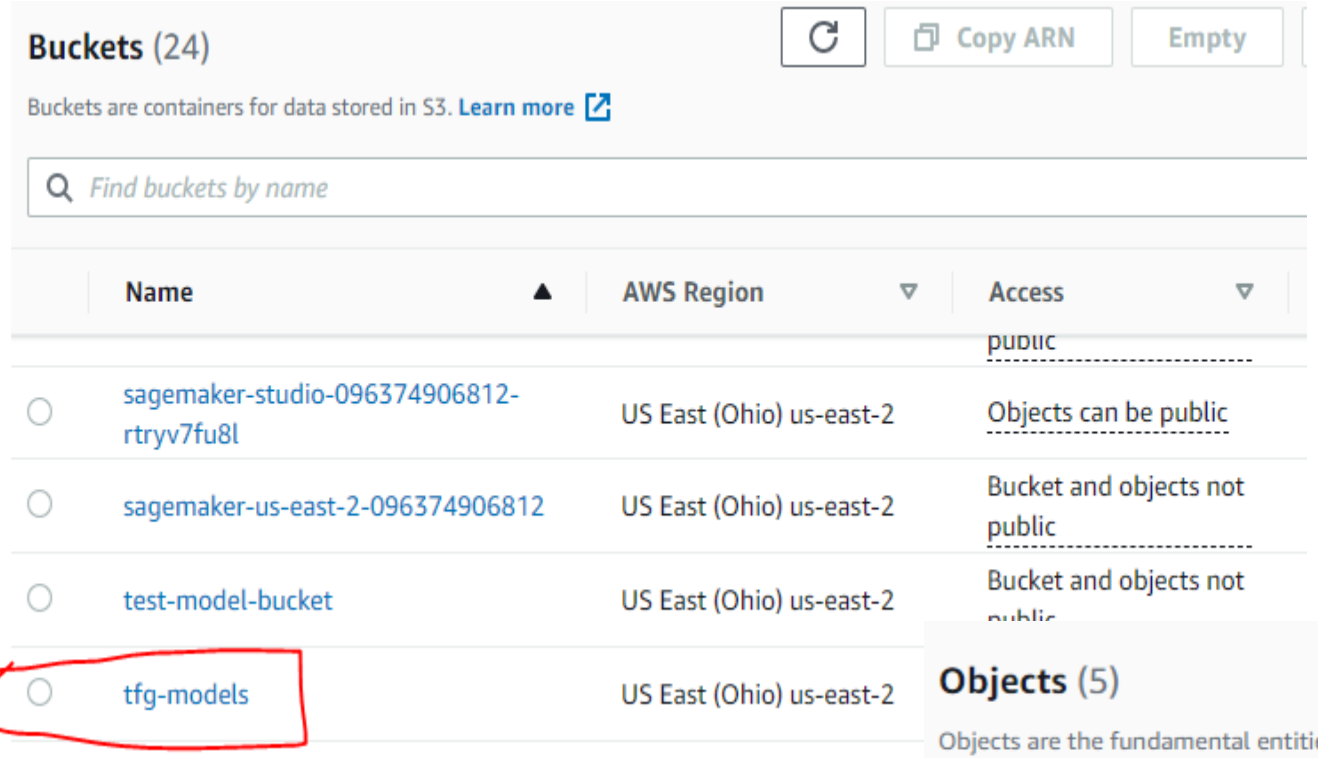
- Load_fn
- Input_fn
- Preprocessing_fn
- Predict_fn
- Output_fn

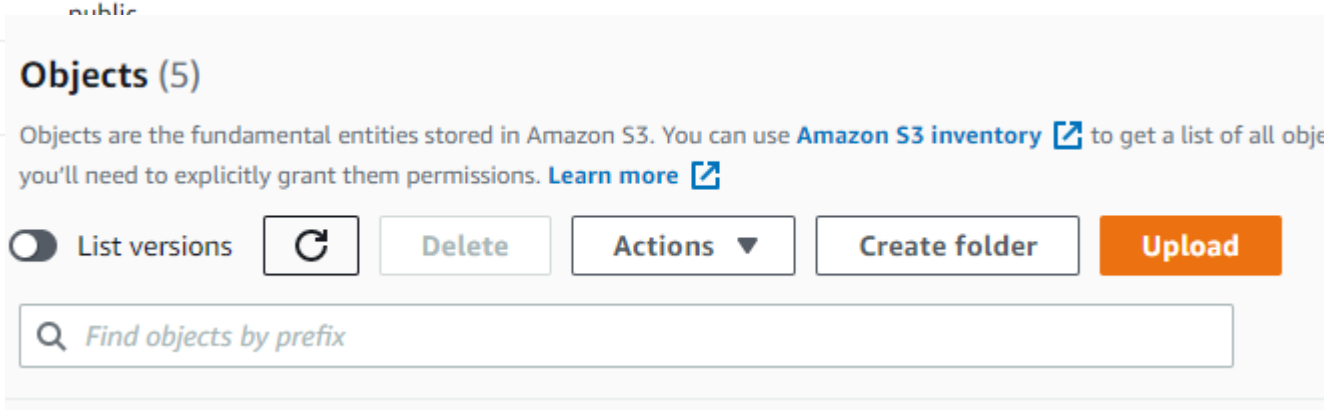
```
temp_file_path = "wine-model-masterclass-2-input-params.p"
with open(temp_file_path, 'rb') as f:
    model = pickle.load(f)

input_ = {"alcohol": 105, "malic_acid": 13}
df_input = pd.DataFrame([input_])
df_input
model.predict(df_input)
```

PART B - Build the Lambda function

- Putting the model on S3 bucket
- Add Layers (Prebuilt)
- Add Environment Variables
- Building Lambda function (Step by Step)

Upload model to AWS





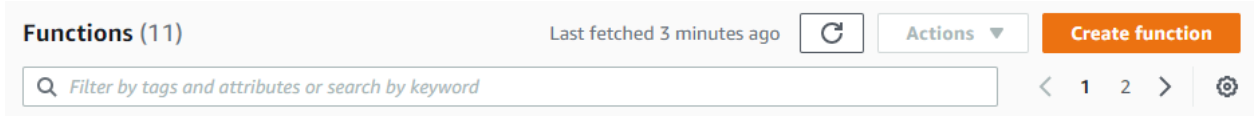
- Login to AWS account with your credentials.
- Goto S3 Services
- Locate **tfg-models**
- **Upload the model file created above.**

