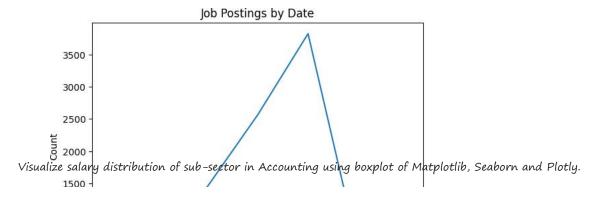
# [Lab 5]

Name: Nguyễn Thành Trung – MSSV: 19522431

Link github: Data Mining/Week5 at main · Shu2301/Data Mining (github.com)

```
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files under the input directory
import os
for dirname, _, filenames in os.walk("_/kaggle/input"):
    for filename in filenames:
        print(os.path.join(dirname, filename))
# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version using "Save & # You can
also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session
In this workshop, you will continue using Matplotlib, Seaborn and Plotly to explore the data. Filter from the dataset all the
"Accounting" job and visualize the total job postings of each sub-sectors.
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import plotly.express as px
df = pd.read_csv("_/kaggle/input/shu/job-market.csv")
accounting_jobs = df[df['Classification'] == 'Accounting']
subsector_counts = accounting_jobs['SubClassification'].value_counts()
fig = px.bar(x=subsector_counts.index, y=subsector_counts.values, labels={'x':'Sub-sector', 'y':'Count'}, color=subsector_counts.index)
fig.update_layout(title='Accounting')
fig.show()
Count the job postings by month and visualize in line graph using Matplotlib, Seaborn and Plotly
import matplotlib.pyplot as plt
# Convert the 'Date' column to datetime format
df['Date'] = pd.to_datetime(df['Date'])
# Filter the data for the month of October 2018
df = df[(df['Date'] >= '2018-10-01') & (df['Date'] <= '2018-10-31')]
# Group the data by day and count the job postings
job_postings_by_day = df.groupby(pd.Grouper(key='Date', freq='D'))['Title'].count()
# Create a line graph of the job postings by day using Matplotlib
plt.plot(job_postings_by_day.index, job_postings_by_day.values)
plt.xticks(job_postings_by_day.index, [d.strftime('%d.%b') for d in job_postings_by_day.index], rotation=45)
plt.xlabel('Label')
plt.ylabel('Count')
plt.title('Job Postings by Date')
plt.show()
```



import seaborn as sns

```
accounting_jobs = df[df['Classification'] == 'Accounting']
plt.figure(figsize=(10,8))
sns.boxplot(x=accounting_jobs['SubClassification'], y=accounting_jobs['LowestSalary'])
plt.xlabel('Sub-sector')
plt.ylabel('Average Salary')
plt.title('The average salary distribution of sub-sectors in Accounting')
plt.xticks(rotation=90)
plt.show()
```

A NumPy version >=1.16.5 and <1.23.0 is required for this version of SciPy (detected version 1.23.5

## The average salary distribution of sub-sectors in Accounting

I = I

```
Section 2: Moving Beyond Static Visualizations
```

- 1 Animating cumulative values over time import pandas as pd questions\_per\_library = pd.read\_csv( "/kaggle/input/shu/stackoverflow/stackoverflow.csv", parse\_dates=True, index\_col = 'creation\_date' ).loc[:,'pandas':'bokeh'].resample('1M').sum().cumsum().reindex( pd.date\_range('2008-08', '2021-10', freq='M')

	pandas	matplotlib	numpy	seaborn	geopandas	geoviews	altair	yellowbrick	vega	holoviews	hvplot	bokeh
2021-05-31	200734.0	57853.0	89812.0	6855.0	1456.0	57.0	716.0	46.0	532.0	513.0	84.0	4270.0
2021-06-30	205065.0	58602.0	91026.0	7021.0	1522.0	57.0	760.0	48.0	557.0	521.0	88.0	4308.0
2021-07-31	209235.0	59428.0	92254.0	7174.0	1579.0	62.0	781.0	50.0	572.0	528.0	89.0	4341.0
2021-08-31	213410.0	60250.0	93349.0	7344.0	1631.0	62.0	797.0	52.0	589.0	541.0	92.0	4372.0
2021-09-30	214919.0	60554.0	93797.0	7414.0	1652.0	63.0	804.0	54.0	598.0	542.0	92.0	4386.0

