



ANALYZING POLICE ACTIVITY WITH PANDAS

**Do the genders commit
different violations?**

Kevin Markham

Founder, Data School



Counting unique values

- `value_counts()`: Counts the unique values in a Series
- Best suited for categorical data

```
ri.stop_outcome.value_counts()
```

```
Citation          77091
```

```
Warning           5136
```

```
Arrest Driver     2735
```

```
No Action         624
```

```
N/D               607
```

```
Arrest Passenger  343
```

```
Name: stop_outcome, dtype: int64
```

```
ri.stop_outcome.value_counts().sum()
```

```
86536
```

```
ri.shape
```

```
(86536, 13)
```



Expressing counts as proportions

```
ri.stop_outcome.value_counts()
```

```
Citation          77091
Warning           5136
Arrest Driver     2735
No Action         624
N/D              607
Arrest Passenger   343
Name: stop_outcome, dtype: int64
```

```
77091/86536
```

```
0.8908546731995932
```

```
ri.stop_outcome.value_counts(normalize=True)
```

```
Citation          0.890855
Warning           0.059351
Arrest Driver     0.031605
No Action         0.007211
N/D              0.007014
Arrest Passenger   0.003964
Name: stop_outcome, dtype: float64
```



Filtering DataFrame rows

```
ri.driver_race.value_counts()
```

```
White      61870  
Black      12285  
Hispanic   9727  
Asian      2389  
Other       265  
Name: driver_race, dtype: int64
```

```
white = ri[ri.driver_race == 'White']
```

```
white.shape  
(61870, 13)
```



Comparing stop outcomes for two groups

```
white.stop_outcome.value_counts(normalize=True)
```

```
Citation          0.902263
Warning           0.057508
Arrest Driver     0.024018
No Action         0.007031
N/D               0.006433
Arrest Passenger  0.002748
Name: stop_outcome, dtype: float64
```

```
asian = ri[ri.driver_race == 'Asian']
```

```
asian.stop_outcome.value_counts(normalize=True)
```

```
Citation          0.922980
Warning           0.045207
Arrest Driver     0.017581
No Action         0.008372
N/D               0.004186
Arrest Passenger  0.001674
Name: stop_outcome, dtype: float64
```



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Let's practice!



ANALYZING POLICE ACTIVITY WITH PANDAS

**Does gender affect
who gets a ticket for
speeding?**

Kevin Markham

Founder, Data School

Filtering by multiple conditions (1)

```
female = ri[ri.driver_gender == 'F']  
  
female.shape  
(23774, 13)  
  
female_and_arrested = ri[(ri.driver_gender == 'F') &  
                          (ri.is_arrested == True)]
```

- Each condition is surrounded by parentheses
- Ampersand (&) represents the and operator

```
female_and_arrested.shape  
(669, 13)
```

- Only includes female drivers who were arrested

Filtering by multiple conditions (2)

```
female_or_arrested = ri[(ri.driver_gender == 'F') |  
                        (ri.is_arrested == True)]
```

- Pipe (|) represents the or operator

```
female_or_arrested.shape  
(26183, 13)
```

- Includes all females
- Includes all drivers who were arrested



Rules for filtering by multiple conditions

- Ampersand (&): only include rows that satisfy both conditions
- Pipe (|): include rows that satisfy either condition
- Each condition must be surrounded by parentheses
- Conditions can check for equality (==), inequality (!=), etc.
- Can use more than two conditions



Correlation, not causation

- Analyze the relationship between gender and stop outcome
 - Assess whether there is a correlation
- Not going to draw any conclusions about causation
 - Would need additional data and expertise
 - Exploring relationships only



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ANALYZING POLICE ACTIVITY WITH PANDAS

**Does gender affect
whose vehicle is
searched?**

Kevin Markham

Founder, Data School



Math with Boolean values

```
ri.isnull().sum()

stop_date           0
stop_time           0
driver_gender       0
driver_race         0
violation_raw       0
violation           0
search_conducted    0
search_type        83229
...
```

- True = 1, False = 0

```
import numpy as np

np.mean([0, 1, 0, 0])
0.25

np.mean([False, True, False, False])
0.25
```

