INTRODUCTION:

Census tracts are relatively permanent small-area geographic divisions of a county or statistically equivalent entity defined for the tabulation and presentation of data from the decennial census and selected other statistical programs

Census tracts defined by these criteria will also be used to tabulate and publish estimates from the American Community Survey (ACS) and potentially data from other Bureau of the Census (Census Bureau) censuses and surveys. In addition to census tracts, the Participant Statistical Areas Program (PSAP) encompasses the review and update of census block groups, census designated places, and census county divisions.

Census tracts generally have a population size between 1,200 and 8,000 people, with an optimum size of 4,000 people. A census tract usually covers a contiguous area; however, the spatial size of census tracts varies widely depending on the density of settlement. Census tract boundaries are delineated with the intention of being maintained over a long time so that statistical comparisons can be made from census to census. Census tracts occasionally are split due to population growth or merged as a result of substantial population decline.

IMPORTANCE OF CENCUS TRACT IN DATA ANALYSIS:

- .. census tracts are comprised on one or more coterminous block groups.
- .. on average, a census tract is comprised of three block groups.
- Census tracts are used by many Federal, state and local governments for compliance and program management

Census tracts are important for many reasons. It is easy to misidentify or misunderstand patterns and characteristics within cities, counties and metros which become obfuscated using these higher levels, more aggregate, geographies. Many cities and counties that might be experiencing demographic-economic decline will often have bright spots that are groups of a few or many census tracts.

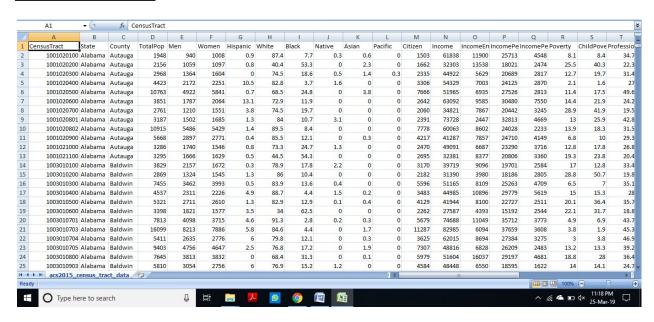
- Importance of Census Tracts for Data Analytics
 Census tracts are important for many reasons. A partial list of reasons is provided below.
 - Covering the U.S. wall-to-wall, census tracts are the preferred "small area" geography for superior data analytics.
 - The Census Bureau now produces annual tract demographic-economic data from the American Community Survey; there is an evolving timeseries at the tract level creating new analytical opportunities.
 - Originally developed to equivalence neighborhoods, many still do.
 - Defined by the Census Bureau in collaboration with local groups, tracts typically reflect boundaries meaningful for local area analysis.
 - Defined generally for use with each new decennial census, most tract boundaries are stable and non-changing for ten years and many much longer.
 - Designed to average 4,000 population, there are more than twice as many census tracts (73,056) than ZIP code areas (33,129).
 - Tract boundaries are well-defined; unlike ZIP code areas which are subject to multi-sourced geographic definitions.
 - Many data developers (e.g., epidemiologists) use census tract geography to tabulate their own small area data enabling more effective

use of those data with Census Bureau census tract data.

- As a statistical geographic area (in contrast to politically defined areas, census tracts are coterminous with counties; data at the census tract level can be aggregated to the county level.
- Small area estimates for tracts are typically more reliable than for block groups..

DATA SET DESCRIPTION:

Link for dataset:

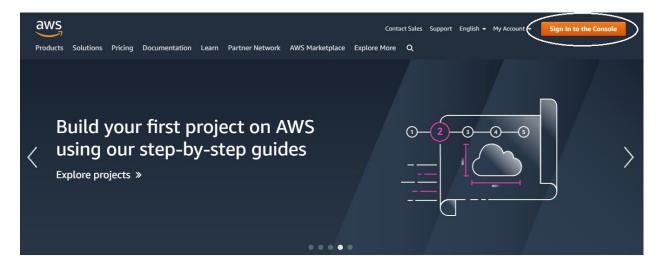


STEPS FOR SETTING UP OF CLOUD ENVIRONMENT:

