## CMPE 462 Project Step 2

For the step 2 we did all of the coding together, sharing screens.

To classify the reviews we have tried 3 methods.

Multinomial Naive Bayes (MNB):

```
0.65466666666666
      precision recall f1-score support
        0.63 0.72
                     0.67
                           245
        0.76 0.79
                     0.77
                            259
        0.55 0.45
                     0.50
                           246
                    0.65 750
 macro avg 0.65 0.65 0.65
                               750
weighted avg 0.65 0.65 0.65
                               750
```

## Support Vector Classifier (SVC):

```
0.636
      precision recall f1-score support
         0.65
               0.68
                     0.66
                             245
         0.71
               0.75
                      0.73
                             259
         0.53
               0.48
                     0.50
                             246
                     0.64 750
 accuracy
            0.63 0.63 0.63
                                750
 macro avg
weighted avg 0.63 0.64 0.63
                                750
```

## K Nearest Neighbors (KNN):

```
0.5213333333333333
      precision recall f1-score support
         0.53 0.58 0.56
                            245
         0.55 0.65
                     0.59
                            259
         0.46 0.33 0.38
                            246
                    0.52
                           750
 accuracy
 macro avg 0.51 0.52 0.51
                               750
weighted avg 0.51 0.52 0.51
                                750
```

Among those 3, the best performing one was the multinomial naive bayes, as can be seen from the metrics. We think that is because dictionary of words used in different classes of reviews is pretty specific to each class.

## Our result on val data:

0.624	precisio	on rec	all f1-s	core	suppo	rt	
_	0 0.6 1 0.7 2 0.5	4 0.	72 (	0.64 0.73 0.50	250 250 250		
accur macro weighte	avg	0.62 0.62	0 0.62 0.62	.62 0.6 ! 0.	750 2 62	750 750	