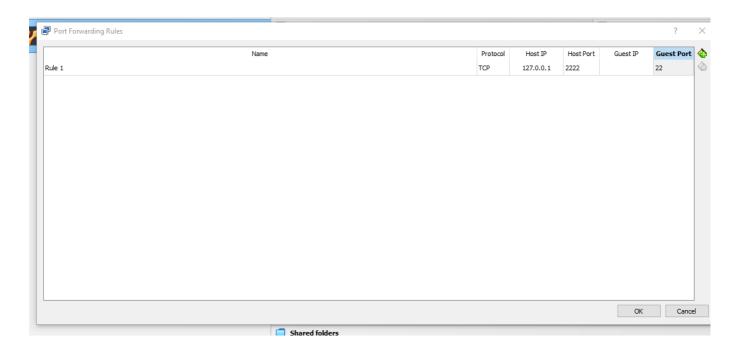
Lab Report 5 - 9

Author: Joo Kai Tay (22489437)

Lab 5: Networking

Section 1: Configure inbound IP on VM

1. Configure the network adapted in VirtualBox Manager using the rule: host IP 127.0.0.1 and host port 2222 mapped to Guest Port 22



2. Install tasksel and openssh-server

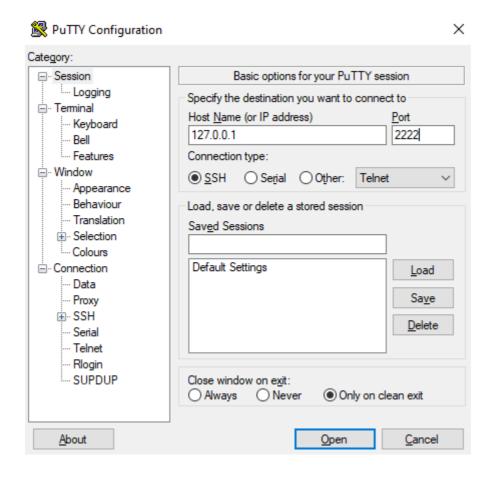
```
jookai@jookai:~$ sudo apt install tasksel

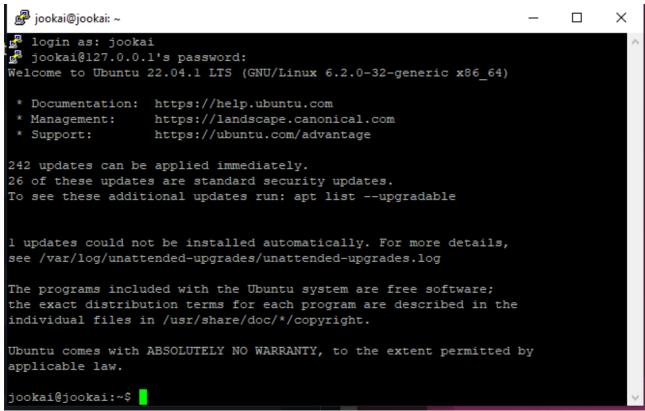
| Toccssting triggers for him as (2.10.2 1) ...
| jookai@jookai:~$ sudo tasksel install openssh-server
```

3. Starting the ssh service on the ubuntu VM

```
jookai@jookai:~$ sudo service ssh start
```

4. SSH into the Ubuntu VM from the hostOS using Putty:





5. Terminate the SSH service:

```
jookai@jookai:~$ sudo service ssh stop
```

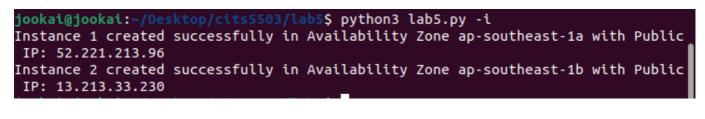
Section 2: Setting up an Application Load Balancer

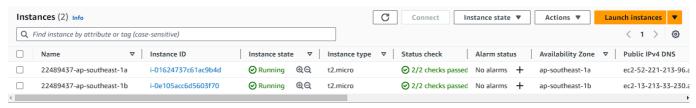
1. The following function is used to create 2 EC2 instances in two different availability zones of apsoutheast-1. The reason ap-southeast-1 was used instead of ap-southeast-2 was due to the limit in VPCUs on ap-southeast-2 which did not allow for any new EC2 instances to be created on the region at the time of attempting this lab.

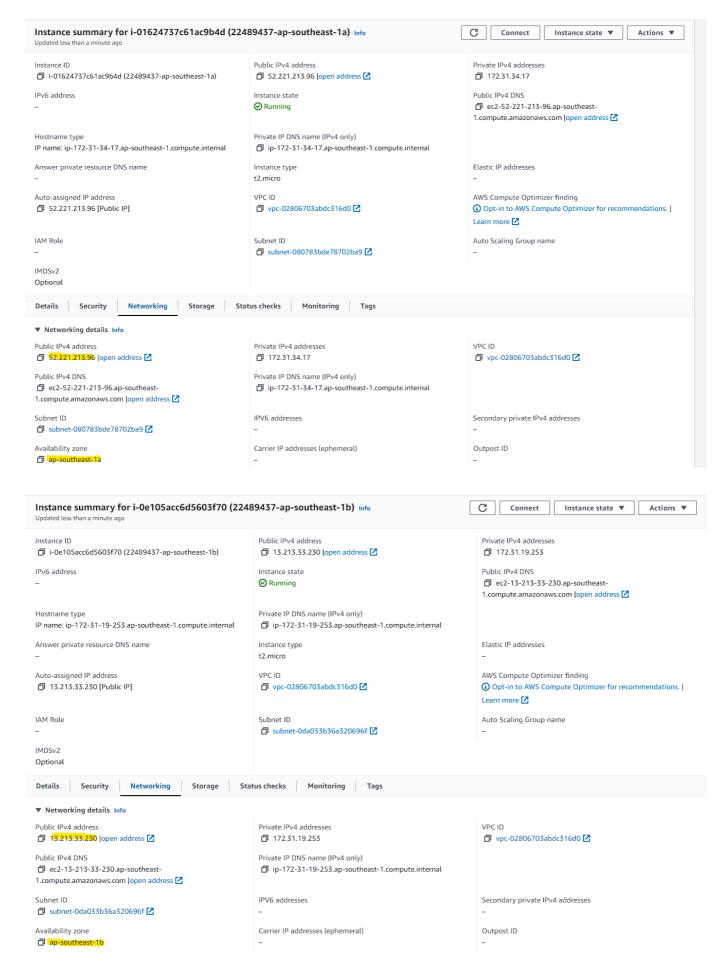
```
def launch_ec2_instances():
   # Create a security group
   response = ec2.create_security_group(
        GroupName=f"{student_number}-sg",
        Description="security group for development environment"
    security_group_id = response['GroupId']
   # Authorize inbound SSH traffic for the security group
   ec2.authorize_security_group_ingress(
        GroupId=security_group_id,
        IpProtocol="tcp",
        FromPort=22,
        ToPort=22,
       CidrIp="0.0.0.0/0"
   )
   # Create a key pair and save the private key to a file
   response = ec2.create_key_pair(KeyName=f"{student_number}-key")
   private_key = response['KeyMaterial']
   private_key_file = f"{student_number}-key.pem"
   # Allow writing to the private key file
   os.chmod(private_key_file, 0o666)
   with open(private_key_file, 'w') as key_file:
        key file.write(private key)
   # Set the correct permissions for the private key file
   os.chmod(private_key_file, 0o400)
   # Copy the private key file to ~/.ssh directory
   ssh directory = os.path.expanduser("~/.ssh")
   if not os.path.exists(ssh_directory):
        os.makedirs(ssh directory)
    shutil.copy(private_key_file, ssh_directory)
   availability zones = ["ap-southeast-1a", "ap-southeast-1b"]
   for i, az in enumerate(availability_zones):
        instance_name = f"{student_number}-{az}"
        instance_params = {
            'ImageId': 'ami-0df7a207adb9748c7',
            'InstanceType': 't2.micro',
            'KeyName': f"{student_number}-key",
            'SecurityGroupIds' : [security_group_id],
            'MinCount': 1,
            'MaxCount': 1,
```

```
'Placement': {'AvailabilityZone': az},
            'TagSpecifications': [
                    'ResourceType': 'instance',
                    'Tags': [{'Key': 'Name', 'Value': instance_name}]
                }
            ]
        }
        # Launch an EC2 instance
        response = ec2.run_instances(**instance_params)
        instance_id = response['Instances'][0]['InstanceId']
        # Wait for the instance to be up and running
        ec2.get_waiter('instance_running').wait(InstanceIds=[instance_id])
        # Describe the instance to get its public IP address
        response = ec2.describe instances(InstanceIds=[instance id])
        public_ip_address = response['Reservations'][0]['Instances'][0]
['PublicIpAddress']
        print(f"Instance {i+1} created successfully in Availability Zone {az} with
Public IP: {public_ip_address}")
```