- Mean of Boolean Series represents percentage of True values



Taking the mean of a Boolean Series

```
ri.is_arrested.value_counts(normalize=True)
```

```
False    0.964431
```

```
True     0.035569
```

```
Name: is_arrested, dtype: float64
```

```
ri.is_arrested.mean()
```

```
0.0355690117407784
```

```
ri.is_arrested.dtype
```

```
dtype('bool')
```

Comparing groups using groupby

- Study the arrest rate by police district

```
ri.district.unique()

array(['Zone X4', 'Zone K3', 'Zone X1', 'Zone X3', 'Zone K1', 'Zone K2'],
      dtype=object)

ri[ri.district == 'Zone K1'].is_arrested.mean()
0.024349083895853423

ri[ri.district == 'Zone K2'].is_arrested.mean()
0.030800588834786546

ri.groupby('district').is_arrested.mean()

district
Zone K1    0.024349
Zone K2    0.030801
Zone K3    0.032311
Zone X1    0.023494
Zone X3    0.034871
Zone X4    0.048038
Name: is_arrested, dtype: float64
```




Grouping by multiple categories

```
ri.groupby(['district', 'driver_gender']).is_arrested.mean()
```

district	driver_gender	
Zone K1	F	0.019169
	M	0.026588
Zone K2	F	0.022196
	M	0.034285
Zone K3	F	0.025156
	M	0.034961
Zone X1	F	0.019646
	M	0.024563
Zone X3	F	0.027188
	M	0.038166
Zone X4	F	0.042149
	M	0.049956

```
ri.groupby(['driver_gender', 'district']).is_arrested.mean()
```

driver_gender	district	
F	Zone K1	0.019169
	Zone K2	0.022196
...		
M	Zone K1	0.026588
	Zone K2	0.034285
...		



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**Does gender affect
who is frisked during a
search?**

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Founder, Data School



Examining the search types (1)

```
ri.search_conducted.value_counts()
```

```
False    83229
```

```
True      3307
```

```
Name: search_conducted, dtype: int64
```

```
ri.search_type.value_counts(dropna=False)
```

```
NaN                                         83229
```

```
Incident to Arrest                        1290
```

```
Probable Cause                           924
```

```
Inventory                                219
```

```
Reasonable Suspicion                     214
```

```
Protective Frisk                         164
```

```
Incident to Arrest,Inventory             123
```

```
Incident to Arrest,Probable Cause        100
```

```
Probable Cause,Reasonable Suspicion       54
```

```
Probable Cause,Protective Frisk          35
```

```
...
```

- `value_counts()` excludes missing values by default
- `dropna=False` displays missing values

Examining the search types (2)

```
ri.search_type.value_counts()

Incident to Arrest          1290
Probable Cause              924
Inventory                   219
Reasonable Suspicion        214
Protective Frisk            164
Incident to Arrest,Inventory 123
Incident to Arrest,Probable Cause 100
Probable Cause,Reasonable Suspicion 54
Incident to Arrest,Inventory,Probable Cause 35
Probable Cause,Protective Frisk 35
Incident to Arrest,Protective Frisk 33
Inventory,Probable Cause     25
Protective Frisk,Reasonable Suspicion 19
Incident to Arrest,Inventory,Protective Frisk 18
...
```

- Multiple values are separated by commas
- 219 searches in which "Inventory" was the only search type
- Locate "Inventory" among multiple search types

Searching for a string

```
ri['inventory'] = ri.search_type.str.contains('Inventory', na=False)
```

- `str.contains()` returns True if string is found, False if not found
- `na=False` returns False when it finds a missing value

```
ri.inventory.dtype  
dtype('bool')
```

- True means an inventory was done, False means it was not

```
ri.inventory.sum()  
441
```

Calculating the inventory rate

```
ri.inventory.mean()  
0.0050961449570121106
```

- 0.5% of all traffic stops resulted in an inventory

```
searched = ri[ri.search_conducted == True]  
  
searched.inventory.mean()  
0.13335349259147264
```

- 13.3% of searches included an inventory



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