1. AUTHENTICATION:

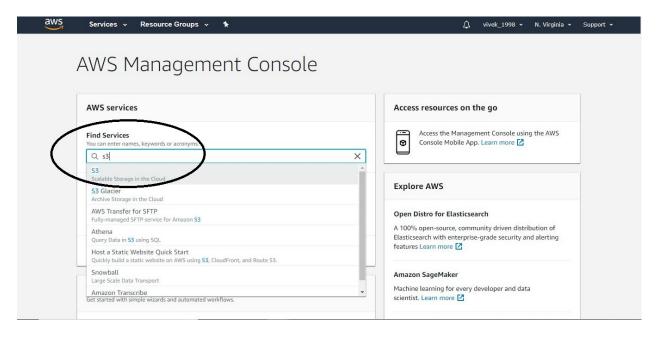
STEP 1.1: create a account at https://aws.amazon.com/

STEP 1.2: Sign into your AWS Amazon console using your credentials

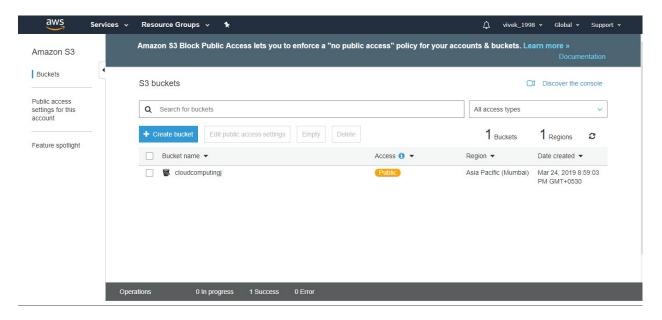


2. CONSOLE (OPENING S3)

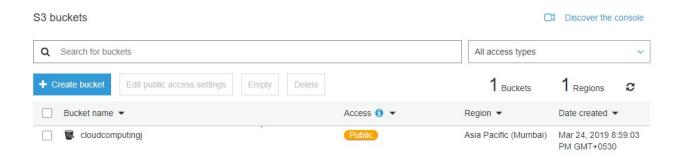
STEP 2.1: Search for S3 storage and open it



RESULT: This screen will be displayed



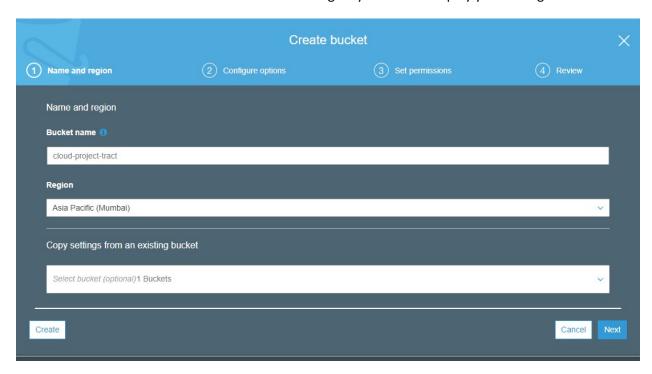
BUCKET INFORMATION WILL BE DISPLAYED HERE



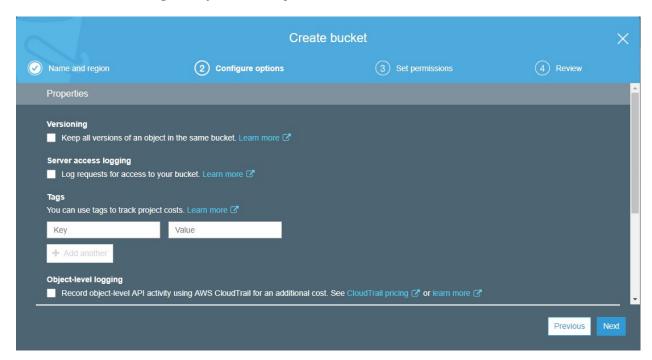
3. CREATING OF BUCKET WITH PUBLIC ACCESS