The created EC2 instances can be observed below. Note that the highlighted public IP addresses and availability zones in the AWS console correspond to the terminal output.







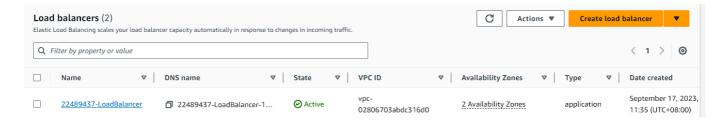
2. The code below creates an application load balancer. a. The code creates the load balancer and specifies the two region subnets retreived from step 1. b. The code creates a listener with a default rule

Protocol: HTTP and Port 80 forwarding on to the target group c. The code creates a target group using the VPC from step 1 d. The code registers the two EC2 instances from step 1 as targets

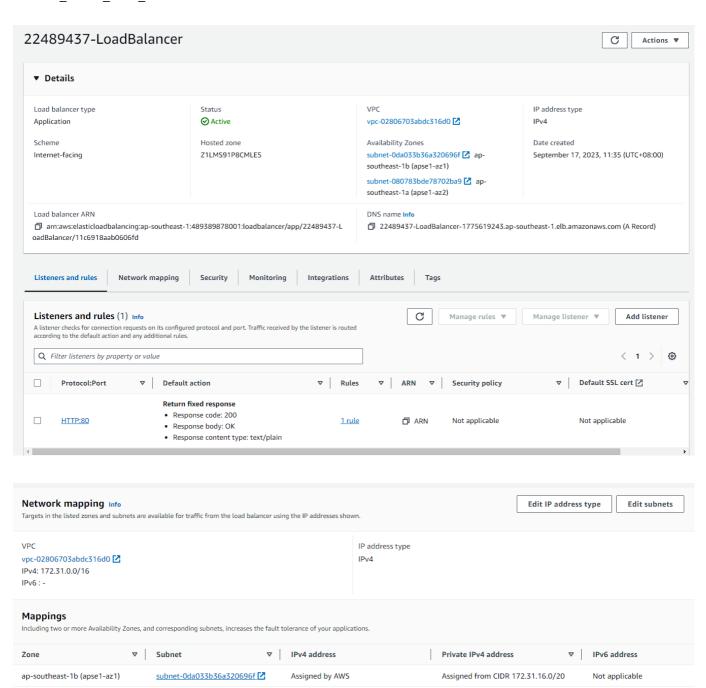
```
def create_load_balancer():
    vpc_id = 'vpc-02806703abdc316d0'
    security_group_id = 'sg-0021774194b407020'
    subnet_ids = ['subnet-080783bde78702ba9', 'subnet-0da033b36a320696f']
    response = elb.create load balancer(
        Name='22489437-LoadBalancer',
        Subnets=subnet_ids,
        SecurityGroups=[security_group_id],
        Scheme='internet-facing',
        Tags=[
            {
                'Key': 'Name',
                'Value': '22489437-LoadBalancer'
            },
        ]
    )
    load balancer arn = response['LoadBalancers'][0]['LoadBalancerArn']
    print(f"Load Balancer ARN: {load_balancer_arn}")
    # Create a target group
    response = elb.create_target_group(
        Name='22489437-target-group',
        Protocol='HTTP',
        Port=80,
        VpcId=vpc id,
        TargetType='instance'
    )
    # Get the ARN of the target group
    target group arn = response['TargetGroups'][0]['TargetGroupArn']
    print(f"Target Group ARN: {target_group_arn}")
    # Create a listener for HTTP traffic (Port 80)
    response = elb.create_listener(
        DefaultActions=[
            {
                'Type': 'forward',
                'TargetGroupArn': target_group_arn,
            },
        LoadBalancerArn=load balancer arn,
        Port=80,
        Protocol='HTTP',
    )
    listener_arn = response['Listeners'][0]['ListenerArn']
    print(f"Listener ARN: {listener_arn}")
```

The following screenshots show the output of running the code as well as the results in the AWS terminal.





ap-southeast-1a (apse1-az2)

