## Do the genders commit different violations?

**ANALYZING POLICE ACTIVITY WITH PANDAS** 



**Kevin Markham**Founder, Data School



### Counting unique values (1)

- .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
```

### Counting unique values (2)

ri.stop\_outcome.value\_counts().sum()

86536

ri.shape

(86536, 13)

### Expressing counts as proportions

ri.stop\_outcome.value\_counts()

77091/86536

0.8908546731995932

Citation	77091	
Warning	5136	
Arrest Driver	2735	
No Action	624	
N/D	607	
Arrest Passenger	343	

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

### Filtering DataFrame rows

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

```
White 61870
Black 12285
Hispanic 9727
Asian 2389
Other 265
```

```
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
```

Citation	0.922980
Warning	0.045207
Arrest Driver	0.017581
No Action	0.008372
N/D	0.004186
Arrest Passenger	0.001674

## Let's practice!

ANALYZING POLICE ACTIVITY WITH PANDAS



# Does gender affect who gets a ticket for speeding?

**ANALYZING POLICE ACTIVITY WITH PANDAS** 



**Kevin Markham**Founder, Data School



### Filtering by multiple conditions (1)

```
female = ri[ri.driver_gender == 'F']
female.shape
```

(23774, 13)

### Filtering by multiple conditions (2)

- 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 (3)

• 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 (1): 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

## Let's practice!

ANALYZING POLICE ACTIVITY WITH PANDAS



## Does gender affect whose vehicle is searched?

**ANALYZING POLICE ACTIVITY WITH PANDAS** 



**Kevin Markham**Founder, Data School



#### Math with Boolean values

```
• True = 1, False = 0
```

```
import numpy as np
np.mean([0, 1, 0, 0])

0.25

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

 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

ri.is\_arrested.mean()

0.0355690117407784

ri.is\_arrested.dtype

dtype('bool')



### Comparing groups using groupby (1)

• 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
```

### Comparing groups using groupby (2)

```
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
```



### 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
```

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



## Let's practice!

ANALYZING POLICE ACTIVITY WITH PANDAS



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

**ANALYZING POLICE ACTIVITY WITH PANDAS** 



**Kevin Markham**Founder, Data School



ri.search\_conducted.value\_counts()

False 83229

True 3307

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
• • •	

- .value\_counts()excludes missingvalues by default
- dropna=False
  displays missing
  values

### **Examining the search types**

```
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	
•••		

- 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 (1)

```
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

### Searching for a string (2)

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

## Let's practice!

ANALYZING POLICE ACTIVITY WITH PANDAS

