

# Drawing Conclusions

December 5, 2017

## 0.0.1 Calculating Errors

Here are two datasets that represent two of the examples you have seen in this lesson.

One dataset is based on the parachute example, and the second is based on the judicial example. Neither of these datasets are based on real people.

Use the questions below to assist in answering the quiz questions at the bottom of this page.

```
In [1]: import numpy as np
import pandas as pd

jud_data = pd.read_csv('judicial_dataset_predictions.csv')
par_data = pd.read_csv('parachute_dataset.csv')
```

```
In [2]: jud_data.head()
```

```
Out[2]:
```

	defendant_id	actual	predicted
0	22574	innocent	innocent
1	35637	innocent	innocent
2	39919	innocent	innocent
3	29610	guilty	guilty
4	38273	innocent	innocent

```
In [3]: par_data.head()
```

```
Out[3]:
```

	parachute_id	actual	predicted
0	3956	opens	opens
1	2147	opens	opens
2	2024	opens	opens
3	8325	opens	opens
4	6598	opens	opens

```
In [35]: par_data.actual.unique()
```

```
Out[35]: array(['opens', 'fails'], dtype=object)
```

```
In [36]: par_data.predicted.unique()
```

```
Out[36]: array(['opens', 'fails'], dtype=object)
```

1. Above, you can see the actual and predicted columns for each of the datasets. Using the **jud\_data**, find the proportion of errors for the dataset, and furthermore, the percentage of errors of each type. Use the results to answer the questions in quiz 1 below.

```
In [4]: jud_count = jud_data.shape[0]
        jud_count
```

```
Out[4]: 7283
```

```
In [5]: jud_error_count = jud_data[jud_data['actual'] != jud_data['predicted']].shape[0]
        jud_error_count
```

```
Out[5]: 307
```

```
In [6]: print('proportion of errors for the judidical dataset: {}'.format(jud_error_count/jud_count))
proportion of errors for the judidical dataset: 0.042152958945489497
```

```
In [7]: jud_type1_count = jud_data.query("actual == 'innocent' and predicted == 'guilty').count()
        jud_type1_count
```

```
Out[7]: 11
```

```
In [8]: jud_type2_count = jud_data.query("actual == 'guilty' and predicted == 'innocent').count()
        jud_type2_count
```

```
Out[8]: 296
```

```
In [10]: print('proportion of type 1 errors for the judidical dataset: {}'.format(jud_type1_count/jud_count))
         print('proportion of type 2 errors for the judidical dataset: {}'.format(jud_type2_count/jud_count))
proportion of type 1 errors for the judidical dataset: 0.001510366607167376
proportion of type 2 errors for the judidical dataset: 0.04064259233832212
```

2. Using the **par\_data**, find the proportion of errors for the dataset, and furthermore, the percentage of errors of each type. Use the results to answer the questions in quiz 2 below.

```
In [21]: par_count = par_data.shape[0]
        par_count
```

```
Out[21]: 5829
```

```
In [22]: par_error_count = par_data[par_data['actual'] != par_data['predicted']].shape[0]
        par_error_count
```

```
Out[22]: 233
```

```
In [23]: print('proportion of errors for the parachute dataset: {}'.format(par_error_count/par_count))
```

proportion of errors for the parachute dataset: 0.039972551037913875

```
In [27]: par_type1_count = par_data.query("actual == 'fails' and predicted == 'opens').count()[
        par_type1_count
```

Out[27]: 1

```
In [29]: par_type2_count = par_data.query("actual == 'opens' and predicted == 'fails').count()[
        par_type2_count
```

Out[29]: 232

```
In [37]: print('proportion of type 1 errors for the parachute dataset: {}'.format(par_type1_count/par_data.count()[
        print('proportion of type 2 errors for the parachute dataset: {}'.format(par_type2_count/par_data.count()[
```

proportion of type 1 errors for the parachute dataset: 0.00017155601303825698

proportion of type 2 errors for the parachute dataset: 0.03980099502487562