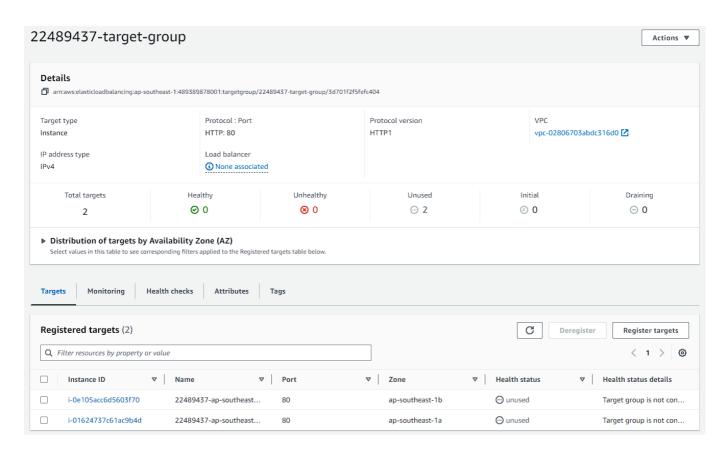


Assigned by AWS

Assigned from CIDR 172.31.32.0/20

Not applicable

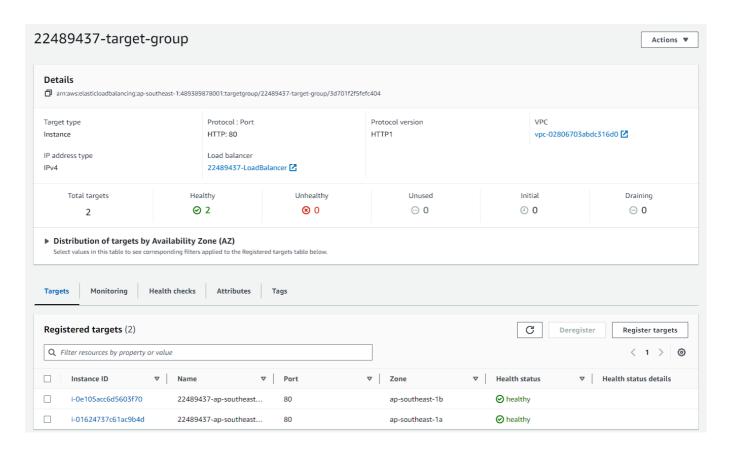
subnet-080783bde78702ba9 🛂



3. In this step, we will SSH into each of the instances created in step 1 and install Apache2. Screenshots showing this process for one of the EC2 instances have been attached:

```
jookai@jookai:~/.ssh$ ssh -i 22489437-key.pem ubuntu@52.221.213.96
The authenticity of host '52.221.213.96 (52.221.213.96)' can't be established.
ED25519 key fingerprint is SHA256:PHQL/z7oZ1klTmFoiao+r0jS708r4bLdfADQwiAlBIg.
This key is not known by any other names
Are you sure you want to continue connecting (yes/no/[fingerprint])? yes
Warning: Permanently added '52.221.213.96' (ED25519) to the list of known hosts.
Welcome to Ubuntu 22.04.2 LTS (GNU/Linux 5.19.0-1025-aws x86 64)
 * Documentation: https://help.ubuntu.com
 * Management:
                   https://landscape.canonical.com
 * Support:
                   https://ubuntu.com/advantage
 System information as of Sun Sep 17 03:50:30 UTC 2023
                                                         96
 System load: 0.0
                                  Processes:
                20.6% of 7.57GB Users logged in:
 Usage of /:
                                                         0
 Memory usage: 24%
                                  IPv4 address for eth0: 172.31.34.17
 Swap usage:
Expanded Security Maintenance for Applications is not enabled.
0 updates can be applied immediately.
Enable ESM Apps to receive additional future security updates.
See https://ubuntu.com/esm or run: sudo pro status
The list of available updates is more than a week old.
To check for new updates run: sudo apt update
The programs included with the Ubuntu system are free software;
the exact distribution terms for each program are described in the
individual files in /usr/share/doc/*/copyright.
Ubuntu comes with ABSOLUTELY NO WARRANTY, to the extent permitted by
applicable law.
To run a command as administrator (user "root"), use "sudo <command>".
See "man sudo root" for details.
ubuntu@ip-172-31-34-17:~$ S
```

```
ubuntu@ip-172-31-34-17:~$ sudo apt install apache2
Reading package lists... Done
Building dependency tree... Done
Reading state information... Done
The following additional packages will be installed:
  apache2-bin apache2-data apache2-utils bzip2 libapr1 libaprutil1
  libaprutil1-dbd-sqlite3 libaprutil1-ldap liblua5.3-0 mailcap mime-support
  ssl-cert
Suggested packages:
  apache2-doc apache2-suexec-pristine | apache2-suexec-custom www-browser
  bzip2-doc
The following NEW packages will be installed:
  apache2 apache2-bin apache2-data apache2-utils bzip2 libapr1 libaprutil1
  libaprutil1-dbd-sqlite3 libaprutil1-ldap liblua5.3-0 mailcap mime-support
  ssl-cert
O upgraded, 13 newly installed, O to remove and 127 not upgraded.
Need to get 2137 kB of archives.
After this operation, 8505 kB of additional disk space will be used.
Do you want to continue? [Y/n]
```



4. In this step we will edit the /var/www/html/index.html file to report the instance name and availability zone.

```
<!DOCTYPE html>
<html>
<body>
<h1>This is Instance 1 from availability zone ap-southeast-1a</h1>
</body>
</hd>
</rr>
</body>
```

```
<!DOCTYPE html>
<html>
<body>
<h1>This is instance 2 from availability zone ap-southeast-1b</h1>
</body>
</html>
```

5. By refreshing the page repeatedly, we can access both EC2 instances



This is instance 2 from availability zone ap-southeast-1b

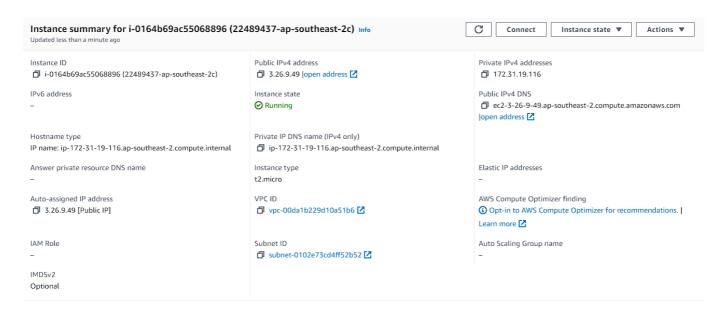
Section 1: Create an EC2 Instance

1. The code below was used to create an EC2 instance on ap-southeast-2c. For the lab this week, there was available capacity on ap-southeast-2 so there was no need to create the instance on another region. The AMI provided in lab 2 ami-d38a4ab1 had a heavily outdated version of python and other utilities that are not compatible with modern programs, therefore an updated AMI was selected instead ami-0310483fb2b488153.

```
jookai@jookai:~/Desktop/cits5503/lab6$ python3 lab6.py -i
Instance 1 created successfully in Availability Zone ap-southeast-2c with Public
IP: 3.26.9.49
```

```
def launch_ec2_instances():
   # Create a security group
   response = ec2.create_security_group(
        GroupName=f"{student_number}-sg",
        Description="security group for development environment"
    security_group_id = response['GroupId']
   # Authorize inbound SSH traffic for the security group
   ec2.authorize_security_group_ingress(
        GroupId=security_group_id,
        IpProtocol="tcp",
        FromPort=22,
       ToPort=22,
        CidrIp="0.0.0.0/0"
    )
    ec2.authorize security group ingress(
        GroupId=security group id,
        IpProtocol="tcp",
        FromPort=80,
        ToPort=80,
        CidrIp="0.0.0.0/0"
    )
   # Create a key pair and save the private key to a file
   response = ec2.create_key_pair(KeyName=f"{student_number}-key")
   private key = response['KeyMaterial']
   private_key_file = f"{student_number}-key.pem"
   # Allow writing to the private key file
   os.chmod(private key file, 0o666)
   with open(private_key_file, 'w') as key_file:
        key_file.write(private_key)
   # Set the correct permissions for the private key file
   os.chmod(private_key_file, 0o400)
   # Copy the private key file to ~/.ssh directory
   ssh_directory = os.path.expanduser("~/.ssh")
   if not os.path.exists(ssh_directory):
        os.makedirs(ssh_directory)
```

```
shutil.copy(private_key_file, ssh_directory)
   availability_zones = ["ap-southeast-2b", "ap-southeast-2c"]
   for i, az in enumerate(availability_zones):
        instance_name = f"{student_number}-{az}"
        instance_params = {
            'ImageId': 'ami-0310483fb2b488153',
            'InstanceType': 't2.micro',
            'KeyName': f"{student_number}-key",
            'SecurityGroupIds' : [security_group_id],
            'MinCount': 1,
            'MaxCount': 1,
            'Placement': {'AvailabilityZone': az},
            'TagSpecifications': [
                {
                    'ResourceType': 'instance',
                    'Tags': [{'Key': 'Name', 'Value': instance_name}]
                }
            ]
        }
        # Launch an EC2 instance
        response = ec2.run_instances(**instance_params)
        instance_id = response['Instances'][0]['InstanceId']
        # Wait for the instance to be up and running
        ec2.get waiter('instance running').wait(InstanceIds=[instance id])
        # Describe the instance to get its public IP address
        response = ec2.describe_instances(InstanceIds=[instance_id])
        public_ip_address = response['Reservations'][0]['Instances'][0]
['PublicIpAddress']
        print(f"Instance {i+1} created successfully in Availability Zone {az} with
Public IP: {public_ip_address}")
```



2. Using the private key obtained and public IP address obtained from step 1, SSH into the EC2 instance and install the Python 3 virtual environment package.

```
The authenticity of host '3.26.9.49 (3.26.9.49)' can't be established.
ED25519 key fingerprint is SHA256:U83QWvm3/tMOx/A5pm0k9WqrPlr0CgYxUyQjKTAFdys.
This key is not known by any other names
Are you sure you want to continue connecting (yes/no/[fingerprint])?         yes
Warning: Permanently added '3.26.9.49' (ED25519) to the list of known hosts.
Welcome to Ubuntu 22.04.2 LTS (GNU/Linux 5.19.0-1025-aws x86 64)
* Documentation: https://help.ubuntu.com
  Management:
                 https://landscape.canonical.com
 Support:
                 https://ubuntu.com/advantage
 System information as of Sat Sep 23 00:47:17 UTC 2023
 System load:
              0.20166015625
                                Processes:
                                                     99
              20.6% of 7.57GB
                               Users logged in:
 Usage of /:
                                                     0
                                IPv4 address for eth0: 172.31.19.116
 Memory usage: 24%
              0%
 Swap usage:
```

```
ubuntu@ip-172-31-19-116:~$ sudo apt-get update
ubuntu@ip-172-31-19-116:~$ sudo apt-get upgrade
```

```
ubuntu@ip-172-31-19-116:~$ sudo apt-get install python3-venv
```