Building lambda function

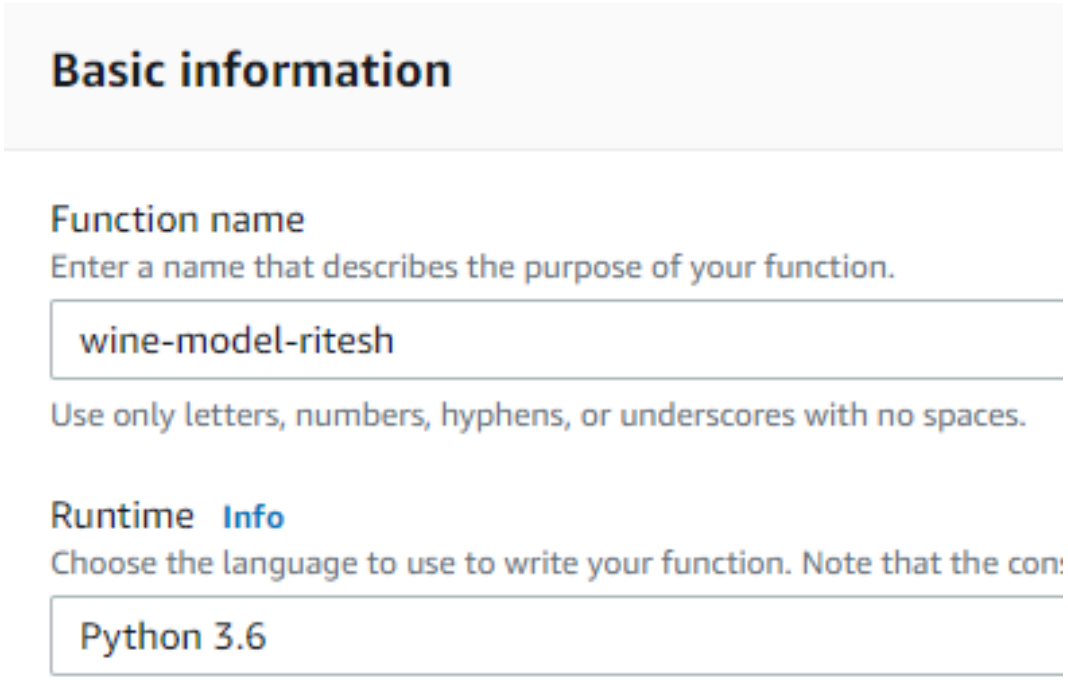
Go to Lambda services

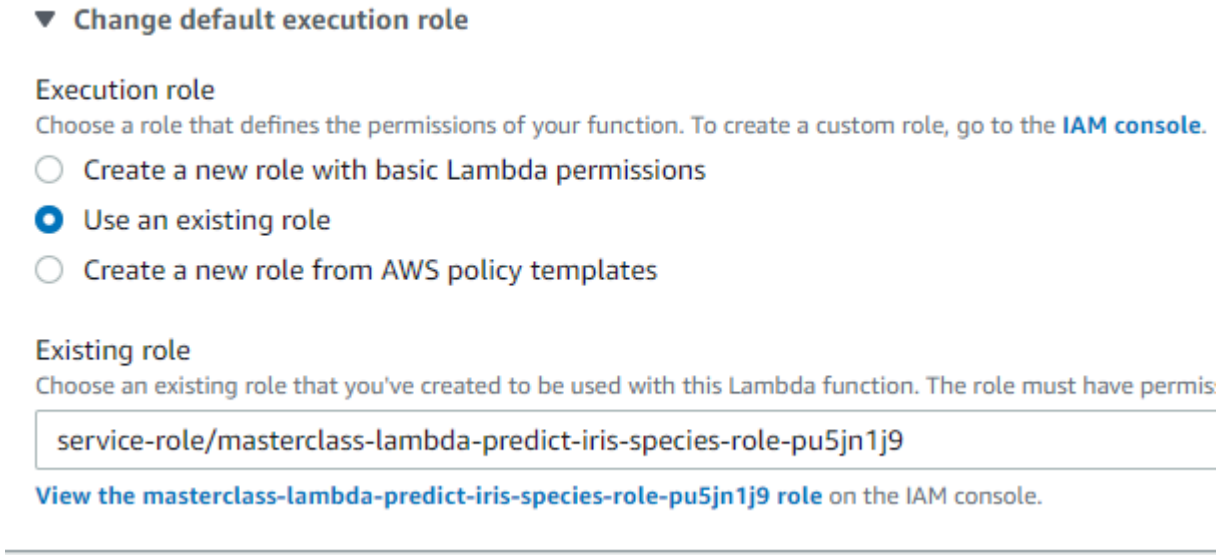
https://us-east-2.console.aws.amazon.com/lambda/home?region=us-east-2#/functions

Create Function



Building lambda function

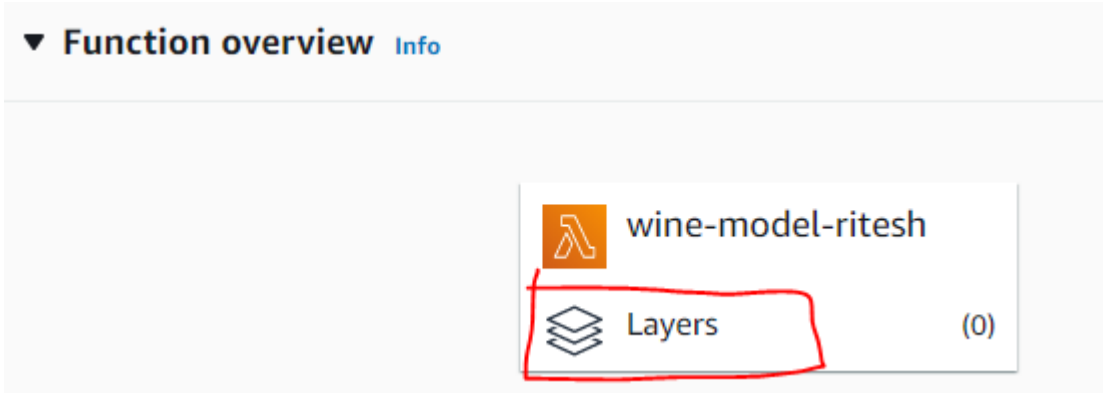




Advanced Settings - Leave as is

Adding Layers - Building lambda function

Layers provide supporting libraries.