STEP 3.1: Click on create button option

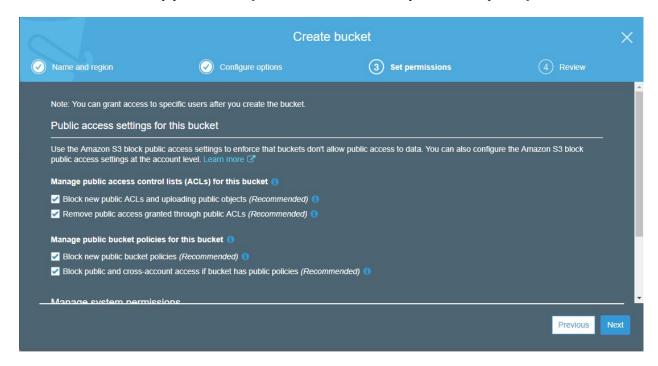
STEP 3.2: Fill details like name and select region you want to deploy your Storage.



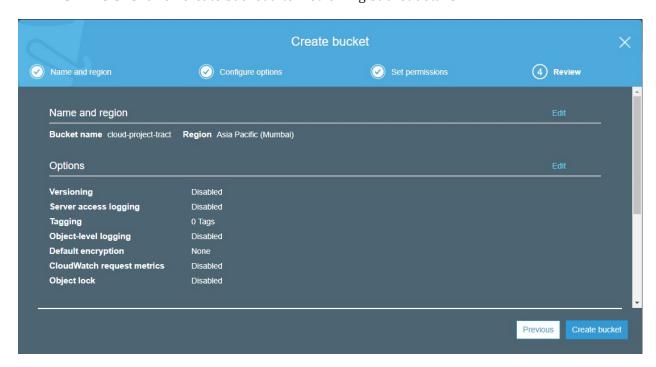
STEP 3.3: Configure key value if required



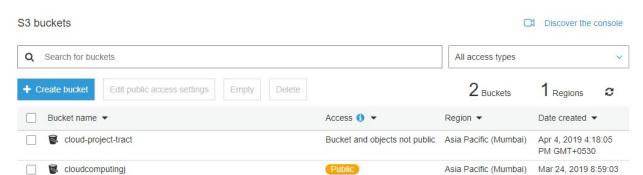
STEP 3.4: Set up permissions(these will allow to make your bucket public)



STEP 3.5: Click on create bucket after reviewing bucket details



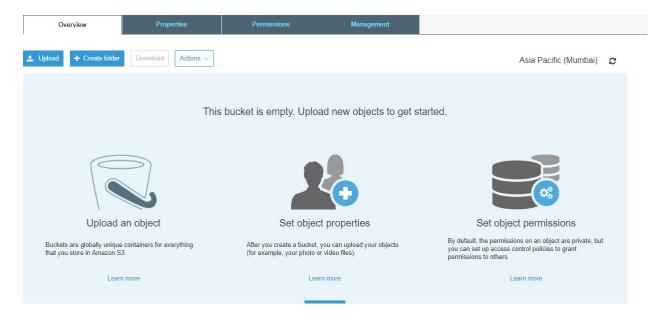
RESULT:



