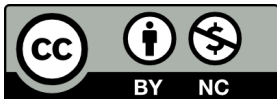


# Business 4720 - Class 8

## Data Visualization with Python

Joerg Evermann

Faculty of Business Administration  
Memorial University of Newfoundland  
[jevermann@mun.ca](mailto:jevermann@mun.ca)



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## What You Will Learn:

- ▶ Visualizing data with Python using the Plotly Express library
- ▶ Interactive data dashboards with Plotly Dash

# Example Dataset

- ▶ Government of Canada, Open Government Portal
- ▶ Fuel Consumption Ratings – Battery-electric vehicles – 2012–2023; last updated Oct 10, 2023
- ▶ <https://open.canada.ca/data/en/dataset/98f1a129-f628-4ce4-b24d-6f16bf24dd64>

Column	Data Type
Make	Discrete
Model	Discrete
Year	Numeric
Category	Discrete
City	Numeric <sup>1</sup>
Hwy	Numeric
Comb	Numeric
Range	Numeric <sup>2</sup>

---

<sup>1</sup>Fuel consumption in l/100km equivalent

<sup>2</sup>Range in km

# Data Preparation

Import required packages:

```
import pandas as pd
import plotly.express as px
import plotly.io as pio
pio.kaleido.scope.mathjax = None
```

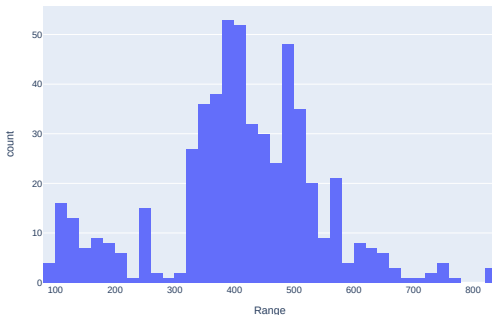
```
# Read data
data = pd.read_csv('https://evermann.ca/busi4720/fuel.csv')
```

# Histogram

```
# Create histogram
fig = px.histogram(data, x='Range', nbins=50)

# Show histogram, interactive in browser
fig.show()

# Save figure to image
fig.write_image("px.histogram.pdf", height=500, width=750)
```



# Column Chart

Prepare data in "long" format using `pd.melt()`:

```
data_grouped = \
data.groupby('Year') \
    .agg(
        meanCity=('City', 'mean'),
        meanHwy=('Hwy', 'mean')) \
    .reset_index()
```

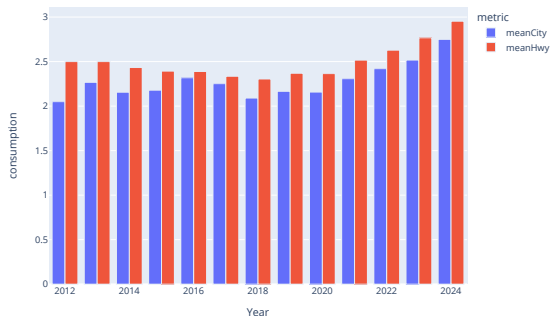
```
data_long = \
    pd.melt(data_grouped,
            id_vars=['Year'],
            value_vars=['meanCity',
                        'meanHwy'],
            var_name='metric',
            value_name='consumption')
```

```
>>> data_grouped
   Year  meanCity  meanHwy
0  2012   2.050000   2.500000
1  2013   2.266667   2.500000
2  2014   2.155556   2.433333
3  2015   2.178571   2.392857
4  2016   2.318519   2.388889
```

```
>>> data_long
   Year  metric  consumption
0  2012  meanCity      2.050000
1  2013  meanCity      2.266667
2  2014  meanCity      2.155556
3  2015  meanCity      2.178571
4  2016  meanCity      2.318519
```

# Column Chart [cont'd]

```
fig = px.bar(data_long,  
             x='Year', y='consumption', color='metric',  
             barmode='group')
```

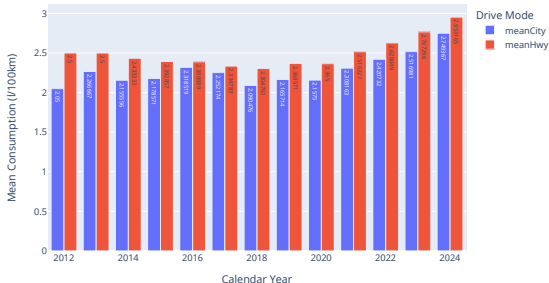


- The `barmode='group'` places bars of different categories next to each other (instead of stacking them)