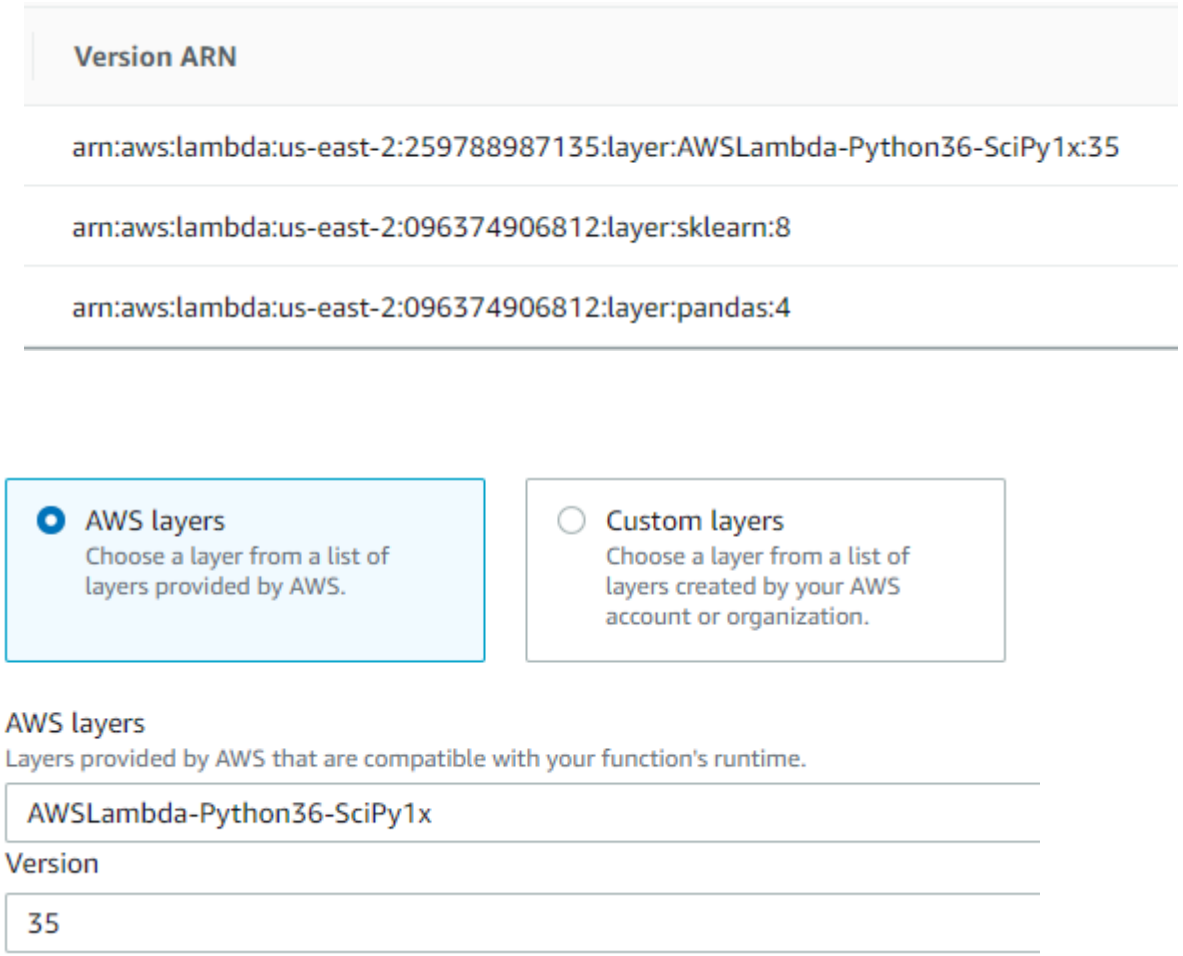


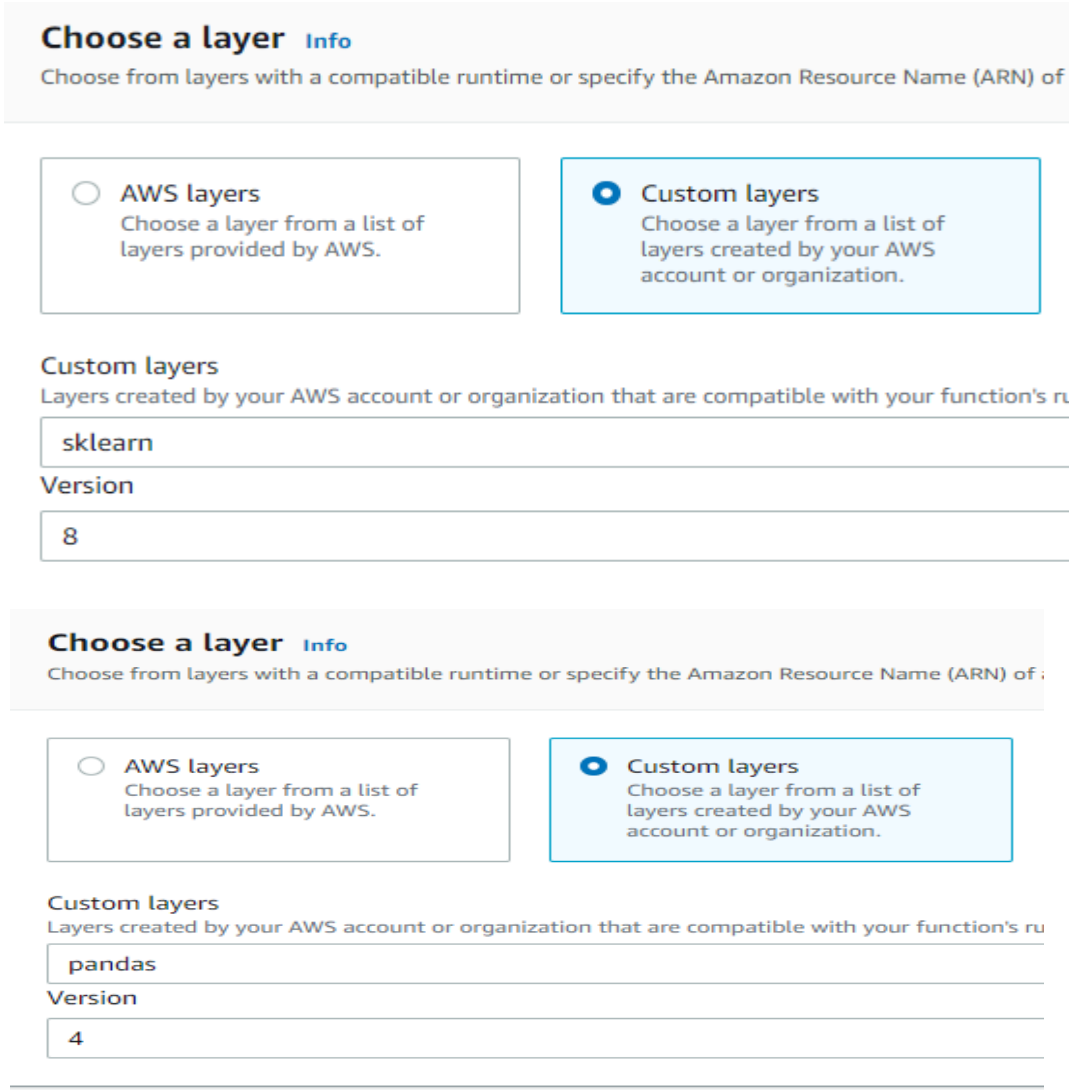
Add following layers:

