

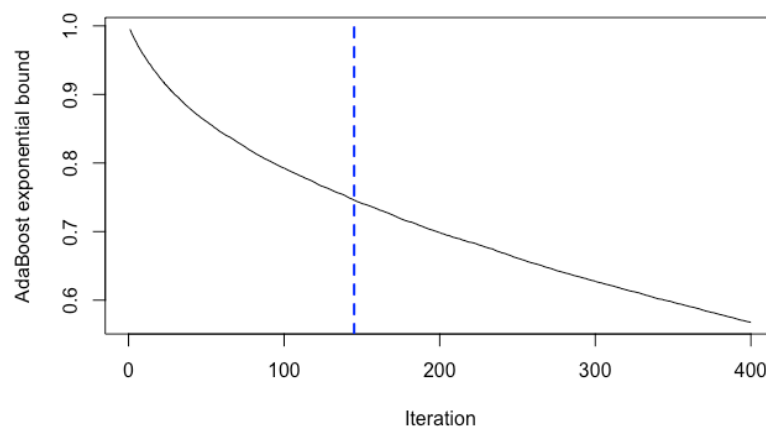
Puppy or Fried Chicken?

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Summary:

In this project, we created a classifier for grey images of puppies versus images of fried chickens. We tried different features (SIFT, LBP) and different classifiers (GBM, BP Neural Networks, SVM, Random Forest, Logistic Regression and Majority Vote). When pursuing low error rate, we also keep an eye on processing time.

Baseline: GBM + SIFT



Tune parameters: n.trees=145, shrinkage=0.1

Error rate 27.5%

Training time 76s

Other models + SIFT

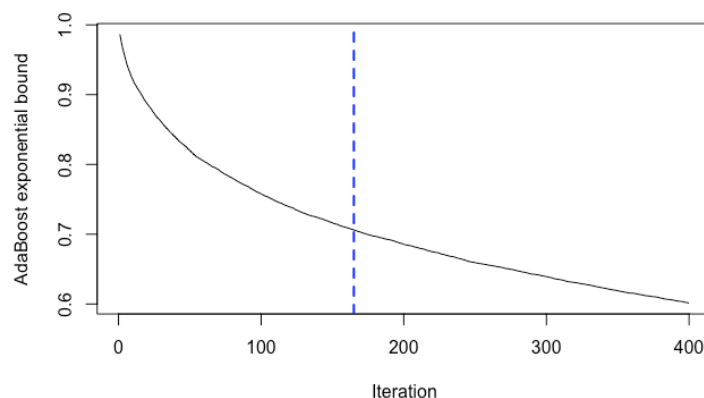
We tried to apply other models on 5000-dimensional SIFT features. When the accuracy rate increased to ~80%, the processing time increased dramatically. So we used PCA to reduce the dimension of SIFT features. However, when the dimension decreased to 500, the results of models didn't seem satisfying. Thus, we started to explore other features.

Local Binary Patterns (LBP)

- Divide the examined window into cells (e.g. 16x16 pixels for each cell).
- For each pixel in a cell, compare the pixel to each of its 8 neighbors
- Where the center pixel's value is greater than the neighbor's value, write "0". Otherwise, write "1". This gives an 8-digit binary number.
- Compute the histogram over the cell. This histogram can be seen as a 256-dimensional feature vector.
- Optionally normalize the histogram to 59-dimensional feature vector.
- Concatenate histograms of all cells. This gives a feature vector for the entire window.

Then we extracted LBP features in MATLAB. The processing time of 2000 images is 210s. The column dimension of the result feature matrix is 59.

GBM+LBP



Tune parameters: n.trees=165, shrinkage=0.1

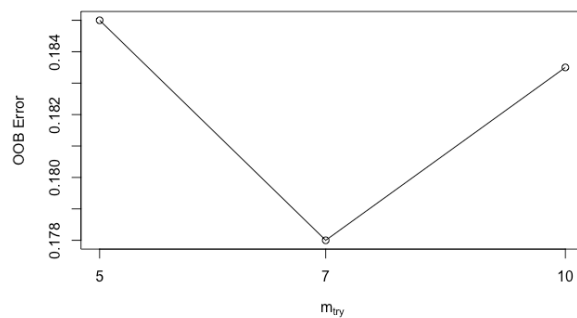
```
gbm.pre  0  1
         0 157 42
         1  45 156
```

BP Neural Networks + LBP

Tune Parameters: size = 1, decay = 0.01

| bp.pre | 0 | 1 |
|--------|-----|-----|
| 0 | 175 | 23 |
| 1 | 27 | 175 |

Random Forest + LBP



Tune Parameter: m.try=7

| rf.pre | 0 | 1 |
|--------|-----|-----|
| 0 | 169 | 35 |
| 1 | 33 | 163 |

SVM + LBP

Tune Parameters: cost=10, gamma=0.01

| svm.pre | 0 | 1 |
|---------|-----|-----|
| 0 | 186 | 25 |
| 1 | 16 | 173 |

Logistic Regression + LBP

| log.pre | 0 | 1 |
|---------|-----|-----|
| 0 | 177 | 25 |
| 1 | 25 | 173 |

Majority Vote + LBP

We choose three models with higher accuracy for a majority vote.

```
pre  0   1
     0 177  23
     1  25 175
```

Summary of Models with LBP features

| | Parameters | CV Error Rate | Training Time |
|--|------------------------------|---------------|---------------|
| GBM | n.trees=133 shrinkage=0.1 | 23% | 1.08s |
| Neural Networks | size = 1 decay = 0.01 | 13.5% | 0.23s |
| Radom Forest | m.try=7 | 19.75% | 35.43s |
| SVM | cost=10 gamma=0.01 | 12.06% | 0.86s |
| Logistic Regression | | 13.56% | 0.09s |
| Majority Vote (NN, SVM, Log) | | 13.12% | |

Final Model

We choose Majority Vote as our final model. Since training time of each model is very short, time won't be a problem for majority vote. Although we sacrifice little accuracy, we can get a more robust model.