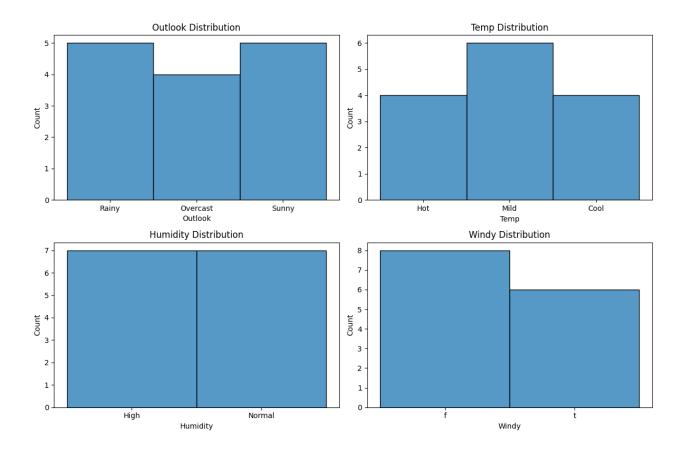
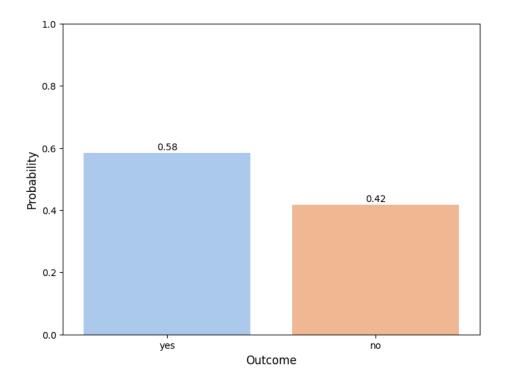
### 2 Naive Bayes

### 2.1 Data Pre-Processing & Visualisation

Data Distribution Plots after CATEGORICAL ENCODING



### 2.2 Prior Probabilities



#### 2.3 Likelihood Probabilities

OUTLOOK	P(yes)	P(no)
Overcast	0.428571	0.000000
Rainy	0.285714	0.600000
Sunny	0.285714	0.400000

# 2.4 Making Predictions based on proportional probabilities

Test-Case	P(yes)	P(no)	Prediction
1	0.024295432458697763	0.005333333333333333	yes
2	0.005830903790087463	0.0	yes

#### 2.5 Likelihood Probabilities after Laplace smoothing

1.  $\alpha = 0.01$ 

OUTLOOK	P(yes)	P(no)
Overcast	0.428165	0.001988
Rainy	0.285917	0.598410
Sunny	0.285917	0.399602

#### Making Predictions based on proportional probabilities after Laplace smoothing

Test-Case	P(yes)	P(no)	Prediction
1	0.02428843241892867	0.005359877943368312	yes
2	0.005854456555063937	0.00015854490113897748	yes

#### 2. $\alpha = 0.1$

OUTLOOK	P(yes)	P(no)
Overcast	0.424658	0.018868
Rainy	0.287671	0.584906
Sunny	0.287671	0.396226

Temp	P(yes)	P(no)
Cool	0.424658	0.207547
Hot	0.287671	0.396226
Mild	0.287671	0.396226

## Making Predictions based on proportional probabilities after Laplace smoothing

Test-Case	P(yes)	P(no)	Prediction
1	0.02422060697754426	0.005588315294667817	yes
2	0.006062135771772901	0.001464182382915439	yes

3.  $\alpha = 1$ 

OUTLOOK	P(yes)	P(no)
Overcast	0.400000	0.125000
Rainy	0.300000	0.500000
Sunny	0.300000	0.375000

Temp	P(yes)	P(no)
Cool	0.400000	0.250000
Hot	0.300000	0.375000
Mild	0.300000	0.375000

## Making Predictions based on proportional probabilities after Laplace smoothing

Test-Case	P(yes)	P(no)	Prediction
1	0.02333333333333333333	0.007174744897959183	yes
2	0.007777777777777776	0.007971938775510204	no

4.  $\alpha = 10$ 

OUTLOOK	P(yes)	P(no)
Overcast	0.351351	0.285714
Rainy	0.324324	0.371429
Sunny	0.324324	0.342857

Temp	P(yes)	P(no)
Cool	0.351351	0.314286
Hot	0.324324	0.342857
Mild	0.324324	0.342857

## Making Predictions based on proportional probabilities after Laplace smoothing

Test-Case	P(yes)	P(no)	Prediction
1	0.018937856775694616	0.010344489795918368	yes
2	0.0131302473644816	0.011885714285714286	yes

#### 5. $\alpha = 100$

OUTLOOK	P(yes)	P(no)
Overcast	0.335505	0.327869
Rainy	0.332248	0.337705
Sunny	0.332248	0.334426

Temp	P(yes)	P(no)
Cool	0.335505	0.331148
Hot	0.332248	0.334426
Mild	0.332248	0.334426

## Making Predictions based on proportional probabilities after Laplace smoothing

Test-Case	P(yes)	P(no)	Prediction
1	0.016568313127134977	0.01142360680677749	yes
2	0.015788362604415428	0.01164533786645278	yes

#### 2.6 observed differences in predictions

- We see that probability of not playing while OUTLOOK is Overcast is zero
- Assigning zero probability means ruling out unseen events from consideration, which is not be desired
- After Laplace smoothing with smoothing parameter alpha we overcome this problem
- Hence probability of not playing in Test-Case 2 is not 0 anymore and makes more sense
- We see a regular trend not matter what alpha value we have taken except for the case where alpha = 1
- Here the probability of not playing overtakes that of playing in test case 2 and hence accuracy drops down from 1 to 0.5
- Later on (alpha more than 1) the overtaking doesn't occur
- hence hyper-parameter tuning has also been done