# Labelling

```
fig = px.bar(data_long,  
             x='Year', y='consumption', color='metric', barmode='group',  
             text_auto=True,  
             title = 'Electric Vehicle Range',  
             labels={'consumption': 'Mean Consumption (l/100km)',  
                    'Year': 'Calendar Year',  
                    'metric': 'Drive Mode'})
```

Electric Vehicle Range

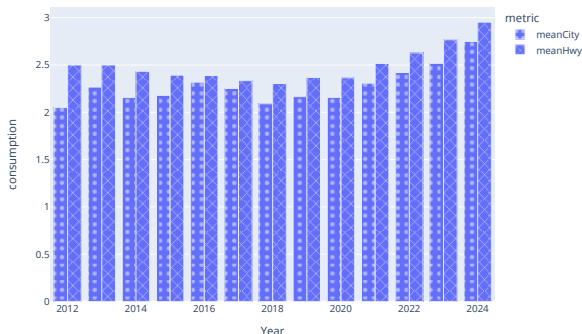


- ▶ No subtitles or captions
- ▶ Labels for each plot element
- ▶ Values with `text_auto=True` option



# Column Chart (with Patterns)

```
fig = px.bar(data_long,  
             x='Year', y='consumption',  
             pattern_shape = 'metric',  
             pattern_shape_sequence = ['.', 'x', '+', '|', '-', '/'],  
             barmode='group')
```



► Note the different pattern shapes

# Hands-On Exercises

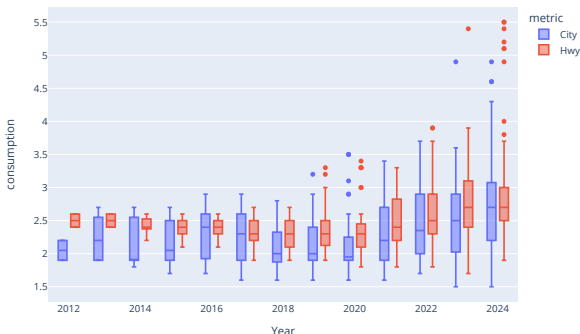
- 1 Read the EV fuel efficiency data set into a Pandas data frame.
- 2 Create a histogram of highway fuel efficiency with 25 bins.
- 3 Add labels for the axes, and add a title.

## Tips

- ▶ Use the `pd.read_csv()` function from Pandas
- ▶ The column name is `Hwy`
- ▶ Use the `px.histogram()` function from Plotly Express
- ▶ Use the `title=...` option
- ▶ Use the `labels='...'` option

# Box Plot

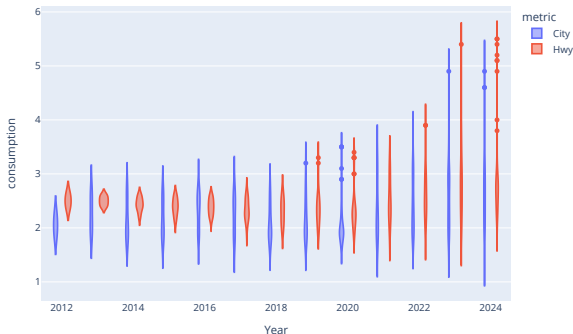
```
data_long = pd.melt(data,  
    id_vars=['Year'], value_vars=['City', 'Hwy'],  
    var_name='metric', value_name='consumption')  
fig = px.box(data_long,  
    x='Year', y='consumption', color='metric')
```



- Shows distribution
- Median
- 1st quartile  $Q_1$
- 3rd quartile  $Q_3$
- "Inter-quartile range"
- $IQR = Q_3 - Q_1$
- "Whiskers"
- $Q_3 + 1.5 \times IQR$
- $Q_1 - 1.5 \times IQR$
- "Outliers"

# Violin Plot

```
data_long = pd.melt(data,  
    id_vars=['Year'], value_vars=['City', 'Hwy'],  
    var_name='metric', value_name='consumption')  
fig = px.violin(data_long,  
    x='Year', y='consumption', color='metric')
```