Layers Info

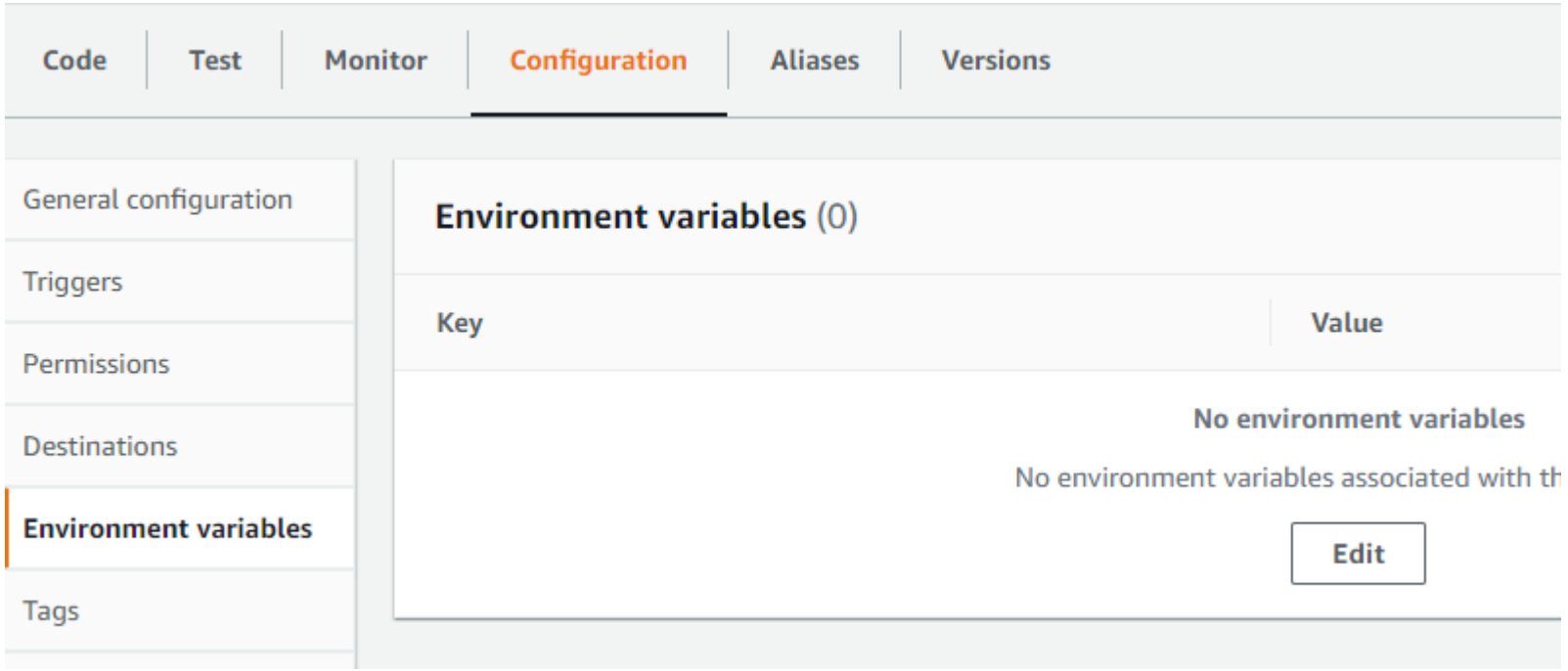
Merge order	Name	Layer version
1	AWSLambda-Python36-SciPy1x	35
2	sklearn	8
3	pandas	4

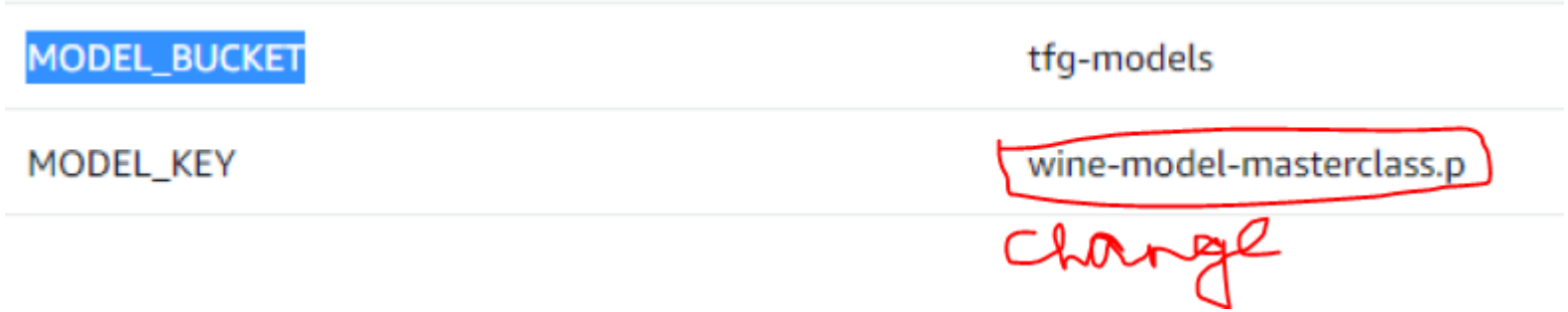
Building lambda function





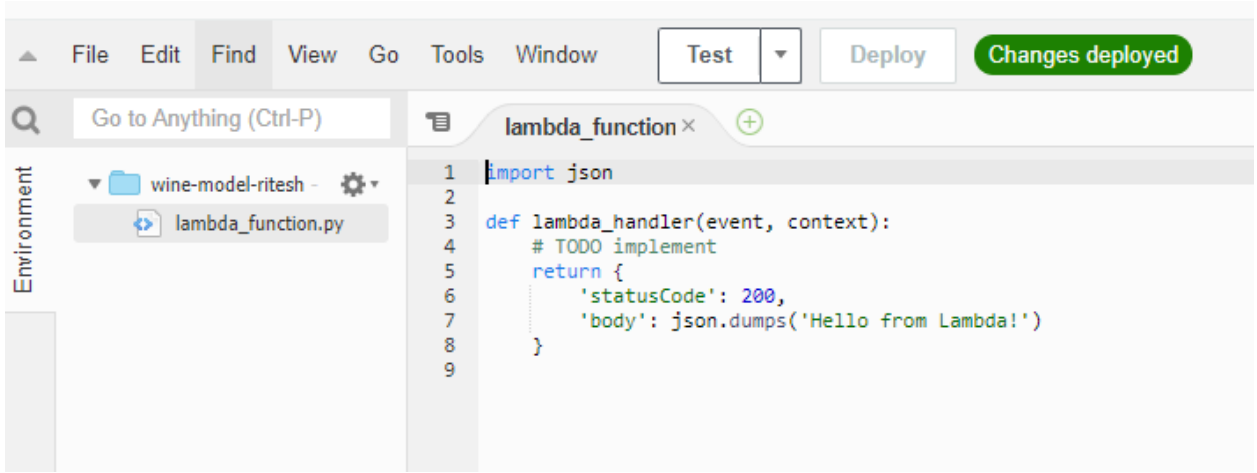
Define environment variables





Building lambda function

IMPLEMENT CODE: click on lambda_function.py



Building lambda function

Step 1: Test that various parameters and libraries are being imported properly:

```
import json
import sklearn
import boto3
import os
import json
import pickle
import pandas as pd
```

```
def lambda_handler(event, context):
    message = f'{MODEL_BUCKET}-{MODEL_KEY}-{sklearn.__version__}'
    # TODO implement
    return {
        'statusCode': 200,
        'body': json.dumps(message)
    }
```