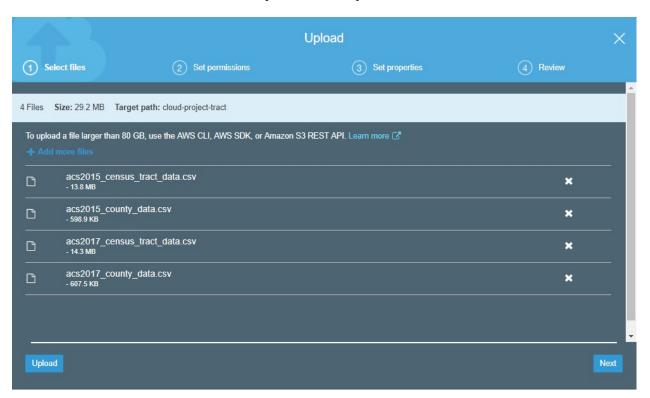
PM GMT+0530

4. UPLOADING OBJECTS TO S3:

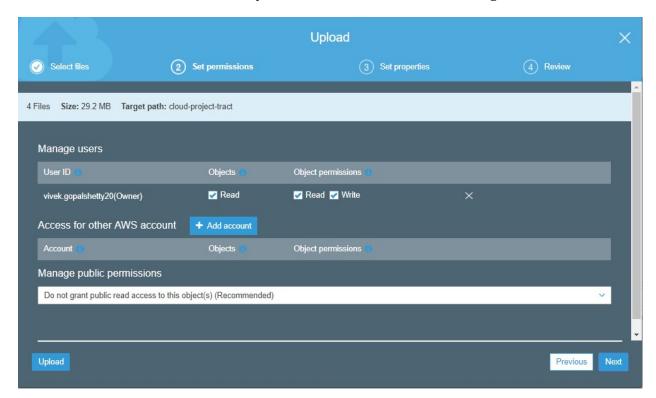
STEP 4.1: Click on upload



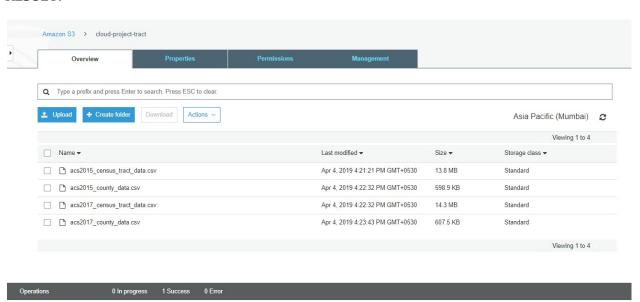
STEP 4.2: Select files from local system and drop them into bucket



STEP 4.3: While collaboration you can add other AWS accounts and give access to them.



RESULT:

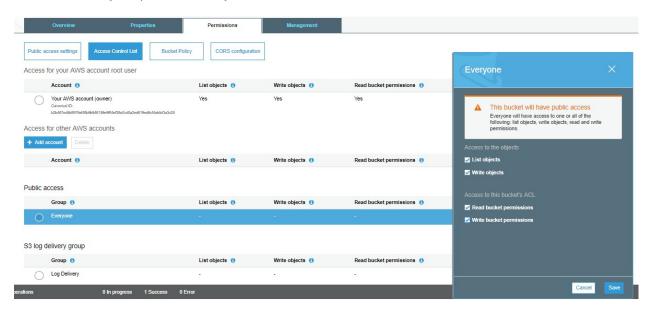


5. MAKING BUKCET PUBLIC

Important: Make sure your initial configuration is as shown in picture.



STEP 5: Select bucket and then go to permissions and then click on Public access-->everyone and then tick mark (read/write access) and confirm.



RESULT:



SETTING UP OF CODING ENVIRONMENT:

We have done data processing with data from AWS S3.

A small rest API has been provided to you. We have implemented one function that API needs to call to find the average census tract population size for a given city.

Setup

Before you start this lab you should make a new virtual environment in this directory and start it. Next, notice that there is a file called `requirements.txt`. This is a format that pip understands and that contains all the python packages needed for this project. You can install them with `pip install -r requirements.txt`

The REST API

'Georgia',

'Hawaii',

This API only has two endpoints. The / endpoint just returns a string saying "Welcome to cloud computing project!"

The second endpoint is documented as follows

```
'/average_pop?city`

City: name of a US city. Can be one of
' Alaska',
'Arizona',
'Arkansas',
'California',
'Colorado',
'Connecticut',
'Delaware',
'District of Columbia',
'Florida',
```



```
'Rhode Island',
'South Carolina',
'South Dakota',
'Tennessee',
'Texas',
'Utah',
'Vermont',
'Virginia',
'Washington',
'West Virginia',
'Wisconsin',
'Wyoming'
```

Virtual Environments

Virtual environments are a popular tool with python developers for self containing dependencies within a directory. You simply make a new virtual environment, and anything you install or remove with pip will only apply to that directory, not your global pip packages.

To make a new virtual environment: `virtualenv ./project_directory`

To start that environment: _**On Mac/Linux**_`source ./project_directory/bin/activate` _**On windows**_`.\project_directory\Source\activate`

To stop that environment: _**On Mac/Linux**_`source ./project_directory/bin/deactivate` _**On Windows**_`.\project_directory\Source\activate`

Testing Code

To test your code, we have written a small program in `test.py` which just calls the `average_pop` endpoint for every city and prints out the results.

