DSC 204A: Scalable Data Systems

Programming Assignment 0

Released: 10 April 2023, Due: 25 April 2023

VERY IMPORTANT: Download your progress to your local machine at regular intervals and terminate your instance when you decide to pause working. You have only \$50 for both PA0 and PA1 and so DO NOT leave instances running. If you terminate without downloading, you WILL LOSE all your work. Every time you start a new instance, you must download the dataset from S3 to your instance. Also, start only AWS Spot Instances and NOT On-Demand instances.

1 Introduction

The goal of this programming assignment is to get you comfortable with datasets that do not fit in the RAM of a single machine and hence are not suitable for analysis using packages like Pandas or NumPy. In PAO and PA1 you will be using the Dask library to explore secondary storage aware data access on a single machine. In this assignment, you will be learn to setup Dask on AWS and compute several descriptive statistics about the data to build intuitions for feature engineering for the final assignment.

2 Dataset Description

You are provided with the Amazon Reviews dataset with the *reviews* table as CSV file. The schema is provided in Table 1.

Column name	Column description	Example
reviewerID	ID of the reviewer	A32DT10X9WS4D0
asin	ID of the product	B003VX9DJM
reviewerName	name of the reviewer	Slade
helpful	helpfulness rating of the review	[0, 0]
reviewText	text of the review	this was a gift for my friend who loves touch lamps.
overall	rating of the product	1
summary	summary of the review	broken piece
unixReviewTime	summary of the review	1397174400
reviewTime	time of the review (raw)	04 11, 2014

Table 1: Schema of Reviews table

The helpful attribute is a tuple of two integer values. The first value represents the number of people who found the review helpful, and the second value represents the total number of people who voted.

3 Tasks

You will use the *reviews* table to explore features related to users. Your task is to create a users table with the schema given in Table 2.

A code stub with the function signature has been provided to you. The input to this function is the path to the reviews CSV file and you will be carrying out a series of transformations to produce the required users table as a DataFrame. Plug in the DataFrame you obtained as a result in <YOUR_USERS_DATAFRAME>. The last line converts the dataframe into a json file and writes it to results_PAO.json file. Do not remove this line. We will time the execution of the function PAO.

Column name	Column description
reviewerID (PRIMARY KEY)	ID of the reviewer
number_products_rated	Total number of products rated by the reviewer
avg_ratings	Average rating given by the reviewer across all the reviewed products
reviewing_since	The year in which the user gave their first review
helpful_votes	Total number of helpful votes received for the users' reviews
total_votes	Total number of votes received for the users' reviews

Table 2: Schema of users table

We have shared with you the "development" dataset and our accuracy results. Our code's runtime on 1 node is roughly 615s. You can use this to validate your results and debug your code. The final evaluation will happen on a separate held-out test set. The runtime will be different for the held-out test set.

4 Deliverables

Submit your source code as <NAME>_<PID>.py on Canvas. Your source code must conform to the function signatures provided to you. Make sure that your code is writing results to results_PAO.json.

5 Getting Started

- 1. Access your ETS account using single sign-on ID: https://ets-apps.ucsd.edu/individual/DSC204A_SP23_A00/. To open the AWS console click "Click here to access AWS" at the bottom of the page. To get your AWS credentials for CLI / API usage click "Generate API Keys (for CLI/scripting)".
- 2. We have setup the Dask environment on an AMI with the name "dsc204a-dask-environment-public." Go to "AMIs" (under "Images") in your EC2 dashboard, select private images, and then search by name to find it. Select this AMI. See Figure 1.