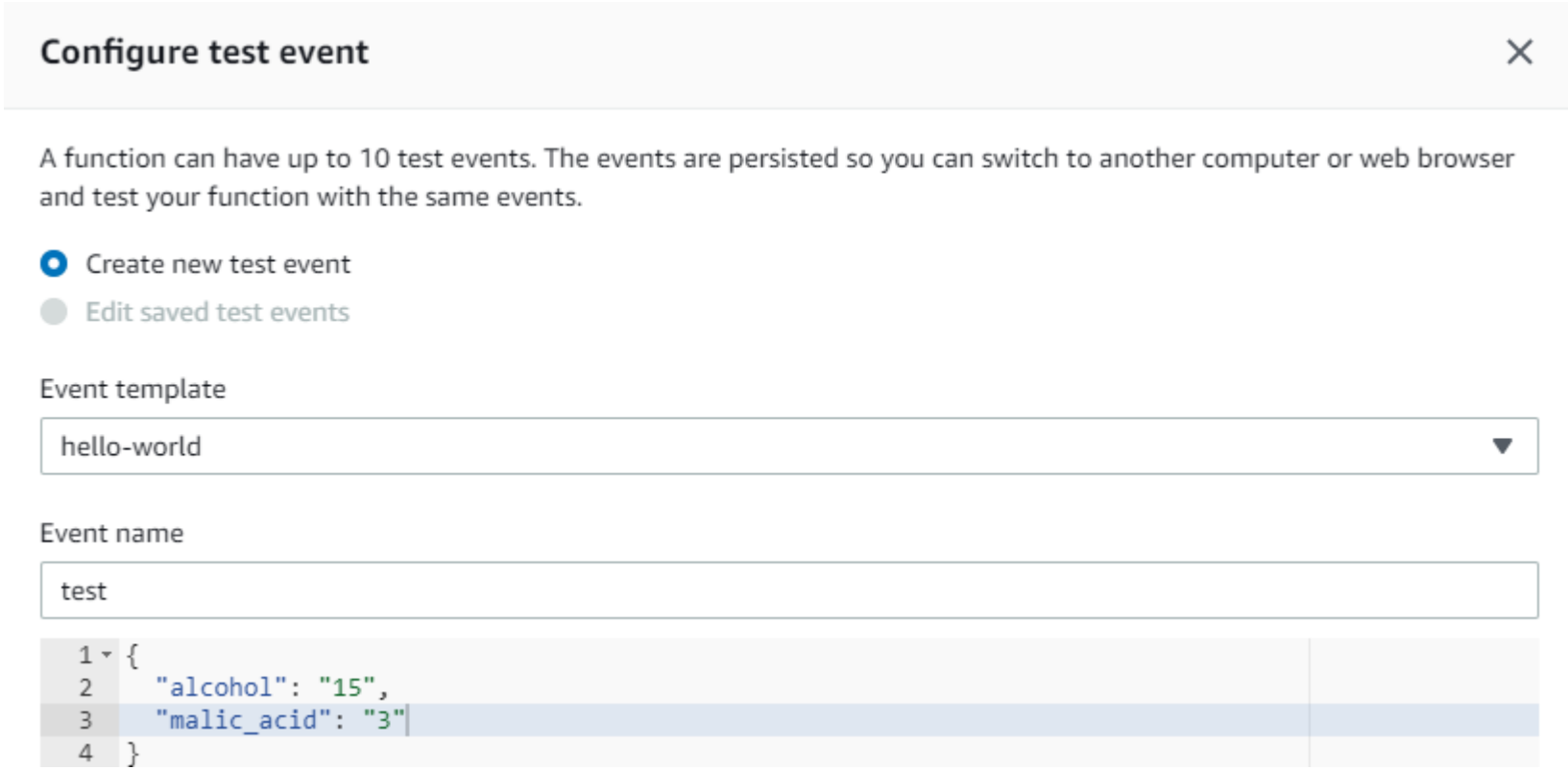
```
s3 = boto3.client('s3')
MODEL_BUCKET=os.environ.get('MODEL_BUCKET')
# get bucket prefix from ENV variable
MODEL_KEY=os.environ.get('MODEL_KEY')
```

OUTPUT:

Should be something like following:

```
{
  "statusCode": 200,
  "body": "\"tfg-models-wine-model-ritesh.p-0.24.1\""
}
```

Configure Test Event



Loading the model - Building lambda function

MAKE THE FOLLOWING CHANGES TO LOAD THE MODEL

```
s3 = boto3.client('s3')

MODEL_BUCKET=os.environ.get('MODEL_BUCKET')

# get bucket prefix from ENV variable
MODEL_KEY=os.environ.get('MODEL_KEY')

temp_file_path = '/tmp/' + MODEL_KEY

# ** Model Init **
s3.download_file(MODEL_BUCKET, MODEL_KEY, temp_file_path)

print(temp_file_path)

with open(temp_file_path, 'rb') as f:
    model = pickle.load(f)    strclasses = str(model.classes_)
```

```
def lambda_handler(event, context):
    message = f'{MODEL_BUCKET}-{MODEL_KEY}-{strclasses}'
    return {
        'statusCode': 200,
        'body': json.dumps(message)
    }
```

OUTPUT:

Response

```
{
  "statusCode": 200,
  "body": "\"tfg-models-wine-model-ritesh.p-[0 1 2]\""
}
```

Prediction - Building lambda function

PREDICTION - Capture Input and Call Predict Function

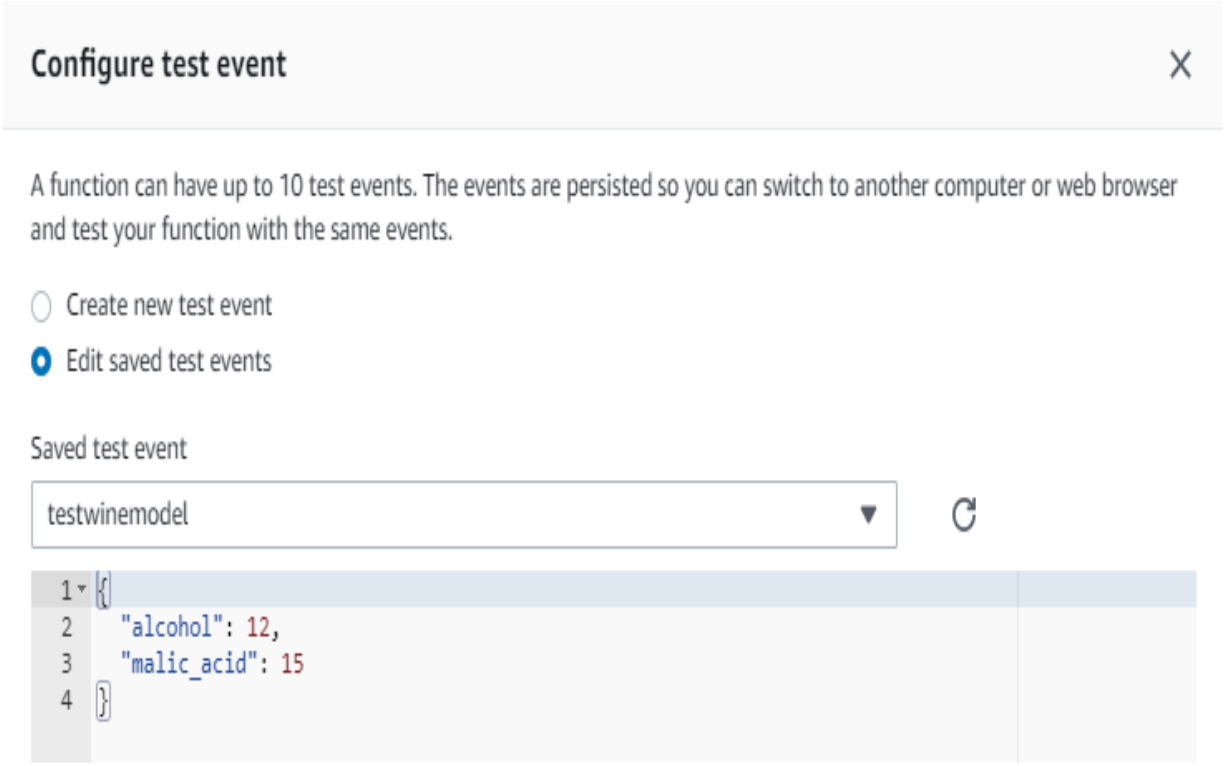
Inside the lambda handler module:

```
input_ = event
y_pred=model.predict(pd.DataFrame([input_]))
message = f'{MODEL_BUCKET}-{MODEL_KEY}-{y_pred}'
```