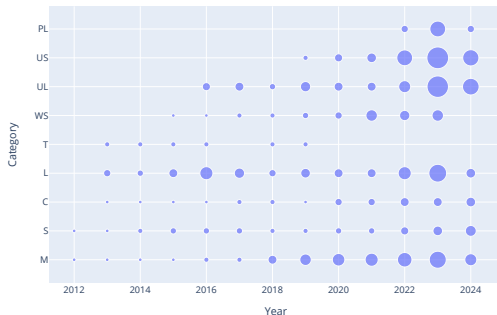


- Shows detailed density
- But no summary statistics

# Count Plot

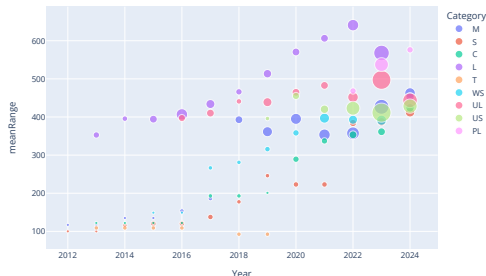
```
count_data = data \
    .groupby(['Year', 'Category']) \
    .agg(counts = ('Range', 'size')) \
    .reset_index()

fig = px.scatter(count_data, x='Year', y='Category', size='counts')
```



# Points Plot

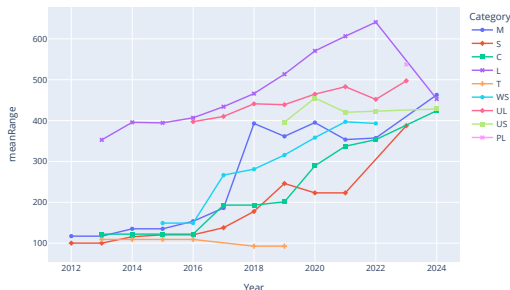
```
grouped_data = data \
    .groupby(['Year', 'Category']) \
    .agg(totalcount=('Range', 'size'),
         meanRange=('Range', 'mean')) \
    .reset_index()
fig = px.scatter(grouped_data,
                 x='Year', y='meanRange',
                 size='totalcount', color='Category')
```



► Shows 4 variables

# Lines and Points Plot

```
filtered_data =  
    data.groupby(['Year', 'Category']) \  
        .agg(meanRange=('Range', 'mean')) \  
        .reset_index() \  
    [data['Category'].isin(['C', 'L', 'M', 'S', 'US', 'UL'])]  
fig = px.line(filtered_data,  
              x='Year', y='meanRange', color='Category',  
              symbol='Category', markers=True)
```

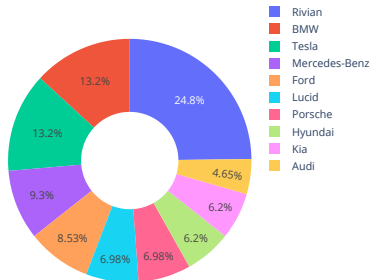


► Category mapped to two plot elements

# Pie/Donut Chart

```
data_pie = \
    data[data['Year'] == 2023] \
        .groupby('Make') \
        .agg(totalcount=('Model', 'size')) \
        .reset_index() \
        .query('totalcount >= 5')

fig = px.pie(data_pie,
             names='Make',
             values='totalcount',
             hole=0.4)
```





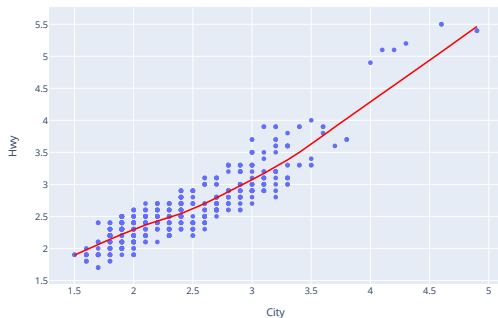
# Radar Plot

```
from sklearn.preprocessing \
    import MinMaxScaler
grouped = data \
    .query('Year == 2023') \
    .groupby('Make') \
    .agg(
        Cty=('City', lambda x: 1/x.mean()),
        Hwy=('Hwy', lambda x: 1/x.mean()),
        Rng=('Range', lambda x: x.mean()/100),
        nModels=('Make', 'size')) \
    .query('nModels >= 5')
grouped[['Cty', 'Hwy', 'Rng']] = \
    MinMaxScaler().fit_transform(
        grouped[['Cty', 'Hwy', 'Rng']])
melted = grouped \
    .reset_index() \
    .melt(id_vars='Make',
          value_vars=['Cty', 'Hwy', 'Rng'])
fig = px.line_polar(melted,
                    r='value', theta='variable',
                    color='Make',
                    line_close=True)
```