3. Creating a directory with path /opt/wwc/mysites and setting up the virtual environment.

```
root@ip-172-31-19-116:/home/ubuntu# sudo mkdir -p /opt/wwc/mysites

root@ip-172-31-19-116:/home/ubuntu# cd /opt/wwc/mysites

root@ip-172-31-19-116:/opt/wwc/mysites# python3 -m venv myvenv

root@ip-172-31-19-116:/opt/wwc/mysites# source myvenv/bin/activate

(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites#
```

- 4. A Django project is a collection of configurations and apps for a particular website. In this step we install Django and create a new Django app named polls.
 - lab: The configuration directory
 - o polls: The directory containing the app
 - manage.py: The command line utility that lets us interact with the new app

```
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites# pip install django
Collecting django
```

```
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites# django-admin startproject lab
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites# cd lab
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites/lab# python3 manage.py startapp
polls
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites/lab# ls
lab manage.py polls
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites/lab#
```

Section 2: Install and Congigure Nginx

- 1. Installing and configuring nginx:
 - Nginx is a popular open-source web server software that can also be used as a reverse proxy, load balancer, mail proxy, and HTTP cache.
 - The configuration file is edited to tell Nginx to pass requests to the backend server running on the same machine 127.0.0.1 at port 8000.

```
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites/lab# apt install nginx
```

```
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites/lab# vi /etc/nginx/sites-enabled
/default
```

```
server {
  listen 80 default_server;
  listen [::]:80 default_server;

  location / {
    proxy_set_header X-Forwarded-Host $host;
    proxy_set_header X-Real-IP $remote_addr;

    proxy_pass http://127.0.0.1:8000;
  }
```

2. Restarting Nginx so that the changes from step 1 take effect:

```
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites/lab# service nginx restart
```

3. Using the command python3 manage.py runserver 8000 to start Django's development web server at port 8000.

```
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites/lab# python3 manage.py runserver
8000
Watching for file changes with StatReloader
Performing system checks...
System check identified no issues (0 silenced).

You have 18 unapplied migration(s). Your project may not work properly until you apply the migrations for app(s): admin, auth, contenttypes, sessions.
Run 'python manage.py migrate' to apply them.
September 23, 2023 - 01:07:45
Django version 4.2.5, using settings 'lab.settings'
Starting development server at http://127.0.0.1:8000/
Quit the server with CONTROL-C.
```

4. Trying the access the public IP address of the EC2 instance results in an error:



This site can't be reached

3.26.9.49 took too long to respond.

Try:

- · Checking the connection
- · Checking the proxy and the firewall
- Running Windows Network Diagnostics

ERR_CONNECTION_TIMED_OUT

Reload

Details

```
Not Found: /polls/
[23/Sep/2023 08:14:39] "GET /polls/ HTTP/1.0" 404 2092
Not Found: /polls/
[23/Sep/2023 08:14:56] "GET /polls/ HTTP/1.0" 404 2092
```

Section 3: Change the code

- 1. Editing polls/views.py
 - Thsi code creates a simple view that returns an HTTP response with the text "Hello, world." when it's called.

```
from django.shortcuts import render
from django.http import HttpResponse

def index(request):
    return HttpResponse("Hello, world.")
```

2. Edit polls/urls.py

• This code defines a URL pattern for this view in the urls.py file, so that Django knows which view to call for a given URL.

3. Edit lab/urls.py

• The code configures the URL patterns for the Django project.

```
URL configuration for lab project.
The `urlpatterns` list routes URLs to views. For more information please see:
   https://docs.djangoproject.com/en/4.2/topics/http/urls/
Examples:
Function views
    1. Add an import: from my_app import views
    Add a URL to urlpatterns: path('', views.home, name='home')
Class-based views

    Add an import: from other_app.views import Home

    Add a URL to urlpatterns: path('', Home.as_view(), name='home')
Including another URLconf
    1. Import the include() function: from django.urls import include, path
    Add a URL to urlpatterns: path('blog/', include('blog.urls'))
from django.contrib import admin
from django.urls import include, path
from django.contrib import admin
urlpatterns = 🛮
    path('polls/', include('polls.urls')),
    path('admin/', admin.site.urls),
  INSERT --
                                                                            Bot
                                                              24,2
```

4. Running the application and getting Hello, world

```
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites/lab# vi polls/views.py
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites/lab# vi polls/urls.py
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites/lab# vi lab/urls.py
(myvenv) root@ip-172-31-19-116:/opt/wwc/mysites/lab# python3 manage.py runserver
8000
Watching for file changes with StatReloader
Performing system checks...
System check identified no issues (0 silenced).

You have 18 unapplied migration(s). Your project may not work properly until you
apply the migrations for app(s): admin, auth, contenttypes, sessions.
Run 'python manage.py migrate' to apply them.
September 23, 2023 - 01:31:53
Django version 4.2.5, using settings 'lab.settings'
Starting development server at http://127.0.0.1:8000/
Quit the server with CONTROL-C.
```

```
← → C 🛕 Not secure | 3.26.9.49/polls
```

Hello, world.

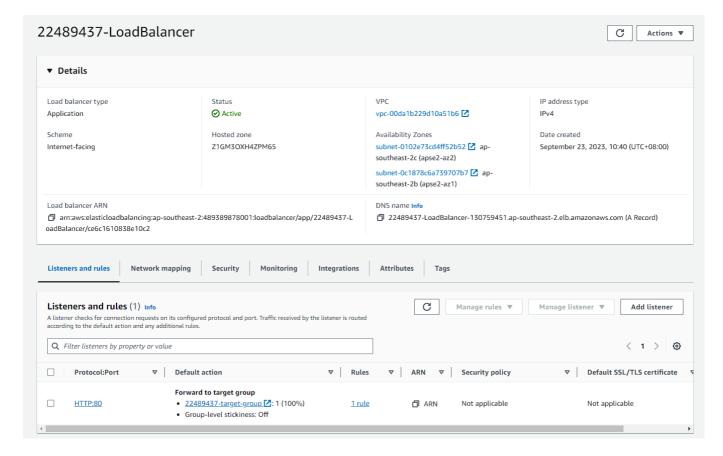
Section 4: Adding an application load balance (ALB)