from matplotlib.animation import FuncAnimation

import matplotlib.pyplot as plt from matplotlib import ticker

).fillna(0)

questions\_per\_library.tail()

```
def bar_plot(data):
 fig, ax = plt.subplots(figsize=(8, 6))
 sort_order = data.last('1M').squeeze().sort_values().index
 bars = [
      bar.set_label(label) for label, bar in
      zip(sort_order, ax.barh(sort_order, [0] * data.shape[1]))
 ]
  ax.set_xlabel('total questions', fontweight='bold')
 ax.set_xlim(0, 250_000)
  ax.xaxis.set_major_formatter(ticker.EngFormatter())
  ax.xaxis.set_tick_params(labelsize=12)
  ax.yaxis.set_tick_params(labelsize=12)
  for spine in ['top', 'right']:
   ax.spines[spine].set_visible(False)
 fig.tight_layout()
 return fig, ax
%config InlineBackend.figure_formats = ['svg']
%matplotlib inline
bar_plot(questions_per_library)
```

```
(<Figure size 800x600 with 1 Axes>, <AxesSubplot: xlabel='total questions'>)
          pandas
          numpy
       matplotlib
         seaborn
           bokeh
      geopandas
            altair
             vega -
def generate_plot_text(ax):
 annotations =
     [ ax.annotate(
          "", xy=(0, bar.get_y() + bar.get_height()/2),
         ha='left', va='center'
     ) for bar in ax.patches
 ]
 time_text = ax.text(
     0.9, 0.1, **, transform = ax.transAxes,
     fontsize=15, ha='center', va='center'
 return annotations, time_text
def update(frame, *, ax, df, annotations, time_text):
 data = df.loc[frame, :]
 # update bars
 for rect, text in zip(ax.patches, annotations):
   col = rect.get_label()
   if data[col]:
     rect.set width(data[col])
     text.set_x(data[col])
     text.set_text(f" {data[col]:,.Of}")
 # update time
 time\_text.set\_text(frame.strftime('\%b\n\%Y'))
from pandas.core.strings.base import annotations
from functools import partial
def bar_plot_init(questions_per_library):
 fig, ax = bar_plot(questions_per_library)
 annotations, time_text = generate_plot_text(ax)
 bar_plot_update = partial(
     update, ax=ax, df=questions_per_library,
     annotations = annotations, time_text = time_text
 )
 return fig, bar_plot_update
from numpy import repeat
fig, update_func = bar_plot_init(questions_per_library)
ani = FuncAnimation(
   fig, update_func, frames=questions_per_library.index, repeat=False
ani.save(
    'stackoverflow_questions.mp4',
   writer='ffmpeg', fps=10, bitrate=100, dpi=300
plt.close()
```

```
from IPython import display

display.Video(
    '/kaggle/working/stackoverflow_questions.mp4', width=600, height=400, embed=True, html_attributes='controls muted autoplay'
)
```