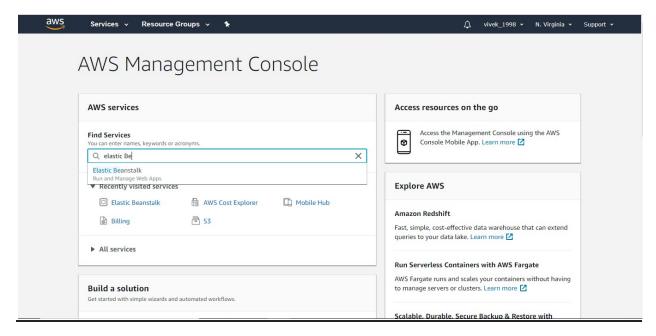
<u>s</u>The goal is to set up for Elastic Beanstalk, where we will run this API on a cluster of multiple machines to increase performance.

STEPS FOR SETTING UP OF BEANSTALK:

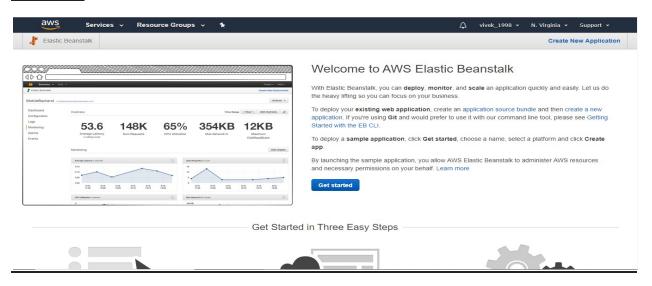
1. AUTHENTICATION:

- **STEP 1.1:** create a account at https://aws.amazon.com/
- STEP 1.2: Sign into your AWS Amazon console using your credentials
- 2. CONSOLE (OPENING Elastic Beanstalk)

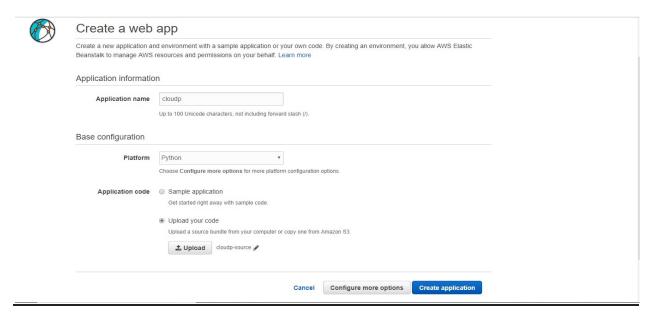
STEP 2.1: Search for Elastic Beanstalk and open it



RESULT:



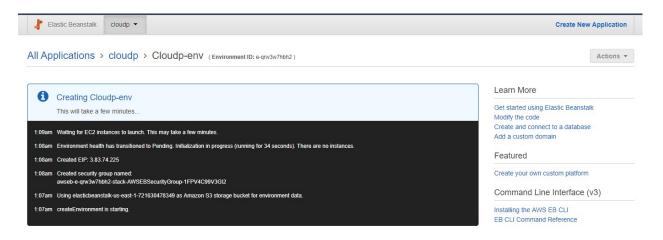
4. Now click on create and application and fill details as in image exactly



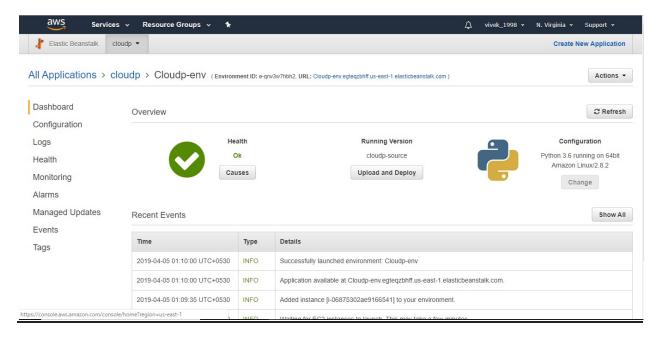
MAKE ZIP FILE OF IT AND UPLOAD IN APPLICATION CODE



RESULT: wait for 5-10 mins for your web app to deploy on cloud



RESULT: you can check health and status also



ACTION: Use link given at URL in test.py now to get average population



FLASK END TO END API:

```
from flask import Flask, request
import processing as p
app = Flask(__name__)
@app.route("/")
def index():
  return "Welcome to our cloud computing project!"
# /average_pop?city
@app.route("/average_pop")
def process():
  cityname = request.args.get('city')
  if cityname == None:
    return "No city name"
  avg = p.get_avg_pop(cityname)
  return str(avg)
if __name__ == "__main__":
  app.run()
```

OUTPUT:

ON WEB APP:



Welcome to our cloud computing project!