Output:

Response

```
{
  "statusCode": 200,
  "body": "\"tfg-models-wine-model-ritesh.p-[2]\""
}
```



Building lambda function – Exercise to Participants

Change the lambda function to calculate the time taken in prediction

HINT: start_time = time.time()

Attach this calculated time taken to the output

Building lambda function – Exercise to Participants

SOLUTION

Building lambda function – Exercise to Participants

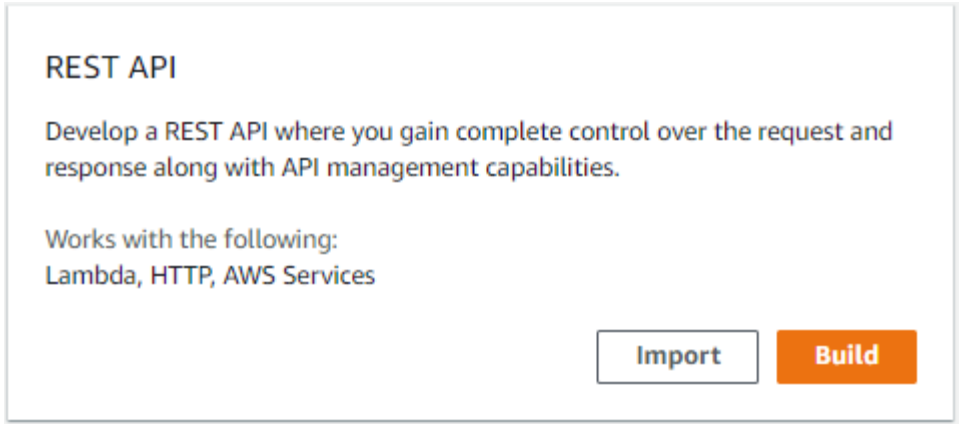
lambda_function × Execution results ×

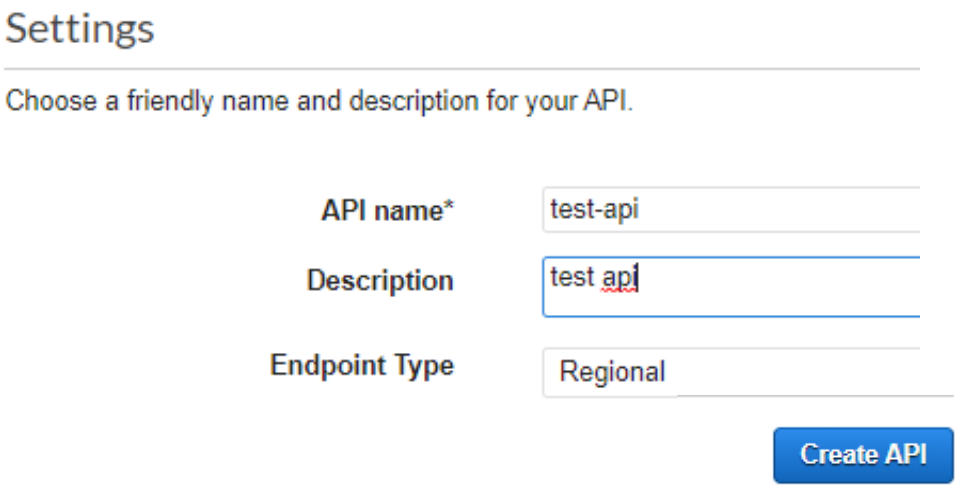
```
31
32  def lambda_handler(event, context):
33      start_time = time.time()
34      input_ = event
35      #print(input_)
36      #df_input = pd.DataFrame([input_])
37      y_pred=model.predict(pd.DataFrame([input_]))
38      message = f'{MODEL_BUCKET}-{MODEL_KEY}-{y_pred}'
39      end_time = time.time()
40      time_taken = end_time - start_time
41      message = f'time taken is: {time_taken}'
42      # TODO implement
43      return {
44          'statusCode': 200,
45          'body': json.dumps(message),
46          'time': time_taken,
47          'predict':f'class_{y_pred}'
48      }
49
```

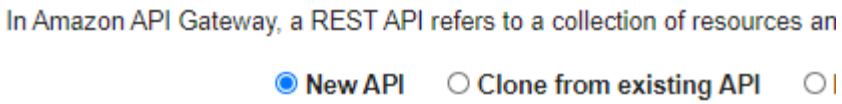
Integrate with API gateway

- Create API
- Create Resource
- Create Method
- Test on Platform
- Test with Curl

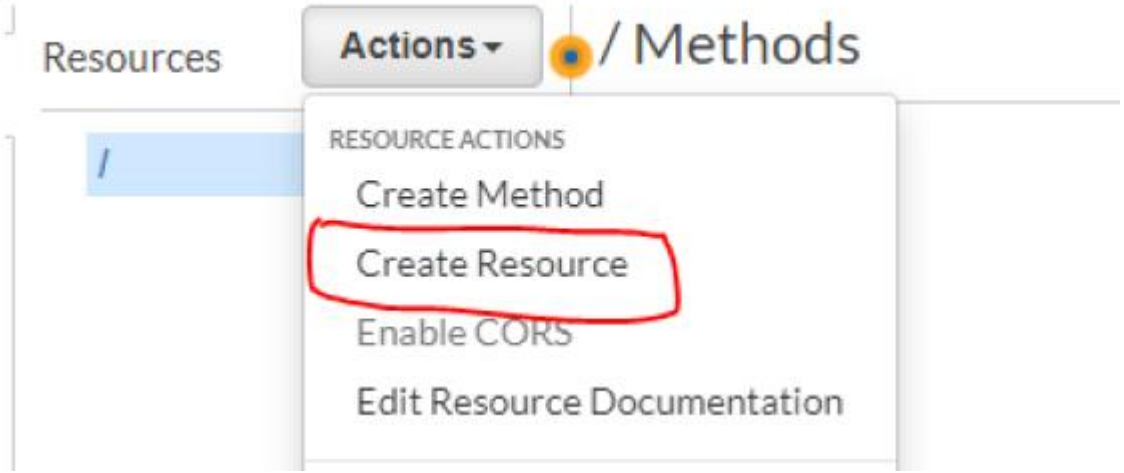
Create API


REST API
Develop a REST API where you gain complete control over the request and response along with API management capabilities.
Works with the following:
Lambda, HTTP, AWS Services
Import
Build


Settings
Choose a friendly name and description for your API.
API name*
test-api
Description
test api
Endpoint Type
Regional
Create API


In Amazon API Gateway, a REST API refers to a collection of resources an
New API
Clone from existing API

Create Resource



Use this page to create a new child resource for your resource.

