CS 573 – Homework 3

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Problem 1

Code

The screenshot below shows the output of LR/SVM models for test cases of Homework 2

```
priyank@priyank:~/Desktop/Dropbox/Spring 2017/DM/My HW Solutions/HW3$ python hw3.py yelp_train1.txt yelp_test1.txt 2
ZERO-ONE-LOSS-SVM 0.11
priyank@priyank:~/Desktop/Dropbox/Spring 2017/DM/My HW Solutions/HW3$
```

Figure 1: Output for testcases of Homework 2

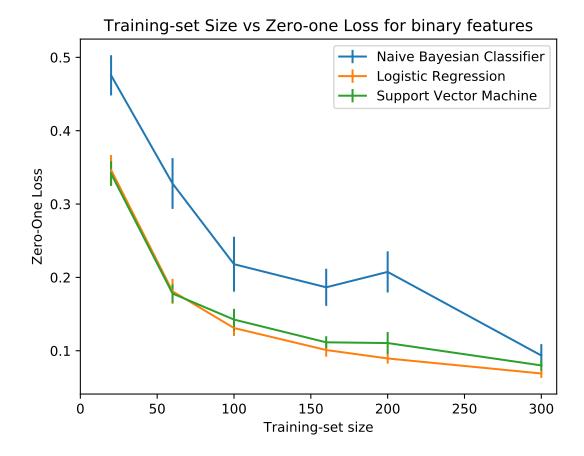


Figure 2: TSS vs Zero-one loss for NBC vs LR vs SVM with binary features

Problem 2

Analysis

1.

(a)

Figure 2 shows the learning curves for the three models with binary features.

(b)

Null Hypothesis: Logistic regression performs as good as Naive Bayesian Classifier Alternate Hypothesis: There is a performance difference between Logistic Regression and Naive Bayesian Classifier models

OR

Null Hypothesis: Support Vector Machine performs as good as the Naive Bayesian Classifier Alternate Hypothesis: There is a performance difference between Support Vector Machine and Naive Bayesian Classifier models

OR

Null Hypothesis: Support Vector Machine performs as good as Logistic Regression Alternate Hypothesis: There is a performance difference between Support Vector Machine and Logistic Regression models

I select the significance level as $\alpha = 0.05$

Model/TSS	20	60	100	160	200	300
NBC	0.4755	0.328	0.218	0.1865	0.2075	0.0935
LR	0.3465	0.181	0.131	0.101	0.0895	0.069
SVM	0.3415	0.178	0.1425	0.1115	0.1105	0.08

Table 1: Mean zero-one loss for the three models with binary features

```
priyank@priyank: ~/Desktop/Dropbox/Spring 2017/DM/My HW Solutions/HW3
   200
    300
    300
    300
    300
    300
    300
    300
leans with NBC with binary features [0.4754999999999999, 0.3280000000000001, 0.2180000000000003, 0.1865, 0.20
7499999999996, 0.0935]
Means with NBC with features with 3 values [0.444999999999995, 0.38, 0.3175, 0.23349999999999, 0.2400000000
9999999999999, 0.10100000000000001, 0.07599999999999999999
Means with SVM with binary features [0.34150000000000003, 0.17799999999999, 0.1425000000000000, 0.1115, 0.11
50000000000002, 0.0975000000000003, 0.08050000000000002]
ttest_rel for NBC vs LR: t = 5.55279  p = 0.00260349
ttest_rel for NBC vs LR: t = 5.55279
               NBC vs SVM: t = 4.56679
SVM vs LR: t = 1.90513
ttest_rel for
                                          p = 0.0060195
                                         p = 0.115104
     _rel for NBC with binary-valued featuresvs NBC with features with three values: t = -2.37857
_rel for LR with binary-valued featuresvs LR with features with three values: t = 0.891802
_rel for SVM with binary-valued featuresvs SVM with features with three values: t = 1.21569
                                                                                                           p = 0.413353
      nk@priyank:~/Desktop/Dropbox/Spring 2017/DM/My HW Solutions/HW3$
```

Figure 3: Hypothesis Testing Output

(c)

Table 1 shows the mean zero-one loss across 10 folds for the three models with binary features. As we can see from figure 3:

For the null hypothesis: "Logistic Regression (LR) performs as good as Naive Bayes Classifier (NBC)", the probability is p=0.0026, which is less than the significance level. So we reject the null-hypothesis. This implies that the alternate hypothesis is true: There is a performance difference between Logistic Regression and Naive Bayesian Classifier

OR

For the null hypothesis: "Support Vector Machine (SVM) performs as good as Naive Bayesian Classifier", the probability is p=0.006, which is less than the significance level. So we reject the null-hypothesis. This implies that the alternate hypothesis is true: There is a performance difference between Support Vector Machine and Naive Bayesian Classifier.

OR

For the null hypothesis: "Support Vector Machine performs as good as Logistic Regression", the probability is p = 0.115, which is greater than the significance level. So we accept the null-hypothesis. This implies that the null-hypothesis is true: Support Vector Machine performs as

Training-set Size vs Zero-one Loss for features with three values Naive Bayesian Classifier 0.45 Logistic Regression Support Vector Machine 0.40 0.35 Zero-One Loss 0.30 0.25 0.20 0.15 0.10 0.05 0 50 100 200 150 250 300

Figure 4: TSS vs Zero-one loss for NBC vs LR vs SVM with features with three values

Training-set size

good as Logistic Regression"

(a) Figure 4 shows the learning curves for the three models with features with three values.

(b)

Null Hypothesis: LR with binary features performs as good as LR with features with three

Alternate Hypothesis: There is a performance difference between LR with binary features and LR with features with three values

OR

Null Hypothesis: SVM with binary features performs as good as SVM with features with three values

Alternate Hypothesis: There is a performance difference between SVM with binary features and SVM with features with three values

I select the significance level as $\alpha = 0.05$

(c)

Table 2 shows the mean zero-one loss across 10 folds for the three models with features with

Model/TSS	20	60	100	160	200	300
NBC	0.445	0.38	0.3175	0.2335	0.24	0.137
LR	0.297	0.169	0.134	0.0925	0.101	0.076
SVM	0.2875	0.1785	0.1425	0.1125	0.0975	0.08

Table 2: Mean zero-one loss for the three models with features with three values

three values. As we can see from figure 3:

For the null hypothesis: "LR with binary features performs as good as LR with features with three values", the probability is p=0.413, which is above the significance level. So we accept the null-hypothesis. This implies that the null hypothesis is true: LR with binary features performs as good as LR with features with three values.

OR

For the null hypothesis: "SVM with binary features performs as good as SVM with features with three values", the probability is p=0.278, which is above the significance level. So we accept the null-hypothesis. This implies that the null hypothesis is true: SVM with binary features performs as good as SVM with features with three values.