PROCESSING:

```
import boto3 as aws
from botocore.handlers import disable signing
import pandas as pd
import sys
if sys.version info[0] < 3:
  from StringIO import StringIO
                                                   # Check for python version to use correct
library
else:
  from io import StringIO
def get census data(city):
                                #given city name extract that data from AWS S3 storage
  filename = "acs2015.csv"
  s3 = aws.resource('s3')
  s3.meta.client.meta.events.register('choose-signer.s3.*', disable signing) #disable signing
  obj = s3.Object(bucket name="cloudcomputingj", key=filename)
                                                                      #connecting to our
bucket
  res = obj.get()
  raw data = res['Body'].read().decode('UTF-8') #decode data once fetched from server
  data = StringIO(raw data)
  df = pd.read csv(data)
  return df
                 #return dataFrame required
def get avg pop(city):
                           #given a city name finding average population of state using cencus
tract
  df = get census data(city) # Makes a call to get the city data from S3
  condition=(df['State']==city)
  df=df[condition]
  # Calculate and return the average census tract population
  total pop=df['TotalPop'].sum(axis=0)
  num tracts=df.shape[0]
  return str(round(total pop / num tracts,2))
```

```
TEST (input place)
import requests as r
import threading
cities = ['Alabama']
'Alaska',
'Arizona',
'Arkansas',
'California',
'Colorado',
'Connecticut',
'Delaware',
'District of Columbia',
'Florida',
'Georgia',
'Hawaii',
'Idaho',
'Illinois',
'Indiana',
'Iowa',
'Kansas',
'Kentucky',
'Louisiana',
'Maine',
'Maryland',
'Massachusetts',
'Michigan',
'Minnesota',
'Mississippi',
'Missouri',
'Montana',
'Nebraska',
'Nevada',
'New Hampshire',
'New Jersey',
'New Mexico',
'New York',
'North Carolina',
'North Dakota',
```

```
'Ohio',
'Oklahoma',
'Oregon',
'Pennsylvania',
'Puerto Rico',
'Rhode Island',
'South Carolina',
'South Dakota',
'Tennessee',
'Texas',
'Utah'.
'Vermont',
'Virginia',
'Washington',
'West Virginia',
'Wisconsin',
'Wyoming']
def get avg(c):
  res = r.get('http://localhost:5000/average pop?city=' + c)
  print(c + " average population: " + res.text)
for c in cities:
  threading.Thread(target=get avg, args=(c,)).start()
 *Python 3.7.2 Shell*
                                                                                      X
 File Edit Shell Debug Options Window Help
 Python 3.7.2 (tags/v3.7.2:9a3ffc0492, Dec 23 2018, 22:20:52) [MSC v.1916 32 bit
```

Type "help", "copyright", "credits" or "license()" for more information.

======== RESTART: C:\Users\vivek\Desktop\cll\test.py ===========

(Intel)] on win32

>>> Alabama average population: 4090.28

```
*Python 3.7.2 Shell*
                                                                        ×
File Edit Shell Debug Options Window Help
Python 3.7.2 (tags/v3.7.2:9a3ffc0492, Dec 23 2018, 22:20:52) [MSC v.1916 32 bit
(Intel)] on win32
Type "help", "copyright", "credits" or "license()" for more information.
======= RESTART: C:\Users\vivek\Desktop\cll\test.py ========
>>> Alabama average population: 4090.28
Delaware average population: 4249.79
Alaska average population: 4391.47
Arkansas average population: 4312.26
Arizona average population: 4352.51
District of Columbia average population: 3617.23
California average population: 4768.71
Colorado average population: 4226.51
```