Configure as proxy resource

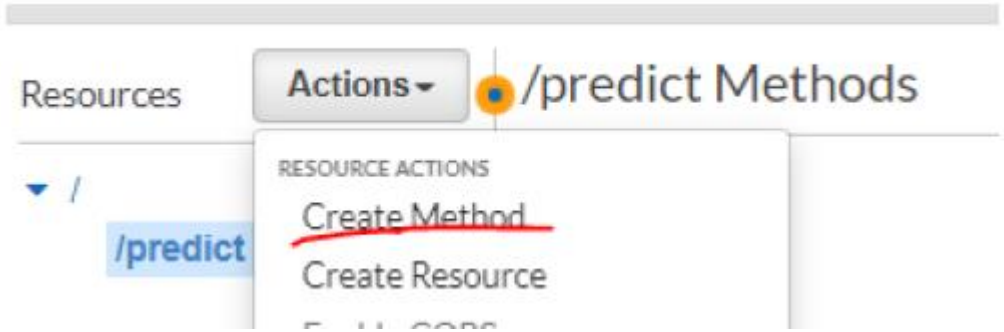
Resource Name* predict

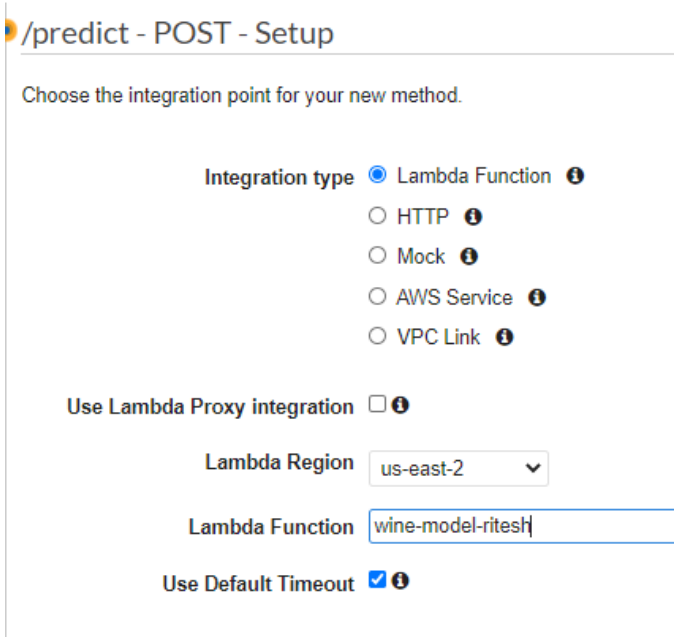
Resource Path* / predict

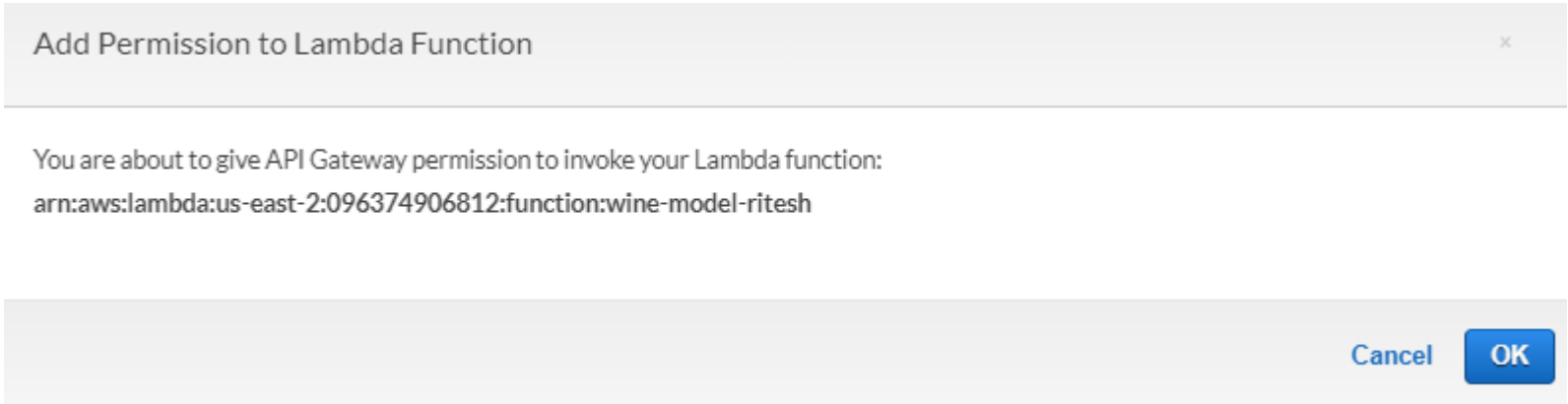
You can add path parameters using brackets. For exa
parameter called 'username'. Configuring /{proxy+} as
example, it works for a GET request to /foo. To handle