0:15 / 0:15

## Animating distributions over time

	Entries				Exits			
Borough	Bk	Вх	M	Q	Bk	Вх	M	Q
Datetime								
2017-02-04	617650.0	247539.0	1390496.0	408736.0	417449.0	148237.0	1225689.0	279699.0
2017-02-05	542667.0	199078.0	1232537.0	339716.0	405607.0	139856.0	1033610.0	268626.0
2017-02-06	1184916.0	472846.0	2774016.0	787206.0	761166.0	267991.0	2240027.0	537780.0
2017-02-07	1192638.0	470573.0	2892462.0	790557.0	763653.0	270007.0	2325024.0	544828.0
2017-02-08	1243658.0	497412.0	2998897.0	825679.0	788356.0	275695.0	2389534.0	559639.0

```
_, config['hist'] = ax.hist(
            data[config['mask']].loc[date_range], bin_ranges, ec='black'
        ax.xaxis.set_major_formatter(ticker.EngFormatter())
        ax.set(
            xlim=(0, None), ylim=(0, config['ymax']),
            xlabel=f"{config["label"]} Entries'
        for spine in ['top', 'right']:
            ax.spines[spine].set_visible(False)
    axes[0].set_ylabel('Frequency')
    fig.suptitle('Histogram of Daily Subway Entries in Manhattan')
   fig.tight_layout()
    return fig, axes, bin_ranges, configs
 = subway_histogram(manhattan_entries, bins=30, date_range='2017')
                             Histogram of Daily Subway Entries in Manhattan
        60
                                                         120
        50
                                                         100
        40
                                                          80
      Frequency
        30
                                                          60
        20
                                                          40
        10
                                                          20
         0
                                                           0 -
                                                                        1 M 1.5 M 2 M 2.5 M
Weekday Entries
                500 k
                                          2.5 M
                                                                  500 k
def add_time_text(ax):
   time_text = ax.text(
        0.15, 0.9, ", transform=ax.transAxes,
        fontsize=15, ha='center', va='center'
   )
    return time_text
def update(frame, *, data, configs, time_text, bin_ranges):
    artists = []
   time = frame.strftime('%b\n%Y')
   if time != time_text.get_text():
        time_text.set_text(time)
        artists.append(time_text)
   for config in configs:
        time_frame_mask = \
            (data.index > frame - pd.Timedelta(days=365)) & (data.index <= frame)
        counts, = np.histogram(
            data[time_frame_mask & config['mask']],
            bin ranges
        for count, rect in zip(counts, config['hist'].patches):
            if count != rect.get_height():
                rect.set_height(count)
                artists.append(rect)
    return artists
def histogram init(data, bins, initial date range):
   fig, axes, bin_ranges, configs = subway_histogram(data, bins, initial_date_range)
    update_func = partial(
        update, data=data, configs=configs,
```

```
time_text=add_time_text(axes[0]),
        bin_ranges=bin_ranges
   return fig, update_func
fig, update_func = histogram_init(
   manhattan_entries, bins=30, initial_date_range=slice('2017', '2019-07')
)
ani = FuncAnimation(
   fig, update_func, frames=manhattan_entries['2019-08':'2021'].index,
   repeat=False, blit=True
)
ani.save(
    'subway_entries_subplots.mp4',
   writer='ffmpeg', fps=30, bitrate=500, dpi=300
plt.close()
from IPython import display
display.Video(
    '/kaggle/working/subway_entries_subplots.mp4', width=600, height=400,
   embed=True, html_attributes='controls muted autoplay'
)
```

0:13 / 0:18

## Animating geospatial data with HoloViz

```
geometry mag time month
```

**0** POINT Z (-67.12750 19.21750 12.00000) 2.75 2020-01-01 00:01:56.590

```
import geoviews as gv
import geoviews.feature as gf
import holoviews as hv
```

gv.extension('matplotlib')



import calendar

```
def plot_earthquakes(data, month_num):
    points = gv.Points(
        data.query(fmonth == {month_num}*),
        kdims = ['longitude', 'latitude'],
        vdims = ['mag']
    ).redim.range(mag=(-2, 5), latitude=(-45,45))

    overlay = gf.land * gf.coastline * gf.borders * points

    return overlay.opts(
        gv.opts.Points(color='mag', cmap='fire_r', colorbar=True, alpha=0.75),
        gv.opts.Overlay(
            global_extent=False, title=f*{calendar.month_name[month_num]}*, fontscale=2
        )
    )

plot_earthquakes(earthquakes, 1).opts(
    fig_inches=(6,3), aspect=2, fig_size=250, fig_bounds=(0.07, 0.05, 0.87, 0.95)
```

/opt/conda/lib/python3.10/site-packages/cartopy/io/\_init\_.py:241: DownloadWarning:

