Pandas Built-in Data Visualization In this lecture we will learn about pandas built-in capabilities for data visualization! It's built-off of matplotlib, but it baked into pandas for easier usage! Let's take a look! **Imports** import numpy as np In [18]: import pandas as pd import matplotlib.pyplot as plt %matplotlib inline The Data There are some fake data csv files you can read in as dataframes: In [19]: df1 = pd.read\_csv('df1',index\_col=0) df2 = pd.read\_csv('df2') In [20]: Out[20]: **0** 0.039762 0.218517 0.103423 0.957904 **1** 0.937288 0.041567 0.899125 0.977680 **2** 0.780504 0.008948 0.557808 0.797510 **3** 0.672717 0.247870 0.264071 0.444358 **4** 0.053829 0.520124 0.552264 0.190008 **5** 0.286043 0.593465 0.907307 0.637898 **6** 0.430436 0.166230 0.469383 0.497701 **7** 0.312296 0.502823 0.806609 0.850519 **8** 0.187765 0.997075 0.895955 0.530390 **9** 0.908162 0.232726 0.414138 0.432007 **Style Sheets** Matplotlib has style sheets you can use to make your plots look a little nicer. These style sheets include plot\_bmh,plot\_fivethirtyeight,plot\_ggplot and more. They basically create a set of style rules that your plots follow. I recommend using them, they make all your plots have the same look and feel more professional. You can even create your own if you want your company's plots to all have the same look (it is a bit tedious to create on though). Here is how to use them. Before plt.style.use() your plots look like this: df1['A'].hist() In [21]: <AxesSubplot:> Out[21]: 250 200 150 100 50 -1Call the style: import matplotlib.pyplot as plt In [22]: plt.style.use('ggplot') Now your plots look like this: df1['A'].hist() In [23]: <AxesSubplot:> Out[23]: 250 200 150 100 50 0\_4 -1In [24]: plt.style.use('bmh') df1['A'].hist() <AxesSubplot:> Out[24]: 200 150 100 50 0\_4 -10 plt.style.use('dark\_background') In [25]: df1['A'].hist() <AxesSubplot:> Out[25]: 250 150 100 50 plt.style.use('fivethirtyeight') In [26]: df1['A'].hist() <AxesSubplot:> Out[26]: plt.style.use('ggplot') In [27]: Let's stick with the ggplot style and actually show you how to utilize pandas built-in plotting capabilities! **Plot Types** There are several plot types built-in to pandas, most of them statistical plots by nature: • df.plot.area df.plot.barh df.plot.density df.plot.hist • df.plot.line • df.plot.scatter • df.plot.bar • df.plot.box • df.plot.hexbin • df.plot.kde • df.plot.pie You can also just call df.plot(kind='hist') or replace that kind argument with any of the key terms shown in the list above (e.g. 'box', 'barh', etc..) Let's start going through them! Area df2.plot.area(alpha=1) <AxesSubplot:> Out[28]: 2.5 2.0 1.5 1.0 0.5 0.0 **Barplots** In [29]: df2.head() Out[29]: b **0** 0.039762 0.218517 0.103423 0.957904 **1** 0.937288 0.041567 0.899125 0.977680 **2** 0.780504 0.008948 0.557808 0.797510 **3** 0.672717 0.247870 0.264071 0.444358 **4** 0.053829 0.520124 0.552264 0.190008 In [30]: df2.plot.bar() <AxesSubplot:> Out[30]: 1.0 0.8 0.6 0.4 0.2 df2.plot.bar(stacked=True) In [31]: <AxesSubplot:> Out[31]: 2.5 2.0 1.5 1.0 0.5 0.0 Histograms In [32]: df1['A'].plot.hist(bins=50) <AxesSubplot:ylabel='Frequency'> Out[32]: 60 50 Frequency 40 30 20 10 -11 **Line Plots** In [33]: df2.plot.line(x=df2.index, y='B', figsize=(12,3), lw=1)KeyError Traceback (most recent call last) Input In [33], in <cell line: 1>() ----> 1 df2.plot.line(x=df2.index,y='B',figsize=(12,3),lw=1) File ~\anaconda3\lib\site-packages\pandas\plotting\ core.py:1041, in PlotAccessor.line(self, x, y, \*\*kwargs) 976 @Appender( 0.00 977 978 See Also (...) 