Enable API Gateway CORS

Create Method



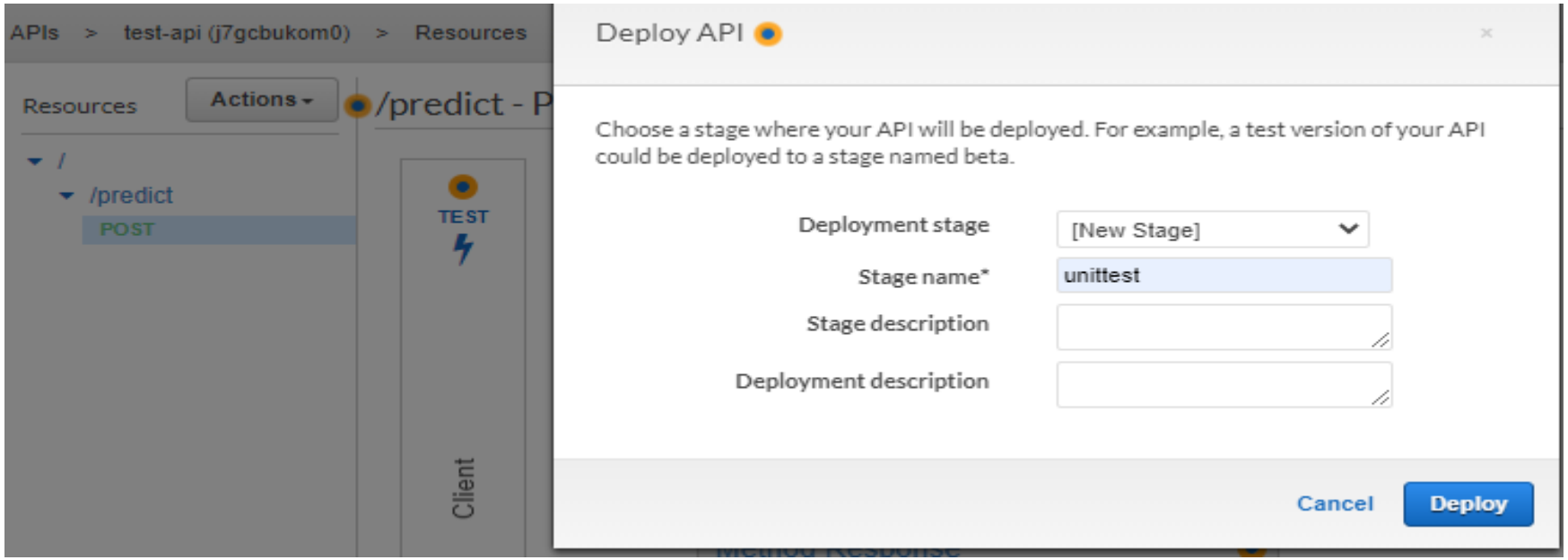




Test Method


Method Execution
/predict - POST - Method Test
Make a test call to your method. When you make a test call, API Gateway skips authorization and directly invokes your method
Path
No path parameters exist for this resource. You can define path parameters by using the syntax {myPathParam} in a resource path.
Query Strings
{predict}
param1=value1&param2=value2
Headers
{predict}
Use a colon (:) to separate header name and value, and new lines to declare multiple headers. eg. Accept:application/json.
Stage Variables
No stage variables exist for this method.
Request Body
1 {"alcohol": "12", "malic_acid": "5"}
Request: /predict
Status: 200
Latency: 4113 ms
Response Body
{
"statusCode": 200,
"body": "\"time taken is: 0.4230184555053711\"",
"time": 0.4230184555053711,
"predict": "class_[2]"
}
Response Headers
{"X-Amzn-Trace-Id":"Root=1-605b2743-e57abf44bbc2983a5
ype":"application/json"}
Logs
Execution log for request 8613143a-82b4-4bb7-8ed3-66e
Wed Mar 24 11:49:23 UTC 2021 : Starting execution for
8ed3-66e009d65fea
Wed Mar 24 11:49:23 UTC 2021 : HTTP Method: POST, Res
Wed Mar 24 11:49:23 UTC 2021 : Method request path: {
Wed Mar 24 11:49:23 UTC 2021 : Method request query s
Wed Mar 24 11:49:23 UTC 2021 : Method request headers
Wed Mar 24 11:49:23 UTC 2021 : Method request body be

Deploy API



COPY THE URL : https://j7gcbukom0.execute-api.us-east-2.amazonaws.com/unittest

Test from command line

curl -v -X POST "https://<subdomain>.execute-api.us-east-2.amazonaws.com/unittest/predict/" -H "content-type: application/json" -d "{ \"alcohol\": \"12\", \"malic_acid\": \"15\" }"

{"statusCode": 200, "body": "\"time taken is: 0.42817234992980957\"", "time": 0.42817234992980957, "predict": "class_[2]"}

Discussion

Creating a layer:

```
import os
import shutil
os.listdir('.')
! pip3 install --ignore-installed --target=python pandas
# ! pip3 install --ignore-installed --target=python scikit-learn ‘ for scikit learn
! rm -rf python/numpy* python/scipy*
! zip -r ./pandas.zip python
```

Deploying a deep NN model

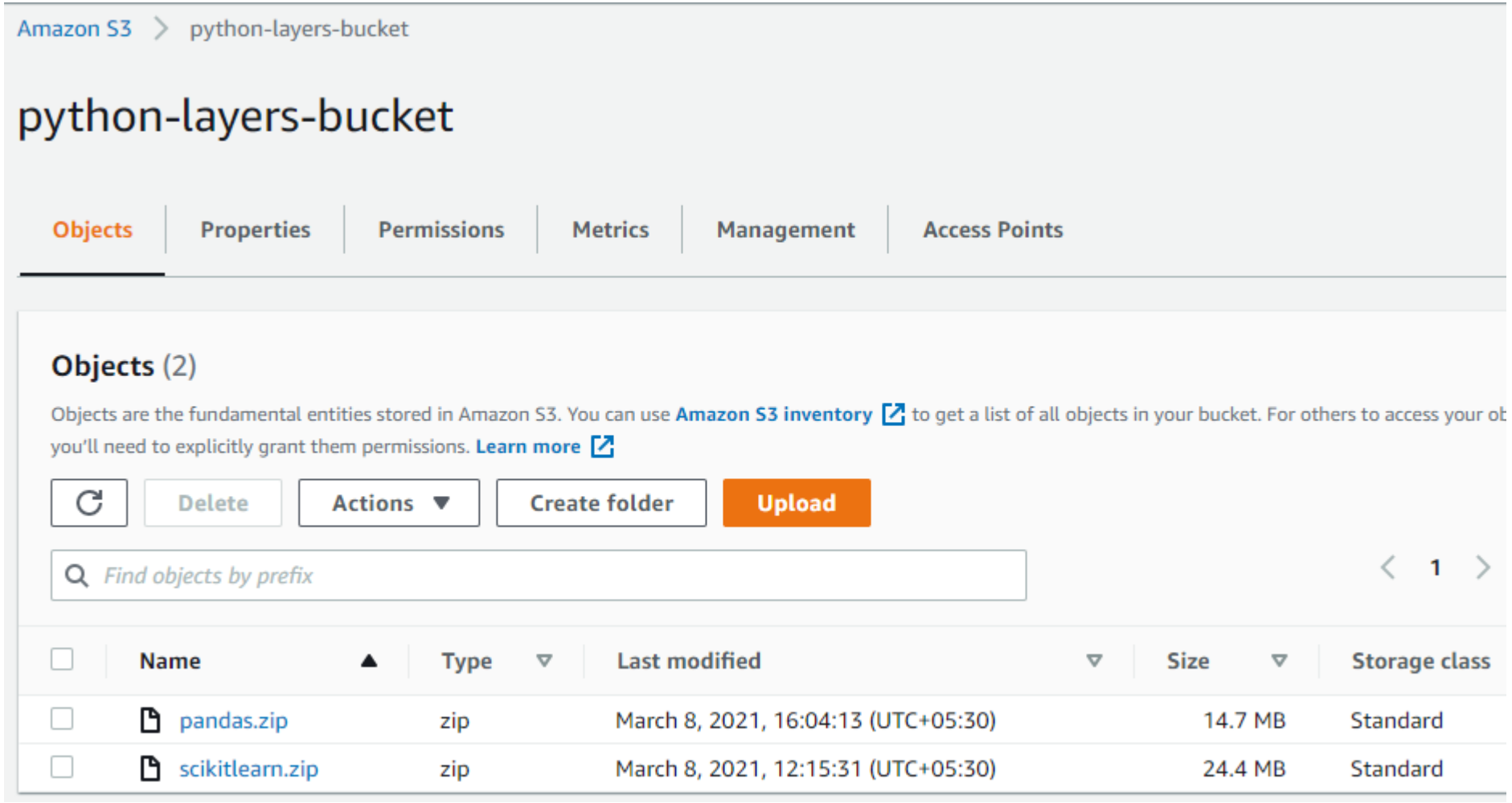
https://github.com/sinharitesh/lambda-multilabel

Exercise for participants

Advanced Exercise:
Move the code loading to lambda handler call and note the time taken. is there a drop in subsequent calls. that will tell if there is time difference.

DELETED

BACKUP


Amazon S3 > python-layers-bucket
python-layers-bucket
Objects
Properties
Permissions
Metrics
Management
Access Points
Objects (2)
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 ob
you'll need to explicitly grant them permissions. Learn more
Delete
Actions
Create folder
Upload
Find objects by prefix
1
Name
Type
Last modified
Size
Storage class
pandas.zip
zip
March 8, 2021, 16:04:13 (UTC+05:30)
14.7 MB
Standard
scikitlearn.zip
zip
March 8, 2021, 12:15:31 (UTC+05:30)
24.4 MB
Standard

Securing the API