

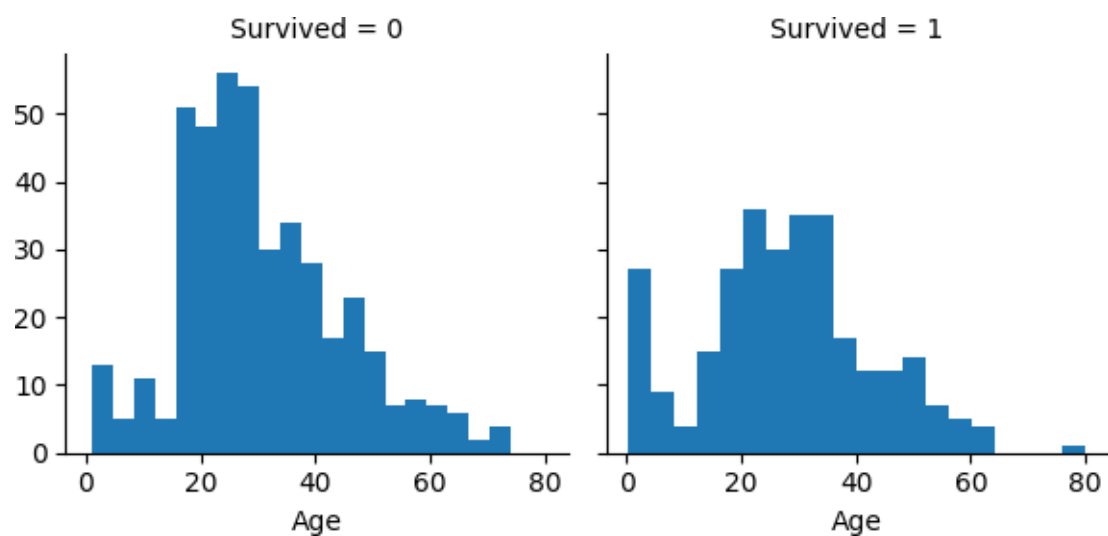
Assignment 3 Documentation

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1. Titanic dataset

- ✓ Read train and test dataset using pandas.
- ✓ Check head to see columns and type of data.
- ✓ Check for data imbalance using value_counts on y variable.
- ✓ Plot histogram for survived vs age.



- ✓ Drop unnecessary columns and separate y variable. Out: X_train, Y_train, X_test
- ✓ Check for null values and fill them with mean or median accordingly.
- ✓ Convert categorical columns to numerical with label encoding.
- ✓ Sample data after preprocessing is as below.

	Pclass	Sex	Age	Fare	Embarked
0	3	male	34.5	7.8292	Q
1	3	female	47.0	7.0000	S
2	2	male	62.0	9.6875	Q
3	3	male	27.0	8.6625	S
4	3	female	22.0	12.2875	S

- ✓ Fit 4 naïve bayes models(gaussian, multinomial, Bernoulli, complement) on X_train and Y_train. Use scikit-learn for the same.

- ✓ Predict on X_test.
- ✓ Since X_test original labels are not available we calculate accuracy on train data itself.
- ✓ Find classification_report, confusion_matrix, accuracy_score for each of the 4 models using scikit-learn
- ✓ Gaussian:

	precision	recall	f1-score	support	
	0	0.82	0.81	0.82	549
	1	0.70	0.72	0.71	342
accuracy				0.78	891
macro avg		0.76	0.77	0.77	891
weighted avg		0.78	0.78	0.78	891

```

[[445 104]
 [ 95 247]]
accuracy is 0.77665544332211

```

- ✓ Multinomial:

	precision	recall	f1-score	support	
	0	0.72	0.83	0.77	549
	1	0.64	0.48	0.55	342
accuracy				0.70	891
macro avg		0.68	0.65	0.66	891
weighted avg		0.69	0.70	0.68	891

```

[[457 92]
 [179 163]]
accuracy is 0.6958473625140292

```

- ✓ Bernoulli:

	precision	recall	f1-score	support	
	0	0.81	0.85	0.83	549
	1	0.74	0.68	0.71	342
accuracy				0.79	891
macro avg		0.78	0.77	0.77	891
weighted avg		0.78	0.79	0.78	891

```

[[468 81]
 [109 233]]
accuracy is 0.7867564534231201