1. The code below creates an application load balancer, specifies the region subnet where the EC2 instance resides, creates a listener with a default rule Protocol: HTTP and Port 80 forwarding.

```
def create_load_balancer():
    vpc id = 'vpc-00da1b229d10a51b6'
    security_group_id = 'sg-0fb8992bd2473b7bc'
    subnet_ids = ['subnet-0c1878c6a739707b7', 'subnet-0102e73cd4ff52b52']
    response = elb.create_load_balancer(
        Name='22489437-LoadBalancer',
        Subnets=subnet ids,
        SecurityGroups=[security_group_id],
        Scheme='internet-facing',
        Tags=[
                'Key': 'Name',
                'Value': '22489437-LoadBalancer'
            },
        ]
    )
    load_balancer_arn = response['LoadBalancers'][0]['LoadBalancerArn']
    print(f"Load Balancer ARN: {load_balancer_arn}")
    # Create a target group
    response = elb.create_target_group(
        Name='22489437-target-group',
        Protocol='HTTP',
        Port=80,
        VpcId=vpc id,
        HealthCheckProtocol='HTTP',
```

```
HealthCheckPort='80',
    HealthCheckPath='/polls/',
    HealthCheckIntervalSeconds=30,
    HealthCheckTimeoutSeconds=5,
    HealthyThresholdCount=5,
    UnhealthyThresholdCount=2,
    Matcher={
        'HttpCode': '200'
    },
    TargetType='instance'
)
# Get the ARN of the target group
target_group_arn = response['TargetGroups'][0]['TargetGroupArn']
print(f"Target Group ARN: {target_group_arn}")
# Create a listener for HTTP traffic (Port 80)
response = elb.create listener(
    DefaultActions=[
        {
            'Type': 'forward',
            'TargetGroupArn': target_group_arn,
        },
    ],
    LoadBalancerArn=load_balancer_arn,
    Port=80,
    Protocol='HTTP',
)
listener_arn = response['Listeners'][0]['ListenerArn']
print(f"Listener ARN: {listener arn}")
instance_1_id = 'i-0164b69ac55068896'
# Register the instances in the target group
elb.register_targets(
    TargetGroupArn=target_group_arn,
    Targets=[
        {'Id': instance_1_id},
    1
)
# Print registration status
print("Targets registered successfully.")
```

jookai@jookai:~/Desktop/cits5503/lab6\$ python3 lab6.py -lb
Load Balancer ARN: arn:aws:elasticloadbalancing:ap-southeast-2:489389878001:load
balancer/app/22489437-LoadBalancer/ce6c1610838e10c2
Target Group ARN: arn:aws:elasticloadbalancing:ap-southeast-2:489389878001:targe
tgroup/22489437-target-group/1448f7d6495f1fac
Listener ARN: arn:aws:elasticloadbalancing:ap-southeast-2:489389878001:listener/
app/22489437-LoadBalancer/ce6c1610838e10c2/e43f09b70777fe3d
Targets registered successfully.



2. Viewing the health check on the /polls/ page

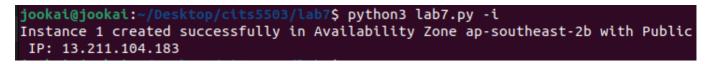
```
[23/Sep/2023 08:09:55] "GET /polls/ HTTP/1.0"
                                              200 13
[23/Sep/2023 08:10:08] "GET /polls/ HTTP/1.0"
                                              200 13
[23/Sep/2023 08:10:25] "GET /polls/ HTTP/1.0" 200 13
                                              200 13
[23/Sep/2023 08:10:39]
                       "GET /polls/ HTTP/1.0"
[23/Sep/2023 08:10:55] "GET /polls/ HTTP/1.0" 200 13
[23/Sep/2023 08:11:09] "GET /polls/ HTTP/1.0" 200 13
[23/Sep/2023 08:11:26] "GET /polls/ HTTP/1.0" 200 13
[23/Sep/2023 08:11:39] "GET /polls/ HTTP/1.0" 200 13
[23/Sep/2023 08:11:56]
                      "GET /polls/ HTTP/1.0" 200 13
[23/Sep/2023 08:12:09] "GET /polls/ HTTP/1.0" 200 13
[23/Sep/2023 08:12:26] "GET /polls/ HTTP/1.0" 200 13
[23/Sep/2023 08:12:39] "GET /polls/ HTTP/1.0" 200 13
[23/Sep/2023 08:12:56] "GET /polls/ HTTP/1.0" 200 13
[23/Sep/2023 08:13:09]
                       "GET /polls/ HTTP/1.0" 200 13
```

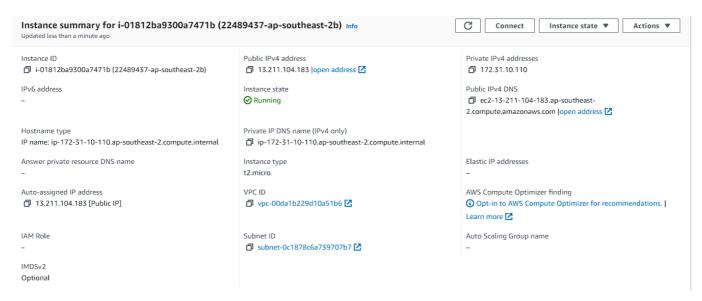
3. Accessing the site using the url:

Lab 7: DevOps

Step 1: Create an EC2 Instance

1. The code used in Labs 5 and 6 was used to create an EC2 instance in availability zone ap-southeast-2b





Step 2: Install Fabric

1. Fabric is a Python library used to perform Linux shell commands remotely over SSH. It is a command-line tool that uses SSH for application deployment or for administration tasks. In this lab we will be using fabric to automate the deployment of a django app on the remote EC2 instance.

```
jookai@jookai:~/Desktop/cits5503/lab7$    pip install fabric
Defaulting to user installation because normal site-packages is not writeable
Collecting fabric
  Downloading fabric-3.2.2-py3-none-any.whl (59 kB)
                                             - 59.4/59.4 KB 2.0 MB/s
                                                                     eta 0:00:00
Requirement already satisfied: paramiko>=2.4 in /usr/lib/python3/dist-packages (
from fabric) (2.9.3)
Collecting invoke>=2.0
  Downloading invoke-2.2.0-py3-none-any.whl (160 kB)
                                            · 160.3/160.3 KB 5.0 MB/s eta 0:00:00
Collecting deprecated>=1.2
  Downloading Deprecated-1.2.14-py2.py3-none-any.whl (9.6 kB)
Collecting decorator>=5
  Downloading decorator-5.1.1-py3-none-any.whl (9.1 kB)
Collecting wrapt<2,>=1.10
  Downloading wrapt-1.15.0-cp310-cp310-manylinux_2_5_x86_64.manylinux1_x86_64.ma
nylinux_2_17_x86_64.manylinux2014_x86_64.whl (78 kB)
                                              78.4/78.4 KB 5.2 MB/s eta 0:00:00
Installing collected packages: wrapt, invoke, decorator, deprecated, fabric
Successfully installed decorator-5.1.1 deprecated-1.2.14 fabric-3.2.2 invoke-2.2
.0 wrapt-1.15.0
```

The configuration file located in ~/.ssh contains the IPV4 DNS of the EC2 instance and the path of the private key file on the local VM to allow SSH access into the EC2 instance.

```
Host <ec2instance>
    Hostname ec2-13-211-104-183.ap-southeast-2.compute.amazonaws.com
    User ubuntu
    UserKnownHostsFile /dev/null
    StrictHostKeyChecking no
    PasswordAuthentication no
    IdentityFile ~/.ssh/22489437-key.pem

"config" [New File]
```

2. Testing Fabric

```
jookai@jookai:~/.ssh$ python3
Python 3.10.12 (main, Jun 11 2023, 05:26:28) [GCC 11.4.0] on linux
Type "help", "copyright", "credits" or "license" for more information.
>>> from fabric import Connection
>>> c = Connection('<ec2instance>')
>>> result = c.run('uname -s')
Linux
>>>
```

Step 3: Python Script to automate installation of nginx

1. The following script is used to automate the instllation of nginx on the EC2 instance.

```
def create_nginx():
    with Connection('<ec2instance>') as c:
        c.sudo('apt update -y')
        c.sudo('apt upgrade -y')
        c.sudo('apt install nginx -y')
        c.sudo('rm /etc/nginx/sites-enabled/default')
        os.system('scp -i ~/.ssh/22489437-key.pem default ubuntu@ec2-13-211-104-
183.ap-southeast-2.compute.amazonaws.com:/home/ubuntu')
        c.sudo('mv /home/ubuntu/default /etc/nginx/sites-enabled/')
        c.sudo('service nginx restart')
```