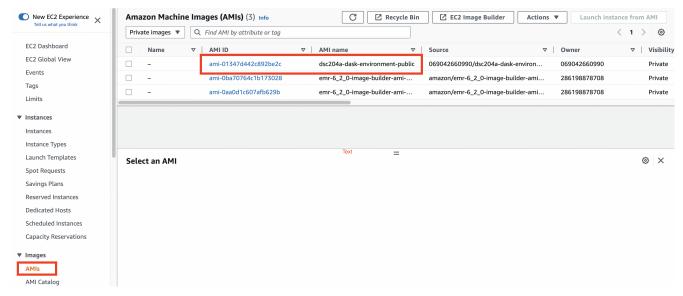


Figure 1

3. After selecting the AMI, click "Launch Instance from AMI" as shown below.

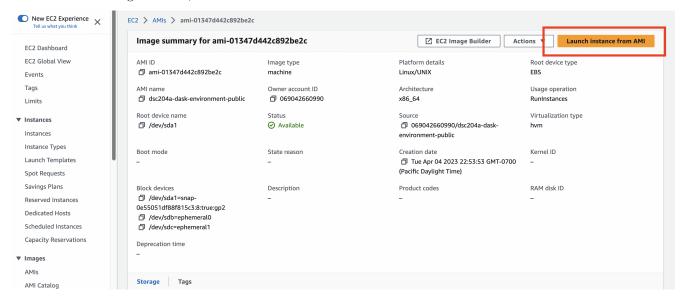


Figure 2

- 4. Now, strictly follow the below instructions to launch one EC2 Spot instance in which you will run your code.
 - **a.** Give any name for your instance.
 - **b.** Number of instances to launch is 1. The instance type is "t2.xlarge".
 - **c.** Create a new key pair for SSH'ing to your instance later. The private key will be downloaded to your local machine.
 - **d.** Let the network settings be same as default.
 - e. Choose 40GB SSD gp2 storage.
 - **f.** Open advanced details. Select "Request Spot Instances". Then click on "customize" just on the right. Open the dropdown for "Request type" and select "One-time".

 ${\bf g.}$ Click "Launch Instance" as shown in Figure 5.

See below figures for the required configuration.

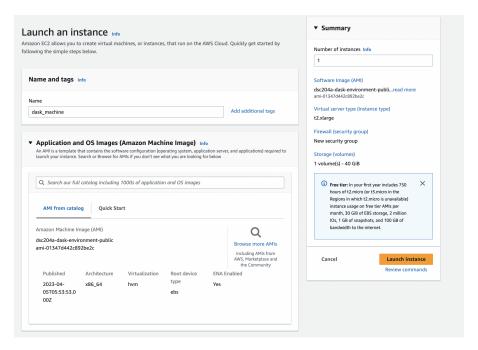


Figure 3

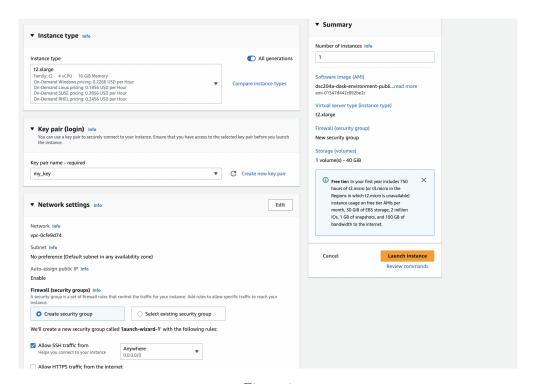


Figure 4

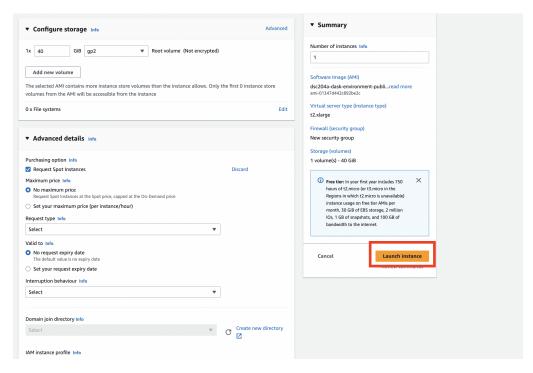


Figure 5

- 5. In these final steps you will SSH into your instance, download the dataset and start jupyter-notebook.
 - a. Change permission of the SSH keyfile to make sure your private key file isn't publicly viewable: chmod 400 <keyfilename>.pem
 - b. SSH into one of the nodes using command:
 ssh -i ''<your_key>.pem'' ubuntu@<ip-address-of-EC2-instance>
 Public IP of your instance can be found inside the instance details of your running instance.
 - c. I will use tmux to manage my terminals. You can use tmux or screen or whatever works for you, but tmux is recommended. Run tmux new -s dt to open a new window. We will use AWS CLI for downloading the dataset and code stub from our S3 bucket into our instance. First, export your AWS credentials (see step 1. on where you can find them) into the current shell environment. Then run aws s3 sync s3://dsc204a-public.

This will start the download.

c. Now, detach from your tmux session by pressing Ctrl B + D. Start another tmux session for running jupyter-notebook. In this session, first activate the Dask environment with the command:

source dask_env/bin/activate

Then, start jupyter notebook with the command: jupyter-notebook.

d. We will forward port 8888 on our AWS instance to our local machine so we can access jupyter notebook on our machine. Open a new terminal on your local machine and run this command -

ssh -i ''<your_key>.pem'' ubuntu@<ip-address-of-

EC2-instance> -L 8888:localhost:8888

Now, on your browser go to http://localhost:8888 where you will be prompted to enter a token. You can find this token in the terminal output where you started jupyter notebook.

e. Dask also provides us with a Dask Dashboard (like Tensorboard for those who've worked with DL) where we can see the progress of our tasks. This is automatically started on port 8787. So, we will forward this port as well. Run this command in yet another terminal on your local machine -

ssh -i ''<your_key>.pem'' ubuntu@<ip-address-of-

EC2-instance> -L 8787:localhost:8787

You can visit http://localhost:8787 but you will only see output once you have started the Dask Scheduler.

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