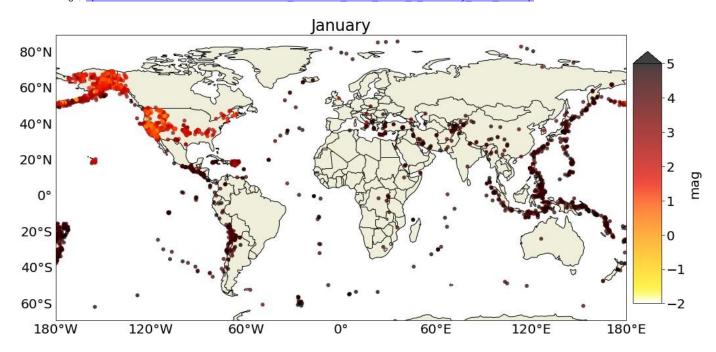
Downloading: https://naturalearth.s3.amazonaws.com/110m\_physical/ne\_110m\_land.zip

/opt/conda/lib/python3.10/site-packages/cartopy/io/\_init\_.py:241: DownloadWarning:

Downloading: https://naturalearth.s3.amazonaws.com/110m\_physical/ne\_110m\_coastline.zip

/opt/conda/lib/python3.10/site-packages/cartopy/io/\_init\_.py:241: DownloadWarning:

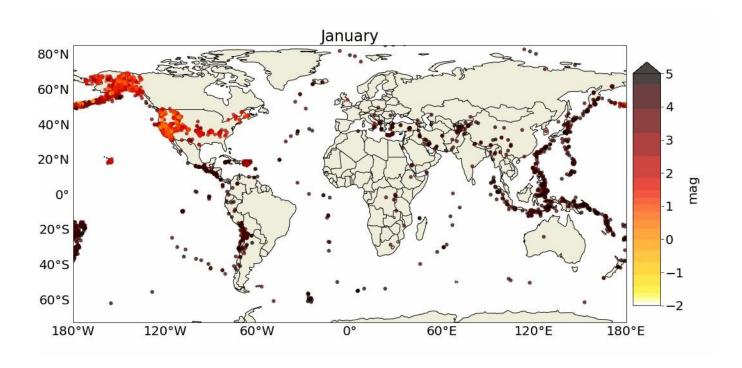
Downloading: https://naturalearth.s3.amazonaws.com/110m\_cultural/ne\_110m\_admin\_0\_boundary\_lines\_land.zip



```
frames = {
    month_num: plot_earthquakes(earthquakes, month_num)
```

```
for month_num in range(1, 13)
}
holomap = hv.HoloMap(frames)

hv.output(
    holomap.opts(
        fig_inches=(6,3), aspect=2, fig_size=250,
        fig_bounds=(0.07, 0.05, 0.87, 0.95)
), holomap='gif', fps = 5
```



```
hv.save(
    holomap.opts(
        fig_inches=(6,3), aspect=2, fig_size=250,
        fig_bounds=(0.07, 0.05, 0.87, 0.95)
    ), 'earthquakes.gif', fps=5
)
```

Up Next: Building interactive Visualizations for Data

ExplorationExcercises Excercises

- 1. Modify the animation of subway entries from this section to show both the weekday and weekend histograms on the same subplot
- 2. Modify the earthquake animation to show earthquakes per day in April 2020.

Section 3: Building interactive visualizations for data exploration

import geopandas as gpd

earthquakes =

gpd.read\_file('/kaggle/input/shu/earthquakes.geojson').assign( time = lambda x: pd.to\_datetime(x.time, unit='ms'), month = lambda x: x.time.dt.month

