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## Cheatsheets / Exploratory Data Analysis in Python



# Summarizing a Single Feature

#### Pandas .describe() method

The pandas method, .describe() provides summary statistics for all features in a dataset. Setting include = 'all' includes summary statistics for both quantitative and categorical features.

df.describe(include = 'all')

#### **Central tendency statistics**

To summarize the central tendency, or typical value, of a quantitative variable, we can use statistics such as the mean, median, and mode. These can be calculated using the pandas methods .mean(), .median(), and .mode(), respectively.

#calculate mean of a column
df.column\_name.mean()

#calculate median of a column
df.column\_name.median()

#calculate mode of a column
df.column\_name.mode()

1 of 3 3/12/24, 06:36

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#### **Spread statistics**

To summarize the spread, or variation, of a quantitative variable, we can use statistics such as the range, interquartile range, variance, standard deviation, and mean absolute deviation. These can be calculated as shown.

```
#calculate range of a column
df.column_name.max() -
df.column_name.min()

#calculate IQR of a column
df.column_name.quantile(0.75) -
df.column_name.quantile(0.25)

#calculate variance of a column
df.column_name.var()

#calculate standard deviation of a column
df.column_name.std()

#calculate MAD of a column
df.column_name.mad()
```

#### Visualize the distribution of a quantitative/continuous feature

To inspect the distribution of a quantitative variable, we can use visualizations such as histograms and box plots. We can create these plots using the seaborn functions histplot() and boxplot(), respectively.

```
import matplotlib.pyplot as plt
import seaborn as sns

#create histogram
sns.histplot(x = 'column_name', data = data_name)
plt.show()

#create boxplot
sns.boxplot(x = 'column_name', data = data_name)
plt.show()
```

2 of 3 3/12/24, 06:36

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#### Summary statistics for categorical data

To summarize the distribution of a categorical/discrete feature, we can calculate the number or proportion of observations in each category using the pandas method .value\_counts .

```
#calculate the number in each category
df.column_name.value_counts()
```

```
#calculate the proportion in each
category
df.column_name.value_counts(normalize =
True)
```

### Visualizing categorical data

To inspect and explore categorical features, we can use visualizations such as bar charts or pie charts. The provided code demonstrates how to create these plots.

```
import matplotlib.pyplot as plt
import seaborn as sns

#create bar chart
sns.countplot(x = 'column_name', data = data_name)
plt.show()

#create pie chart
df.column_name.value_counts().plot.pie()
plt.show()
```



3 of 3