ON SERVER REQUESTS

```
127.0.0.1 - - [10/Apr/2019 01:54:08] "GET /average_pop?city=AlabamaAlaska HTTP/1.1" 200 - 127.0.0.1 - - [10/Apr/2019 01:56:22] "GET /average_pop?city=Arkansas HTTP/1.1" 200 - 127.0.0.1 - - [10/Apr/2019 01:57:34] "GET /average_pop?city=Alabama HTTP/1.1" 200 - 127.0.0.1 - - [10/Apr/2019 01:58:37] "GET /average_pop?city=Delaware HTTP/1.1" 200 - 127.0.0.1 - - [10/Apr/2019 01:59:41] "GET /average_pop?city=Alaska HTTP/1.1" 200 - 127.0.0.1 - - [10/Apr/2019 02:00:29] "GET /average_pop?city=Arkansas HTTP/1.1" 200 - 127.0.0.1 - - [10/Apr/2019 02:01:04] "GET /average_pop?city=Arizona HTTP/1.1" 200 - 127.0.0.1 - - [10/Apr/2019 02:01:47] "GET /average_pop?city=District%20of%20Columbia HTTP/1.1" 200 - 127.0.0.1 - - [10/Apr/2019 02:03:42] "GET /average_pop?city=California HTTP/1.1" 200 - 127.0.0.1 - - [10/Apr/2019 02:03:42] "GET /average_pop?city=Colorado HTTP/1.1" 200 - 127.0.0.1 - - [10/Apr/2019 02:05:54] "GET /average_pop?city=Hawaii HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Hawaii HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:05:54] "GET /average_pop?city=Florida HTTP/1.1" 200 - 127.0.0.1 - [10/Apr/2019 02:0
```

USING DEPLOYED WEB APP FROM AWS SERVER

NOTE: Make no changes to application.py or processing .py In test.py change url given to you on AWS console of webapp

TEST.py

```
import requests as r
import threading
cities = ['Alabama',
'Alaska',
'Arizona',
'Arkansas',
'California',
'Colorado',
'Connecticut',
'Delaware',
'District of Columbia',
'Florida',
'Georgia',
'Hawaii',
'Idaho',
'Illinois',
'Indiana',
'Iowa',
'Kansas',
'Kentucky',
'Louisiana',
'Maine',
'Maryland',
'Massachusetts',
'Michigan',
'Minnesota',
'Mississippi',
'Missouri',
'Montana',
'Nebraska',
'Nevada',
```

'New Hampshire',

```
'New Jersey',
'New Mexico',
'New York',
'North Carolina',
'North Dakota',
'Ohio',
'Oklahoma',
'Oregon',
'Pennsylvania',
'Puerto Rico',
'Rhode Island',
'South Carolina',
'South Dakota',
'Tennessee',
'Texas',
'Utah',
'Vermont',
'Virginia',
'Washington',
'West Virginia',
'Wisconsin',
'Wyoming']
link="" #link given to you on aws webapp console.
def get avg(c):
  res = r.get('http://link)
  print(c + " average population: " + res.text)
for c in cities:
  threading.Thread(target=get avg, args=(c,)).start()
```

THREADING AS A ENHANCER:

Usage of asynchronous requests enhances the cloud request process by usage of the threading process.

An synchronous request doesn't work on the "FIRST COME FIRST SERVE BASIS". It identifies the faster networks and caters to their requests first to reduce load on the server by sending outputs as soon as the outputs are generated.

Furthermore, the multi-threading process speeds up the working of the system. A multi-threaded application is an application whose architecture takes advantage of the multi-threading provided by the operating system. Such applications assign specific jobs to individual threads within the process and the threads communicate, using various means, to synchronize their actions. For example, a data processing application might be designed so that the graphical user interface is completely handled by one thread, while the actual work of the application is performed by another thread. This architecture would enable the complete separation of the business logic from the user-interface logic within the application.

CONCLUSION:

The census tracts calculated above simply the population based analysis process to great extent.

The easy and statistically accurate approaches used lead to solid and reliable reports which can be used in the development of a nation as well as the welfare of the people.

Cloud computing being the backbone of all future Information Technology Endeavours enhances the above process to extents of utter ease. The data and results being calculated and communicated using the cloud the technologies takes us one step closer to fully digitised nations, data and accounts.

The usage of this tutorial for the educating the people about cloud computing and data analysis together can generate curiosity in young minds and build towards a brighter future.