Data'

earthquakes.head()

```
mag
                                       place
                                                                 time tsunami magType
                                                                                                                        geometry month
0 2.75
                80 km N of Isabela, Puerto Rico 2020-01-01 00:01:56.590
                                                                                      md POINT Z (-67.12750 19.21750 12.00000)
1 2.55
                64 km N of Isabela, Puerto Rico 2020-01-01 00:03:38.210
                                                                                            POINT Z (-67.09010 19.07660 6.00000)
                                                                                      md
2 1.81 12 km SSE of Maria Antonia, Puerto Rico 2020-01-01 00:05:09.440
                                                                              0
                                                                                            POINT Z (-66.85410 17.87050 6.00000)
3 1.84
        9 km SSE of Maria Antonia, Puerto Rico 2020-01-01 00:05:36.930
                                                                                            POINT Z (-66.86360 17.89930 8.00000)
        8 km SSE of Maria Antonia, Puerto Rico 2020-01-01 00:09:20.060
4 1.64
                                                                              0
                                                                                      md
                                                                                            POINT Z (-66.86850 17.90660 8.00000)
                                                                                                                                       1
```

```
from cartopy import crs
import geoviews as gv
import geoviews.feature as gf
gv.extension('bokeh')
points =
   gv.Points( earthqua
    kdims = ['longitude', 'latitude'],
    vdims = ['month', 'place', 'tsunami', 'mag', 'magType']
)
points = points.redim.range(
   mag=(-2,10), longitude=(-180,180), latitude=(-90,90)
)
overlay = gf.land * gf.coastline * gf.borders * points.groupby('month')
interactive_map = overlay.opts(
    gv.opts.Feature(projection=crs.PlateCarree()),
    gv.opts.Overlay(width=700, height=450),
   gv.opts.Points(color='mag', cmap='fire_r', colorbar=True, tools=['hover'])
)
import panel as pn
earthquake viz = pn.panel(interactive map, widget location='bottom')
earthquake_viz.embed()
Linking plots
january_earthquakes = earthquakes.query('month ==
    1').assign( longitude = lambda x: x.geometry.x,
    latitude = lambda x: x.geometry.y
).drop(columns=['month', 'geometry'])
import hvplot.pandas
geo = january_earthquakes.hvplot(
    x='longitude', y='latitude', kind='points',
    color='mag', cmap='fire_r', clim=(-2,10),
    titles='CartoLight', geo=True, global_extent=True,
    xlabel='Longitude', ylabel='Latitude', title='January 2020 Earthquakes',
    frame_height=450
)
     WARNING:param.main: titles option not found for points plot with bokeh; similar options include: ['title', 'tiles']
table = january_earthquakes.sort_values(['longitude',
    'latitude']).hvplot(kind='table', width=650, height=450, title='Raw
```

```
layout = geo + table

selection = hv.link_selections.instance()
map_and_table_tabs = selection(layout).opts(tabs=True)
```

map\_and\_table\_tabs

selection.filter(january\_earthquakes).nlargest(3, 'mag')

	mag	place	time	tsunami	magType	longitude	latitude
15154	7.7	123 km NNW of Lucea, Jamaica	2020-01-28 19:10:24.918	1	mww	-78.7560	19.4193
13296	6.7	13 km N of Do?anyol, Turkey	2020-01-24 17:55:14.147	0	mww	39.0609	38.4312
4062	6.4	13 km SSE of Maria Antonia, Puerto Rico	2020-01-07 08:24:25.262	1	mww	-66.8266	17.8686

## Additional plot types

import pandas as pd import numpy as np

```
flight_stats = pd.read_csv(
    * /kaggle/input/shu/T100 MARKET ALL CARRIER/865214564 T T100 MARKET ALL CARRIER.csv*,
   usecols=[
       'CLASS', 'REGION', 'UNIQUE_CARRIER_NAME', 'ORIGIN_CITY_NAME', 'ORIGIN',
        'DEST_CITY_NAME', 'DEST', 'PASSENGERS', 'FREIGHT', 'MAIL'
   ]
).rename(lambda x: x.lower(),
   axis=1).assign( region=lambda x:
   x.region.replace({
        'D': 'Domestic', 'I': 'International', 'A':'Atlantic',
       'L': 'Latin America', 'P': 'Pacific', 'S':'System'
   }),
   route=lambda x:
       np.where( x.origin <
       x.dest,
       x.origin + '-' + x.dest,
       x.dest + '-' + x.origin
)
```