# Local Regression Smoothing Plot

```
fig = px.scatter(data,  
    x='City', y='Hwy',  
    trendline='lowess',  
    trendline_color_override='red')
```



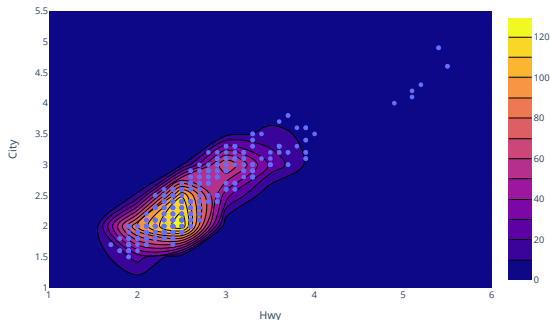
► Local regression line

# 2D Histogram Plot

```
import plotly.graph_objects as go

fig = px.scatter(data, x='Hwy', y='City')

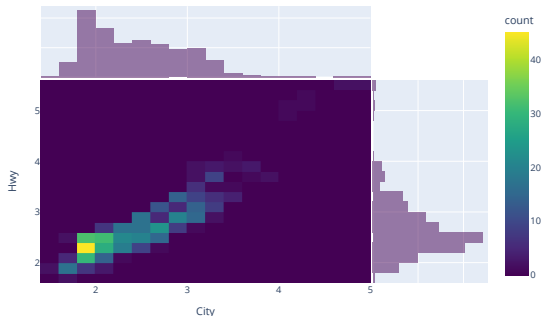
fig.add_trace(
    go.Histogram2dContour(x=data['Hwy'], y=data['City']))
```



► Counts of observations/data points

# Density Heatmap with Marginals

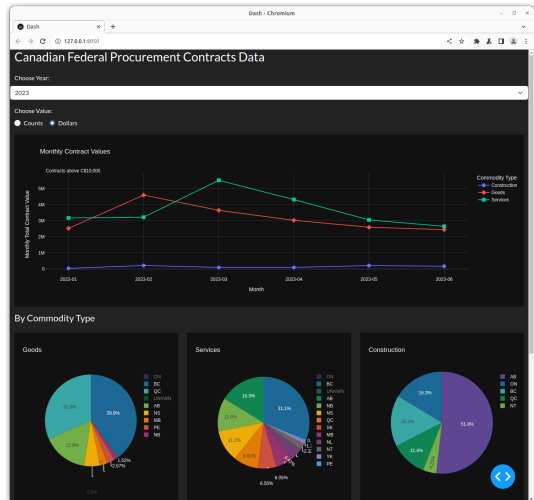
```
fig = px.density_heatmap(data,  
    x = 'City', y = 'Hwy',  
    nbinsx=20, nbinsy=20,  
    marginal_x='histogram', marginal_y='histogram',  
    color_continuous_scale = px.colors.sequential.Viridis)
```



- ▶ Marginals can be applied to other plot types
- ▶ Other options for the marginal plots are rug, box, and violin.
- ▶ Custom colour palette/scale

# Dashboards

- ▶ Set of related plots
- ▶ Interactive and customizable
- ▶ Automatically updated
- ▶ Simple and easy to understand
- ▶ Provides quick overview of key metrics



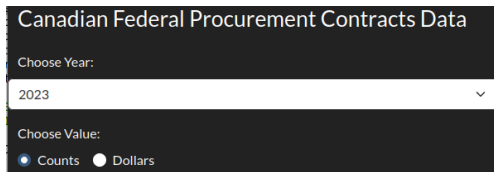
# Step 1 – Import Packages

```
from dash import Dash, html, dcc, callback, Output, Input
import dash_bootstrap_components as dbc

import plotly.express as px
```

## Step 2 – Visual Layout

- ▶ Container element with multiple Row elements
- ▶ Row elements contain Column elements, HTML elements, Label elements, Selection elements, Radio button elements, etc.