The script is run on the local VM:

```
jookai@jookai:~/Desktop/cits5503/lab7$ python3 lab7.py -n
WARNING: apt does not have a stable CLI interface. Use with caution in scripts.
```

Checking if nginx is installed on the EC2 instance:

```
ubuntu@ip-172-31-10-110:~$ nginx -v
nginx version: nginx/1.18.0 (Ubuntu)
```

Step 4: Python script to install Django app

1. The following script is used to automate the installation of the Django app on the EC2 instanc:

```
def create_django_app():
   with Connection('<ec2instance>') as c:
        c.sudo('mkdir -p /opt/wwc/mysites')
        c.sudo('apt-get install python3-pip -y')
        c.sudo('apt install python3-django -y')
        c.sudo('pip3 install django')
        c.run('django-admin startproject lab')
        c.sudo('chmod 777 ./lab')
        c.run('cd ./lab && python3 manage.py startapp polls')
        c.sudo('mv ./lab /opt/wwc/mysites')
        c.sudo('rm /opt/wwc/mysites/lab/polls/views.py')
        c.sudo('rm /opt/wwc/mysites/lab/lab/urls.py')
        os.system('scp -i ~/.ssh/22489437-key.pem ./polls/views.py ubuntu@ec2-13-
211-104-183.ap-southeast-2.compute.amazonaws.com:/opt/wwc/mysites/lab/polls')
        os.system('scp -i ~/.ssh/22489437-key.pem ./polls/urls.py ubuntu@ec2-13-
211-104-183.ap-southeast-2.compute.amazonaws.com:/opt/wwc/mysites/lab/polls')
        os.system('scp -i ~/.ssh/22489437-key.pem ./lab/urls.py ubuntu@ec2-13-211-
104-183.ap-southeast-2.compute.amazonaws.com:/opt/wwc/mysites/lab/lab')
        c.run('cd /opt/wwc/mysites/lab && python3 manage.py runserver 8000')
```

The script is run on the local VM:

```
jookai@jookai:~/Desktop/cits5503/lab7$ python3 lab7.py -d
views.py
                                               100%
                                                      98
                                                             1.8KB/s
                                                                       00:00
urls.py
                                               100%
                                                             2.0KB/s
                                                                       00:00
                                                     108
urls.py
                                               100%
                                                    169
                                                             2.8KB/s
                                                                       00:00
Watching for file changes with StatReloader
Performing system checks...
System check identified no issues (0 silenced).
You have 18 unapplied migration(s). Your project may not work properly until you
apply the migrations for app(s): admin, auth, contenttypes, sessions.
Run 'python manage.py migrate' to apply them.
September 29, 2023 - 14:07:35
Django version 3.2.12, using settings 'lab.settings'
Starting development server at http://127.0.0.1:8000/
Quit the server with CONTROL-C.
```

Checking to see if Hello world is displayed when visiting the Public IPv4 DNS of the EC2 instance:



Lab 8: Al

Step 1: Install and run jupyter notebooks

1. Jupyter notebook was already installed on this machine for CITS5508 machine learning. Therefore, only the command to launch it was needed.

jookai@jookai:~/Desktop/cits5503/lab8\$ jupyter notebook

2. This brings up Jupyter notebook in the browser where we can create a notebook for this lab named lab8.ipynb which is an Interactive Python Notebook.



Step 2: Setup Python Environment

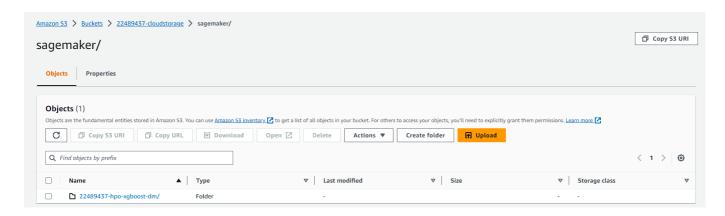
- 1. The following command is run pip3 install sagemaker pandas ipykernel.
 - This installs sagemaker, which is the Python SDK for Amazon SageMaker, that allows us to build, train, and deploy machine learning models. By installing this SDK, we can interact with the SageMaker service directly from the Python environment on the virtual machine.
 - Pandas is a popular data manipulation and analysis library for Python.
 - Ipykernel allows the python environment to interface with Jupyrer.

Step 3: Create Role and Bucket

1. Creating a S3 bucket to store the files generated in the later parts of the machine learning process:



2. Creating the folders sagemaker/22489437-hpo-xgboost-dm



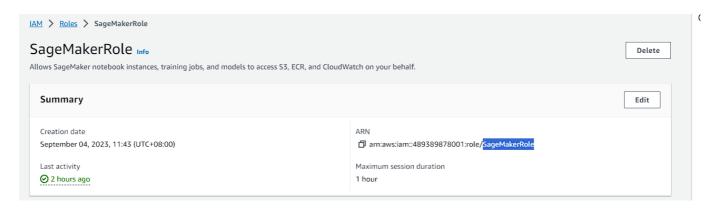
- 3. Due to security reasons, students are not allowed to directly create IAM roles. Therefore, we will use the code provided in the labsheet to use the already created IAM role.
 - smclient = boto3.Session().client("sagemaker"): This initializes a low-level client representing Amazon SageMaker. So, smclient will be used to make requests to the SageMaker service.
 - sagemaker_role = iam.get_role(RoleName='Role_AWS_SageMaker')['Role']['Arn']:
 This uses the IAM client to retrieve the ARN of a specific IAM role named 'Role_AWS_SageMaker'.
 This ARN can later be used to grant permissions or interact with the role in various ways within AWS.

```
region = 'ap-southeast-2'
smclient = boto3.Session().client("sagemaker")

iam = boto3.client('iam')
sagemaker_role = iam.get_role(RoleName='Role_AWS_SageMaker')['Role']['Arn']

student_id = "STUDENTID"
bucket = 'YOUR_BUCKET_NAME_HERE'
prefix = f"sagemaker/{student_id}-hpo-xgboost-dm"
```

4. The code provided fails to execute as the role Role_AWS_SageMaker does not exist. Therefore, we log into the IAM system in the AWS console and find that the actual name of the created roles is SageMakerRole. When using the actual name of the role, the code executes and we can proceed to the next step.



Step 4: Download Dataset

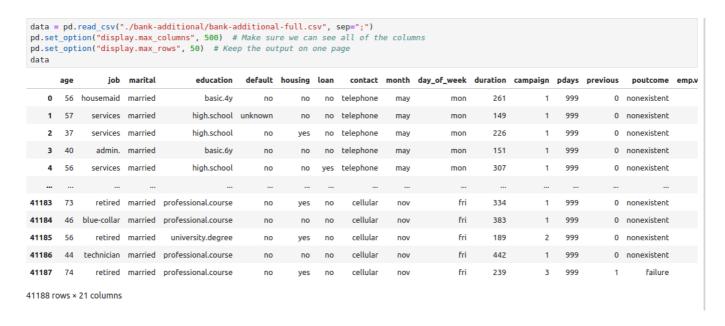
1. We download and extract the direct marketing dataset from UCI's ML Repository.

```
jookai@jookai:~/Desktop/cits5503/lab8$ wget -N https://archive.ics.uci.edu/ml/mac
hine-learning-databases/00222/bank-additional.zip
--2023-10-06 14:54:17-- https://archive.ics.uci.edu/ml/machine-learning-database
s/00222/bank-additional.zip
Resolving archive.ics.uci.edu (archive.ics.uci.edu)... 128.195.10.252
Connecting to archive.ics.uci.edu (archive.ics.uci.edu)|128.195.10.252|:443... co
nnected.
HTTP request sent, awaiting response... 200 OK
Length: unspecified
Saving to: 'bank-additional.zip'
bank-additional.zip
                       [
                              <=>
                                             ] 434.15K
                                                         419KB/s
                                                                    in 1.0s
Last-modified header missing -- time-stamps turned off.
2023-10-06 14:54:19 (419 KB/s) - 'bank-additional.zip' saved [444572]
jookai@jookai:~/Desktop/cits5503/lab8$ unzip -o bank-additional.zip
Archive: bank-additional.zip
  creating: bank-additional/
 inflating: bank-additional/.DS Store
  creating: __MACOSX/
  creating: __MACOSX/bank-additional/
 inflating:
             __MACOSX/bank-additional/._.DS_Store
 inflating: bank-additional/.Rhistory
 inflating: bank-additional/bank-additional-full.csv
 inflating: bank-additional/bank-additional-names.txt
 inflating: bank-additional/bank-additional.csv
 inflating: MACOSX/. bank-additional
```

The dataset is related with direct marketing campaigns of a Portuguese banking institution. The marketing campaigns were based on phone calls. Often, more than one contact to the same client was required, in order to access if the product (bank term deposit) would be ('yes') or not ('no') subscribed.