## flight\_stats.head()

ro te	class	dest_city_name	dest	origin_ ity_name	origin	region	unique_carrier_name	mail	freight	passengers	
C B- IAH	G	Houston, TX	IAH	Dubai, United Arab Emirates	DXB	International	Emirates	0.0	53185.0	0.0	0
C B- JFK	G	New York, NY	JFK	Dubai, United Arab Emirates	DXB	International	Emirates	0.0	9002.0	0.0	1
D B-	G	Chicago, IL	ORD	Dubai, United Arab Emirates	DXB	International	Emirates	0.0	2220750.0	0.0	2
D <sub>B</sub> -		Dubai, United Arab									

```
cities = [
    'Atlanta, GA', 'Chicago, IL', 'New York, NY', 'Los Angeles, CA',
    'Dallas/Fort Worth, TX', 'Denver, CO', 'Houston, TX',
    'San Francisco, CA', 'Seattle, WA', 'Orlando, FL'
]

top_airlines = [
    'American Airlines Inc.', 'Delta Air Lines Inc.', 'JetBlue Airways',
    'Southwest Airlines Co.', 'United Air Lines Inc.'
]
```

total\_flight\_stats = flight\_stats.query(
 f`class` == "F" and origin\_city\_name != dest\_city\_name'

```
f' and origin_city_name.isin({cities}) and dest_city_name.isin({cities})'
).groupby([
    'origin', 'origin_city_name', 'dest', 'dest_city_name'
])[['passengers', 'freight', 'mail']].sum().reset_index().query('passengers > 0')
total_flight_stats.sample(10, random_state=1)
           origin
                    origin_city_name
                                               dest_city_name
                                                                            freight
                                                                                          mail
                                       dest
                                                               passengers
      78
             I GA
                                       DFN
                                                                           506023.0
                                                                                      293108.0
                         New York, NY
                                                   Denver, CO
                                                                 589190.0
             ORD
                           Chicago, IL
      117
                                       SFA
                                                   Seattle, WA
                                                                 810594.0
                                                                           1063463 0 2627325 0
      31
             DFW
                   Dallas/Fort Worth, TX MCO
                                                   Orlando, FL
                                                                 683700.0
                                                                            187672.0
                                                                                       95570.0
       5
             ATL
                           Atlanta, GA
                                       LAX
                                               Los Angeles, CA
                                                                1121378.0
                                                                          8707125.0
                                                                                     3267077.0
      126
             SEA
                           Seattle, WA
                                       LGA
                                                 New York, NY
                                                                     24.0
                                                                                 0.0
                                                                                            0.0
      45
              IAH
                          Houston, TX
                                        ATL
                                                   Atlanta, GA
                                                                 566369.0
                                                                            367543.0
                                                                                      726670.0
                                       HOU
                                                                 305193.0
                                                                            363119.0
      14
             DEN
                           Denver, CO
                                                  Houston, TX
                                                                                            0.0
      44
             HOU
                          Houston, TX
                                       SFO
                                             San Francisco, CA
                                                                   1843.0
                                                                              5523.0
                                                                                            0.0
      73
             LAX
                       Los Angeles, CA MDW
                                                   Chicago, IL
                                                                 277226.0 2022416.0
                                                                                            0.0
      89
             MCO
                           Orlando, FL
                                       DEN
                                                   Denver, CO
                                                                 594878.0
                                                                           368516.0
                                                                                      138811.0
import holoviews as hv
chord = hv.Chord(
    total_flight_stats,
    kdims=['origin', 'dest'],
    vdims=['passengers', 'origin_city_name', 'dest_city_name', 'mail', 'freight']
)
from bokeh.models import HoverTool
tooltips = {
    'Source': '@origin_city_name (@origin)',
    'Target': '@dest_city_name (@dest)',
    'Passengers': '@passengers{0,.},',
    'Mail': '@mail{0,.} lbs.',
    'Freight': '@freight{0,.} lbs.',
hover = HoverTool(tooltips=tooltips)
Sankey plot
top_cities = cities[:5]
domestic_passenger_travel = flight_stats.query(
    'region == "Domestic" and `class` == "F" and origin_city_name != dest_city_name '
    f'and origin_city_name.isin({top_cities})) and dest_city_name.isin({top_cities})'
).groupby([
    'region', 'unique_carrier_name', 'route',
    'origin_city_name', 'dest_city_name'
]).passengers.sum().reset_index()
domestic_passenger_travel.head()
```