Canadian Federal Procurement Contracts Data

Choose Year:

2023

Choose Value:

☒ Counts ☐ Dollars

- ▶ First row is HTML heading
- ▶ Second row is Label element and Selection element, followed by another Label element and Radio button element

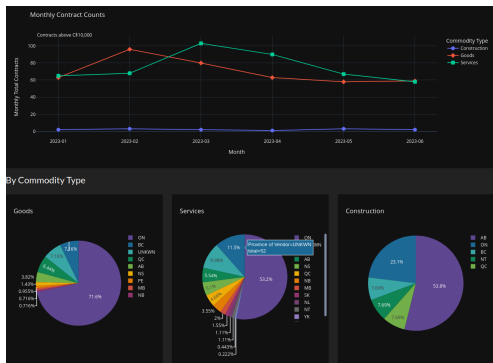
## Step 2 – Visual Layout [cont'd]

```
app.layout = dbc.Container([
    dbc.Row([
        html.H3('Canadian Federal Procurement Contracts Data'))],
    dbc.Row([
        html.P(),
        dbc.Label('Choose Year: '),
        dbc.Select(
            id='year-selection',
            options=[{"label": x, "value": x} for x in years],
            value=years[-1]),
        html.P(),
        dbc.Label('Choose Value: '),
        dbc.RadioItems(
            options=[{"label": "Counts", "value": 0},
                    {"label": "Dollars", "value": 1}],
            value=0, inline = True,
            id = 'value-selection'),
        html.P() ]),
```

- Every component has an ID!



## Step 2 – Visual Layout [cont'd]



- ▶ First row contains single column with a Graph element that contains a figure (line chart)
- ▶ Second row is an HTML heading
- ▶ Third row contains three columns, each with one graph element that contain a figure (pie chart)

## Step 2 – Visual Layout [cont'd]

```
# Continued from previous snippet
dbc.Row([
  dbc.Col([
    dcc.Graph(figure={}, id='line')
  ], width=12),
]),
dbc.Row([
  html.P(),
  html.H4('By Commodity Type'),
  html.P()
]),
dbc.Row([
  dbc.Col([
    dcc.Graph(figure={}, id='pie1')
  ], width=4),
  dbc.Col([
    dcc.Graph(figure={}, id='pie2')
  ], width=4),
  dbc.Col([
    dcc.Graph(figure={}, id='pie3')
  ], width=4),
]),
```

## Step 3 – Create the Figures

```
@callback(  
    Output(component_id='line', component_property='figure'),  
    [  
        Input(component_id='year-selection',  
              component_property='value'),  
        Input(component_id='value-selection',  
              component_property='value')  
    ]  
)  
def update_line(year_chosen, value_chosen):  
    # Figure creation code, for example  
    # fig = px.histogram(...)  
    return fig
```

- ▶ Each figure has an update function that creates and returns the figure
- ▶ Each update function has a callback specification
  - ▶ The `Output` specifies the Graph element for this figure
  - ▶ The `Input` elements specify which interactive elements provide values for creating or updating the figure

## Step 4 – Run the App

```
app.run()
```

# Hands-On Exercises

Use the Pagila film rentals data from

<https://evermann.ca/busi4720/rentals.csv>

- 1 Read the data into a Pandas data frame using `read_csv()`
  - 2 Create a box plot of the rental payment amounts for films by rating
  - 3 Create a violin plot of the rental payment amounts for films by rating
- Compare the information conveyed by a box plot and a violin plot. What are the commonalities and what are the differences?

# Hands-On Exercises

Use the Pagila film rentals data from

<https://evermann.ca/busi4720/rentals.csv>

- 1 Read the data into a Pandas data frame using `read_csv()`
- 2 Use `drop_duplicates()` to drop duplicates of the film titles
  - ▶ Because each film may have been rented multiple times.
- 3 Produce a *histogram* of counts of films by rating

## Tip:

- ▶ Use the `columns` and `shape` properties or the `describe()` method of a data frame to examine it.

# Hands-On Exercises

Use the Pagila film rentals data from

<https://evermann.ca/busi4720/rentals.csv>

- 1 Read the data into a Pandas data frame using `read_csv()`
- 2 Create a data frame with the mean rental payments for films by rating
- 3 Generate a bar chart of the mean rental payments for films by rating
- 4 Generate a pie or donut chart of rental counts by film rating

## Tips:

- ▶ Use the `groupby()` function to group the data
- ▶ Use the `mean()` and `std()` to find the mean and standard deviation of for a data frame column