```

- ✓ Complement:

	precision	recall	f1-score	support	
	0	0.72	0.83	0.77	549
	1	0.64	0.48	0.55	342
accuracy				0.70	891
macro avg		0.68	0.66	0.66	891
weighted avg		0.69	0.70	0.69	891

```
[[455  94]
 [177 165]]
accuracy is 0.6958473625140292
```

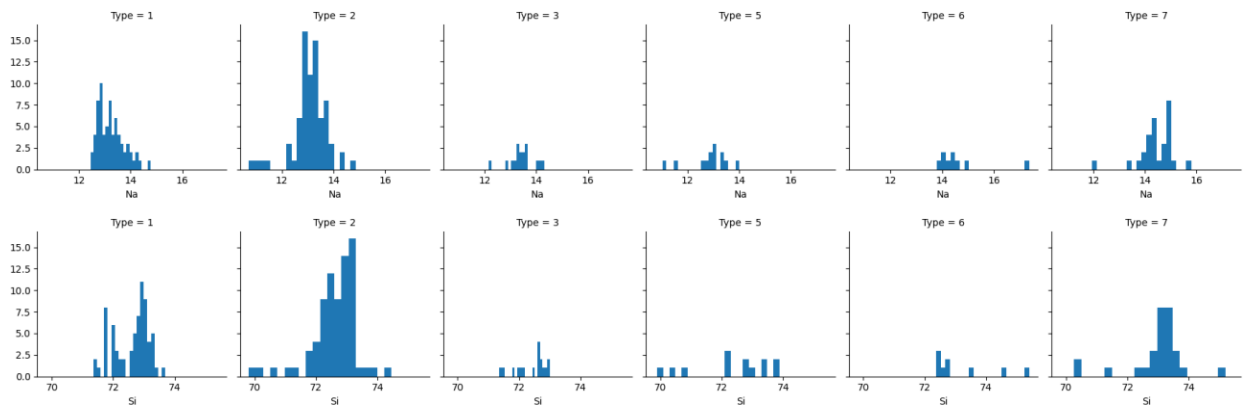
- ✓ Best fit model is Bernoulli naïve as it works well on binary dataset.

2. Glass dataset

- ✓ Read glass dataset using pandas.
- ✓ Check head to see columns and type of data.

RI	Na	Mg	Al	Si	K	Ca	Ba	Fe	Type
0	1.52101		13.64	4.49	1.10	71.78	0.06	8.75	0.0 0.0 1
1	1.51761		13.89	3.60	1.36	72.73	0.48	7.83	0.0 0.0 1
2	1.51618		13.53	3.55	1.54	72.99	0.39	7.78	0.0 0.0 1
3	1.51766		13.21	3.69	1.29	72.61	0.57	8.22	0.0 0.0 1
4	1.51742		13.27	3.62	1.24	73.08	0.55	8.07	0.0 0.0 1

- ✓ Check for data imbalance using value_counts on y variable. This dataset is highly **imbalanced**.
- ✓ Plot histogram for Type vs Na and Type vs Si to observe the relation of dependant variables with independent variable.



- ✓ Separate X and Y data.
- ✓ Split data for train and test with 20% test using train_test_split. Set random_set to get same split on repetition. Out: X_train, Y_train, X_test, Y_test.
- ✓ Fit 4 naïve bayes models(gaussian, multinomial, Bernoulli, complement) , SVC and Linear SVC on X_train and Y_train. Use scikit-learn for the same.
- ✓ Predict on X_test.
- ✓ Calculate accuracy on test data with Y_test and Y_pred_test.
- ✓ Find classification_report, confusion_matrix, accuracy_score for each of the 4 models using scikit-learn.
- ✓ Gaussian:

	precision	recall	f1-score	support	
1		0.19	0.44	0.27	9
2		0.33	0.16	0.21	19
3		0.33	0.20	0.25	5