	region	unique_carrier_name	route or	rigin_city_name des	t_city_name pass	engers
0	Domestic	Air Wisconsin Airlines Corp	ATL- ORD	Atlanta, GA	Chicago, IL	915.0
1	Domestic	Air Wisconsin Airlines Corp	ATL- ORD	Chicago, IL	Atlanta, GA	556.0
2	Domestic	Alaska Airlines Inc.	JFK- LAX	Los Angeles, CA	New York, NY	265307.0
			JFK-			

domestic\_passenger\_travel.unique\_carrier\_name.replace( '^(?
!' + '|'.join(top\_airlines) + ').\*\$',

```
'Other Airlines',
    regex=True, inplace=True
)
domestic_passenger_travel.groupby('unique_carrier_name').passengers.sum().div( domestic_p
   assenger_travel.passengers.sum()
)
     unique_carrier_name
     American Airlines Inc.
                                0.337186
     Delta Air Lines Inc.
                                0.312187
     JetBlue Airways
                                0.049500
     Other Airlines
                                0.120544
     Southwest Airlines Co.
                                0.079074
     United Air Lines Inc.
                                0.101509
     Name: passengers, dtype: float64
def get edges(data, *, source col, target col):
    aggregated = data.groupby([source_col, target_col]).passengers.sum()
    return aggregated.reset_index().rename(
        columns={source_col: 'source', target_col: 'target'}
   ).query('passengers > 0')
carrier_edges = get_edges(
    domestic_passenger_travel,
    source_col='region',
    target_col='unique_carrier_name'
).replace('Domestic', 'Top Routes')
carrier_edges
            source
                                 target
                                          passengers
       0 Top Routes
                     American Airlines Inc.
                                           9426060.0
       1 Top Routes
                        Delta Air Lines Inc.
                                           8727210.0
       2 Top Routes
                          JetBlue Airways
                                           1383776.0
       3 Top Routes
                                           3369815.0
                            Other Airlines
       4 Top Routes Southwest Airlines Co.
                                           2210533.0
       5 Top Routes
                       United Air Lines Inc.
                                           2837682.0
carrier_to_route_edges =
    get_edges( domestic_passenger_travel,
    source_col='unique_carrier_name',
    target_col='route'
)
carrier_to_route_edges.sample(10, random_state=1)
                       source
                                  target passengers
      39
                 Other Airlines
                               DFW-LGA
                                             157366.0
      41
                 Other Airlines
                                 JFK-LAX
                                             523222.0
      2
          American Airlines Inc.
                                 ATL-LAX
                                             294304.0
```

```
48
   Southwest Airlines Co. ATL-MDW
                                       498481.0
50
    Southwest Airlines Co.
                         LAX-MDW
                                       558574.0
44
           Other Airlines
                         LAX-ORD
                                       378552.0
33
           Other Airlines
                           ATL-LAX
                                       146882.0
           Other Airlines ATL-MDW
35
                                         1201.0
40
           Other Airlines DFW-ORD
                                       241147.0
27
          JetBlue Airways
                          DFW-JFK
                                           140.0
```

```
import holoviews as hv
hv.extension('bokeh')

sankey =
    hv.Sankey( all_edge
    s,
    kdims=['source', 'target'],
    vdims=hv.Dimension('passengers', unit='M')
).opts(
    labels='index', label_position='right', cmap='Set1',
    edge_color='lightgray',
    width=750, height=600,
    title='Travel Between the Top 5 Cities in 2019'
)

***Exercises**

Exercises**
```