- 2. We read the dataset into a Pandas dataframe and inspect the contents:
 - We see that the data fame has 41188 rows and 21 columns

```
data = pd.read_csv("./bank-additional/bank-additional-full.csv", sep=";")
pd.set_option("display.max_columns", 500) # Make sure we can see all of the
columns
pd.set_option("display.max_rows", 50) # Keep the output on one page
data
```



- 3. In order to answer which variables are categorical and numerical, the following function was used:
 - o This makes use of the select_dtypes method to filter out the categorical and numeric columns

```
# numeric columns
numeric_cols = data.select_dtypes(include=['number']).columns.tolist()
print("Numeric columns:", numeric_cols)

# categorical columns
categorical_cols = data.select_dtypes(include=['object']).columns.tolist()
print("Categorical columns:", categorical_cols)
```

```
# numeric columns
numeric_cols = data.select_dtypes(include=['number']).columns.tolist()
print("Numeric columns:", numeric_cols)

# categorical columns
categorical_cols = data.select_dtypes(include=['object']).columns.tolist()
print("Categorical columns:", categorical_cols)

Numeric columns: ['age', 'duration', 'campaign', 'pdays', 'previous', 'emp.var.rate', 'cons.price.idx', 'cons.conf.idx', 'euribor3m', 'nr.employed']
Categorical columns: ['job', 'marital', 'education', 'default', 'housing', 'loan', 'contact', 'month', 'day_of_week', 'poutcome', 'y']
```

- 4. The following code performs some transformations on the data:
 - o data["no_previous_contact"] = np.where(data["pdays"] == 999, 1, 0): This line creates a new column in the DataFrame data named "no_previous_contact". The values in this column are determined by the values in the "pdays" column. If a value in "pdays" is 999, the new column gets a value of 1; otherwise, it gets 0.
 - o data["not_working"] = np.where(np.in1d(data["job"], ["student", "retired", "unemployed"]), 1, 0): This line creates another new column named "not_working". It checks the "job" column of the DataFrame. If the job description is either "student", "retired", or "unemployed", the new column gets a value of 1, indicating the person is not actively employed. Otherwise, it gets 0.
 - model_data = pd.get_dummies(data): This line creates a new DataFrame, model_data, by converting the categorical variables in data into "indicator" variables. For each level in a categorical variable, a new column will be created to indicate if it is present or not.

```
data["no_previous_contact"] = np.where(
    data["pdays"] == 999, 1, 0
) # Indicator variable to capture when pdays takes a value of 999
data["not_working"] = np.where(
    np.in1d(data["job"], ["student", "retired", "unemployed"]), 1, 0
) # Indicator for individuals not actively employed
model_data = pd.get_dummies(data) # Convert categorical variables to sets of indicators
model_data
```

This creates the model_data dataframe that has 67 columns.

1:		age	duration	campaign	pdays	previous	emp.var.rate	cons.price.idx	cons.conf.idx	euribor3m	nr.employed	no_previous_contact	not_working	job_admin.	je
	0	56	261	1	999	0	1.1	93.994	-36.4	4.857	5191.0	1	0	0	
	1	57	149	1	999	0	1.1	93.994	-36.4	4.857	5191.0	1	0	0	
	2	37	226	1	999	0	1.1	93.994	-36.4	4.857	5191.0	1	0	0	
	3	40	151	1	999	0	1.1	93.994	-36.4	4.857	5191.0	1	0	1	
	4	56	307	1	999	0	1.1	93.994	-36.4	4.857	5191.0	1	0	0	
411	183	73	334	1	999	0	-1.1	94.767	-50.8	1.028	4963.6	1	1	0	
411	184	46	383	1	999	0	-1.1	94.767	-50.8	1.028	4963.6	1	0	0	
411	185	56	189	2	999	0	-1.1	94.767	-50.8	1.028	4963.6	1	1	0	
411	186	44	442	1	999	0	-1.1	94.767	-50.8	1.028	4963.6	1	0	0	
411	187	74	239	3	999	1	-1.1	94.767	-50.8	1.028	4963.6	1	1	0	
4118	88 го	ws ×	67 column	ıs											

5. The following code removes the the economic features and duration from the data as they would need to be forecasted with high precision to use as inputs in future predictions.

```
model_data = model_data.drop(
    ["duration", "emp.var.rate", "cons.price.idx", "cons.conf.idx", "euribor3m",
    "nr.employed"],
    axis=1,
)
```

Step 5: Split Data into training, validation and test

- 1. The following code splits the dataset into training (70%), validation (20%), and test (10%).
 - The first block of code shuffles model_data using the sample method with frac=1 (which means to sample the whole dataset) and a random seed of 1729. The shuffled data is then split into three segments.
 - For each of the datasets (train, validation, and test), the code reorders the columns so that the target column y_yes is the first column, and then it drops the y_no and y_yes columns from the rest. It then saves the reordered data to a CSV file. This is done for compatibility reasons as Amazon SageMaker's XGBoost algorithm expects the first column to be the target variable and the CSV should not include headers.

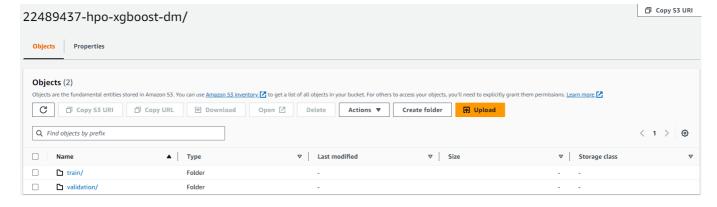
```
train_data, validation_data, test_data = np.split(
    model_data.sample(frac=1, random_state=1729),
    [int(0.7 * len(model_data)), int(0.9 * len(model_data))],
)

pd.concat([train_data["y_yes"], train_data.drop(["y_no", "y_yes"], axis=1)],
    axis=1).to_csv(
        "train.csv", index=False, header=False
)
pd.concat(
    [validation_data["y_yes"], validation_data.drop(["y_no", "y_yes"], axis=1)],
    axis=1
).to_csv("validation.csv", index=False, header=False)

pd.concat([test_data["y_yes"], test_data.drop(["y_no", "y_yes"], axis=1)],
    axis=1).to_csv(
        "test.csv", index=False, header=False
)
```

2. The newly created data is the uploaded to the S3 bucket we created earlier.

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



Step 6: Setup Hyperparameter Optimization

- 1. The following code sets up a configuration for a hyperparameter tuning job:
 - ParameterRanges: This specifies the hyperparameters you want to tune and their possible values.
 - CategoricalParameterRanges: This would contain parameters that have categorical values, but it's empty as we have transformed all the categorical variables into indicator variables in the previous steps.
 - ContinuousParameterRanges: This contains hyperparameters that can take any real value within a range.