5	0.00	0.00	0.00	2
6	0.67	1.00	0.80	2
7	1.00	1.00	1.00	6

accuracy			0.37	43
macro avg	0.42	0.47	0.42	43
weighted avg	0.40	0.37	0.36	43

```
[[ 4  3  1  0  1  0]
 [14  3  1  1  0  0]
 [ 3  1  1  0  0  0]
 [ 0  2  0  0  0  0]
 [ 0  0  0  0  2  0]
 [ 0  0  0  0  0  6]]
```

accuracy is 0.37209302325581395

✓ Multinomial:

precision	recall	f1-score	support	
1	0.28	0.89	0.42	9
2	0.40	0.11	0.17	19
3	0.00	0.00	0.00	5
5	0.00	0.00	0.00	2
6	0.00	0.00	0.00	2
7	0.67	1.00	0.80	6

accuracy			0.37	43
macro avg	0.22	0.33	0.23	43
weighted avg	0.33	0.37	0.27	43

```
[[ 8  1  0  0  0  0]
 [16  2  0  0  0  1]
 [ 5  0  0  0  0  0]
 [ 0  0  0  0  0  2]
 [ 0  2  0  0  0  0]
 [ 0  0  0  0  0  6]]
```

accuracy is 0.37209302325581395

✓ Bernoulli:

precision	recall	f1-score	support	
1	0.27	0.89	0.41	9
2	0.29	0.11	0.15	19
3	0.00	0.00	0.00	5
5	0.00	0.00	0.00	2
6	0.00	0.00	0.00	2
7	0.83	0.83	0.83	6

accuracy			0.35	43
macro avg	0.23	0.30	0.23	43
weighted avg	0.30	0.35	0.27	43

```
[[ 8  1  0  0  0  0]
```

```

[16  2  0  0  0  1]
[ 5  0  0  0  0  0]
[ 0  2  0  0  0  0]
[ 0  2  0  0  0  0]
[ 1  0  0  0  0  5]]
accuracy is 0.3488372093023256

```

✓ Complement:

precision	recall	f1-score	support		
	1	0.28	1.00	0.44	9
	2	0.00	0.00	0.00	19
	3	0.00	0.00	0.00	5
	5	1.00	0.50	0.67	2
	6	0.50	0.50	0.50	2
	7	0.75	1.00	0.86	6
accuracy				0.40	43
macro avg		0.42	0.50	0.41	43
weighted avg		0.23	0.40	0.27	43

```

[[ 9  0  0  0  0  0]
 [17  0  0  0  1  1]
 [ 5  0  0  0  0  0]
 [ 0  0  0  1  0  1]
 [ 1  0  0  0  1  0]
 [ 0  0  0  0  0  6]]
accuracy is 0.3953488372093023

```

✓ SVC:

precision	recall	f1-score	support		
	1	0.21	1.00	0.35	9
	2	0.00	0.00	0.00	19
	3	0.00	0.00	0.00	5
	5	0.00	0.00	0.00	2
	6	0.00	0.00	0.00	2
	7	0.00	0.00	0.00	6
accuracy				0.21	43
macro avg		0.03	0.17	0.06	43
weighted avg		0.04	0.21	0.07	43

```

[[ 9  0  0  0  0  0]
 [19  0  0  0  0  0]
 [ 5  0  0  0  0  0]
 [ 2  0  0  0  0  0]
 [ 2  0  0  0  0  0]
 [ 6  0  0  0  0  0]]
accuracy is 0.20930232558139536

```

✓ Linear SVC(max_iter=1000):

precision	recall	f1-score	support	
	1	0.31	1.00	0.47
	2	1.00	0.11	0.19
	3	0.00	0.00	0.00
	5	0.50	0.50	0.50
	6	0.50	1.00	0.67
	7	1.00	1.00	1.00
accuracy				0.47
macro avg		0.55	0.60	0.47
weighted avg		0.69	0.47	0.38


```
[[ 9  0  0  0  0  0]
 [15  2  0  1  1  0]
 [ 5  0  0  0  0  0]
 [ 0  0  0  1  1  0]
 [ 0  0  0  0  2  0]
 [ 0  0  0  0  0  6]]
```

accuracy is 0.46511627906976744

✓ Complement naïve bayes works well for imbalanced dataset. Overall Linear SVC has better accuracy because data may be linearly related.