1. For the 10 carriers that transported the most freight, create a bar plot showing total freight transported per carrier.

```
import matplotlib.pyplot as plt
# Group by carrier and sum the freight
top_10_carriers = flight_stats.groupby('unique_carrier_name')['freight'].sum().nlargest(10)
# Create a bar plot
plt.bar(top_10_carriers.index, top_10_carriers.values)
# Set labels and title
plt.xlabel('Carrier')
plt.ylabel('Total Freight (in millions)')
plt.title('Total Freight Transported by Top 10 Carriers')
# Rotate x-axis labels diagonally for better visibility
plt.xticks(rotation=45, ha='right')
# Show the plot
plt.show()
import pandas as pd
import geopandas as gpd
from bokeh.plotting import figure, show
from bokeh.models import HoverTool
# Load earthquake data into a geopandas DataFrame
earthquakes = gpd.read_file('/kaggle/input/shu/earthquakes.geojson')
# Convert the 'time' column to a datetime object
earthquakes['time'] = pd.to_datetime(earthquakes['time'], unit='ms')
# Group by day and count the number of earthquakes
earthquakes_per_day = earthquakes.groupby(pd.Grouper(key='time', freq='D')).size()
# Create a line plot with tooltips
p = figure(
   title='Total Earthquakes per Day',
    x_axis_label='Date',
   y_axis_label='Total Earthquakes',
   x_axis_type='datetime',
   tools='hover',
   tooltips=[('Date', '@time{%F}'), ('Total Earthquakes', '@value')],
    sizing_mode="stretch_both"
)
p.line(
   x=earthquakes_per_day.index,
   y=earthquakes_per_day.values,
```

```
line_width=2,
```

```
# Format the hover tooltip to show the date in YYYY-MM-DD format p.hover.formatters = {'@time': 'datetime'}

# Show the plot show(p)
```

3. Make histograms of earthquake magnitude (mag) for each magnitude type (magType) with a dropdown to select the magnitude type.

```
import pandas as pd
import geopandas as gpd
from bokeh.layouts import column
from bokeh.plotting import figure, show
from bokeh.models import ColumnDataSource, Select
# Load earthquake data into a geopandas DataFrame
earthquakes = gpd.read_file('/kaggle/input/shu/earthquakes.geojson')
# Filter out NaN values from the magType column
earthquakes = earthquakes.dropna(subset=['magType'])
# Create a ColumnDataSource for the histograms
source = ColumnDataSource(data=dict(mag=[], magType=[]))
# Create a histogram plot
p = figure(
    title='Histogram of Earthquake Magnitude',
    x_axis_label='Magnitude',
    y axis label='Count',
    tools='box_select, lasso_select',
    sizing_mode="stretch_both"
hist, edges = np.histogram([], bins=50)
p.quad(
    top=hist,
    bottom=0,
    left=edges[:-1],
    right=edges[1:],
    fill_color='navy',
    line_color='white',
    alpha=0.5
)
# Create a dropdown menu to select the magnitude type (sorted by magnitude type)
magTypes = sorted(earthquakes.magType.unique().tolist())
magType_menu = Select(
    title='Magnitude Type',
    options=magTypes,
    value=magTypes[0]
)
# Define a callback function to update the histogram when the magnitude type is changed
def update_hist(attr, old, new):
    selected_data = earthquakes[earthquakes.magType == magType_menu.value]
    hist, edges = np.histogram(selected_data.mag, bins=50)
    source.data = dict(mag=selected_data.mag, magType=[magType_menu.value] * len(selected_data))
    p.title.text = f'Histogram of Earthquake Magnitude ({magType_menu.value})'
    p.quad(
        top=hist,
        bottom=0,
        left=edges[:-1],
        right=edges[1:],
        fill color='navy',
        line_color='white',
        alpha=0.5
    )
# Attach the update_hist callback function to the magnitude type dropdown
magType_menu.on_change('value', update_hist)
# Create a layout with the histogram plot and the magnitude type dropdown
layout = column(magType_menu, p)
```

# Show the plot update\_hist(None, None, magType\_menu.value) show(layout)

0 giây hoàn thành lúc

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