- ResourceLimits: This part specifies how many training jobs can be created in total (MaxNumberOfTrainingJobs) and how many of them can run in parallel (MaxParallelTrainingJobs).
- Strategy: The strategy used for the tuning job. In this case, the "Bayesian" strategy is used, which
 is a popular method for hyperparameter optimization. It builds a probability model of the
 objective function and uses it to select the most promising hyperparameters to evaluate in the
 true objective function.
- HyperParameterTuningJobObjective: This is where you define the metric you want to optimize. The goal here is to maximize the "validation:auc" (Area Under the Curve for the validation set).

```
tuning_job_name = f"{student_id}-xgboost-tuningjob-01"
print(tuning_job_name)
tuning_job_config = {
    "ParameterRanges": {
        "CategoricalParameterRanges": [],
        "ContinuousParameterRanges": [
            {
                "MaxValue": "1",
                "MinValue": "0",
                "Name": "eta",
            },
            {
                "MaxValue": "10",
                "MinValue": "1",
                "Name": "min_child_weight",
            },
            {
                "MaxValue": "2",
                "MinValue": "0",
                "Name": "alpha",
            },
        "IntegerParameterRanges": [
            {
                "MaxValue": "10",
                "MinValue": "1",
                "Name": "max depth",
            }
        ],
    },
    "ResourceLimits": {"MaxNumberOfTrainingJobs": 2, "MaxParallelTrainingJobs":
2},
    "Strategy": "Bayesian",
    "HyperParameterTuningJobObjective": {"MetricName": "validation:auc", "Type":
"Maximize"},
}
```

2. The hyperparameter tuning job will launch training jobs to find an optimal configuration of hyperparameters. These training jobs should be configured using the SageMaker

CreateHyperParameterTuningJob API. To configure the training jobs, we define a JSON object and pass it as the value of the TrainingJobDefinition parameter inside CreateHyperParameterTuningJob.

- AlgorithmSpecification The registry path of the Docker image containing the training algorithm and related metadata. We fetch the XGboost Docker image for our specic usecase.
- InputDataConfig: This specifies where SageMaker should fetch the training and validation data from. For both data sources, the content type is CSV and data distribution is "FullyReplicated", which means each training instance will get the full dataset.
- OutputDataConfig: This defines where SageMaker should store the output of the training job (e.g., model artifacts). The output will be stored in the specified S3 bucket under an "output" prefix.
- ResourceConfig: Specifies the infrastructure to be used for the training job. In this case, a single ml.m5.xlarge instance with a volume size of 10GB.
- RoleArn: This is the Amazon Resource Name (ARN) of the IAM role that SageMaker can assume
 to perform tasks on your behalf. We make use of the ARN of the SageMakerRole that we
 retreived earlier.
- StaticHyperParameters: These are hyperparameters specific to the XGBoost algorithm that will remain fixed during training. For instance, objective is set to "binary:logistic", which is suitable for binary classification problems.
- StoppingCondition: This specifies that the training job should be terminated if it runs for more than 43200 seconds (or 12 hours).

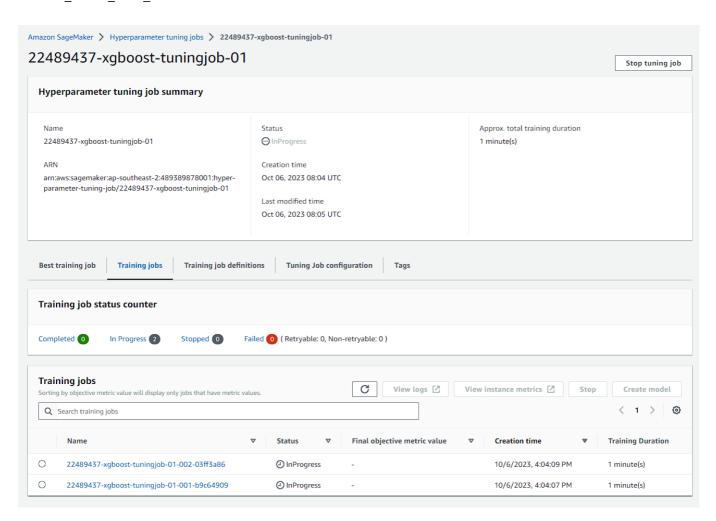
```
from sagemaker.image_uris import retrieve
# Use XGBoost algorithm for training
training_image = retrieve(framework="xgboost", region=region, version="latest")
s3_input_train = "s3://{}/train".format(bucket, prefix)
s3_input_validation = "s3://{}/validation/".format(bucket, prefix)
training job definition = {
    "AlgorithmSpecification": {"TrainingImage": training_image,
"TrainingInputMode": "File"},
    "InputDataConfig": [
        {
            "ChannelName": "train",
            "CompressionType": "None",
            "ContentType": "csv",
            "DataSource": {
                "S3DataSource": {
                    "S3DataDistributionType": "FullyReplicated",
                    "S3DataType": "S3Prefix",
                    "S3Uri": s3 input train,
                }
            },
        },
            "ChannelName": "validation",
            "CompressionType": "None",
            "ContentType": "csv",
            "DataSource": {
                "S3DataSource": {
```

```
"S3DataDistributionType": "FullyReplicated",
                    "S3DataType": "S3Prefix",
                    "S3Uri": s3_input_validation,
                }
            },
        },
    ],
    "OutputDataConfig": {"S3OutputPath": "s3://{}/output".format(bucket,
prefix)},
    "ResourceConfig": {"InstanceCount": 1, "InstanceType": "ml.m5.xlarge",
"VolumeSizeInGB": 10},
    "RoleArn": sagemaker_role,
    "StaticHyperParameters": {
        "eval_metric": "auc",
        "num_round": "1",
        "objective": "binary:logistic",
        "rate_drop": "0.3",
        "tweedie_variance_power": "1.4",
    "StoppingCondition": {"MaxRuntimeInSeconds": 43200},
}
```

3. This code snippet is launching a hyperparameter tuning job on Amazon SageMaker.

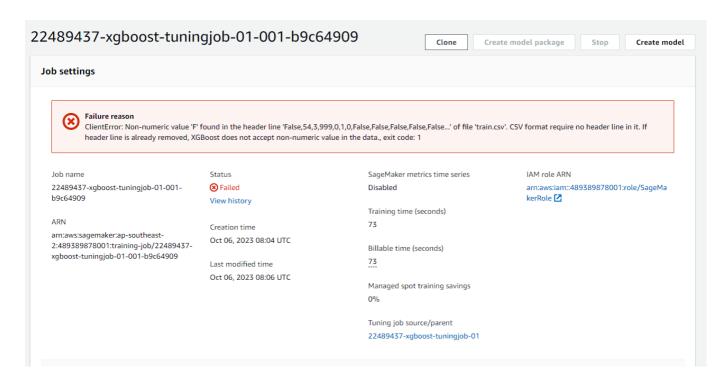
```
#Launch Hyperparameter Tuning Job
smclient.create_hyper_parameter_tuning_job(
    HyperParameterTuningJobName=tuning_job_name,
    HyperParameterTuningJobConfig=tuning_job_config,
    TrainingJobDefinition=training_job_definition,
)
```

4. We can log into the sagemaker service on the AWS console. Here we can see all the running training instances.



5. The hyperparameter optimization failed and we got the message "ClientError: Non-numeric value 'F' found in the header line 'False,54,3,999,0,1,0,False,False,False,False,False...' of file 'train.csv'". In order to resolve this, we convert all the True/False in the dataset to 1/0 using the following code.

```
bool_cols = model_data.select_dtypes(include=['bool']).columns
model_data[bool_cols] = model_data[bool_cols].astype(int)
```



6. Running the code again, we succeed this time and we can view the best training job hyperparaneters either in the AWS console or in the s3://YOUR_BUCKET/sagemaker/22489437-hpo-xgboost-dm/output.