1033 @Appender(\_bar\_or\_line\_doc) 1034 **def** line(self, x=None, y=None, \*\*kwargs): 1035 1036 Plot Series or DataFrame as lines. 1038 This function is useful to plot lines using DataFrame's values 1039 as coordinates. -> 1041 return self(kind="line", x=x, y=y, \*\*kwargs) File ~\anaconda3\lib\site-packages\pandas\plotting\\_core.py:937, in PlotAccessor. call (self, \*args, \*\*kwarg 935 **if** is integer(x) **and not** data.columns.holds integer():  $x = data_cols[x]$ --> 937 elif not isinstance(data[x], ABCSeries): 938 raise ValueError("x must be a label or position") 939 data = data.set index(x) File ~\anaconda3\lib\site-packages\pandas\core\frame.py:3511, in DataFrame.\_\_getitem\_\_(self, key) if is iterator(key): 3509 3510 key = list(key)indexer = self.columns.\_get\_indexer\_strict(key, "columns")
[1] -> 3511 3513 # take() does not accept boolean indexers 3514 if getattr(indexer, "dtype", None) == bool: File ~\anaconda3\lib\site-packages\pandas\core\indexes\base.py:5782, in Index.\_get\_indexer\_strict(self, key, ax is\_name) 5779 **else:** keyarr, indexer, new indexer = self. reindex non\_unique(keyarr) -> 5782 self.\_raise\_if\_missing(keyarr, indexer, axis name) 5784 keyarr = self.take(indexer) 5785 **if** isinstance(key, Index): # GH 42790 - Preserve name from an Index File ~\anaconda3\lib\site-packages\pandas\core\indexes\base.py:5842, in Index.\_raise\_if\_missing(self, key, inde xer, axis\_name) 5840 if use interval msg: key = list(key)-> 5842 raise KeyError(f"None of [{key}] are in the [{axis\_name}]") 5844 not\_found = list(ensure\_index(key)[missing\_mask.nonzero()[0]].unique()) 5845 raise KeyError(f"{not found} not in index") KeyError: 'None of [RangeIndex(start=0, stop=10, step=1)] are in the [columns]' df1.plot.line() In [ ]: **Scatter Plots** df1.plot.scatter(x='A',y='B') In [34]: <AxesSubplot:xlabel='A', ylabel='B'> Out[34]: 2  $\alpha$ 0 You can use c to color based off another column value Use cmap to indicate colormap to use. For all the colormaps, check out: http://matplotlib.org/users/colormaps.html df1.plot.scatter(x='A',y='B',c='C',cmap='coolwarm') In [35]: <AxesSubplot:xlabel='A', ylabel='B'> Out[35]:  $\alpha$ 0 -1Or use s to indicate size based off another column. s parameter needs to be an array, not just the name of a column: df1.plot.scatter(x='A',y='B',s=df1['C']\*200) In [38]: C:\Users\Shubham kumawat\anaconda3\lib\site-packages\matplotlib\collections.py:982: RuntimeWarning: invalid val ue encountered in sqrt scale = np.sqrt(self. sizes) \* dpi / 72.0 \* self. factor <AxesSubplot:xlabel='A', ylabel='B'> Out[38]: 3 2 -1-3 -4 **BoxPlots** df2.plot.box() # Can also pass a by= argument for groupby **Hexagonal Bin Plot** Useful for Bivariate Data, alternative to scatterplot: df = pd.DataFrame(np.random.randn(1000, 2), columns=['a', 'b']) In [39]: df.plot.hexbin(x='a',y='b',gridsize=25,cmap='Oranges') <AxesSubplot:xlabel='a', ylabel='b'> Out[39]: 16 3 14 12 10 -2 -3 0 -3-2**Kernel Density Estimation plot (KDE)** df2['a'].plot.kde() In [36]: <AxesSubplot:ylabel='Density'> Out[36]: 0.8 0.6 0.0 -0.250.00 0.25 0.50 0.75 1.00 1.25 df2.plot.density() In [37]: <AxesSubplot:ylabel='Density'> Out[37]: 1.2 1.0 0.8 Density 0.6 0.4 0.2 0.0 -0.50-0.250.00 0.25 0.50 0.75 1.00 1.25 1.50 That's it! Hopefully you can see why this method of plotting will be a lot easier to use than full-on matplotlib, it balances ease of use with control over the figure. A lot of the plot calls also accept additional arguments of their parent matplotlib plt. call. **Great Job!**