DSC 255: Machine learning

Worksheet

- 1. Regression with one predictor variable
 - (a) We will predict the mean of the y-values: $\hat{y} = (1+3+4+6)/4 = 3.5$. The MSE of this prediction is exactly the variance of the y-values, namely:

$$MSE = \frac{(1 - 3.5)^2 + (3 - 3.5)^2 + (4 - 3.5)^2 + (6 - 3.5)^2}{4} = 3.25.$$

(b) If we simply predict x, the MSE is

$$\frac{1}{4} \sum_{i=1}^{4} (y^{(i)} - x^{(i)})^2 = \frac{1}{4} \left((1-1)^2 + (1-3)^2 + (4-4)^2 + (4-6)^2 \right) = 2.$$

(c) We saw in class that the MSE is minimized by choosing

$$a = \frac{\sum_{i} (y^{(i)} - \overline{y})(x^{(i)} - \overline{x})}{\sum_{i} (x^{(i)} - \overline{x})^{2}}$$
$$b = \overline{y} - a\overline{x}$$

where \overline{x} and \overline{y} are the mean values of x and y, respectively. This works out to a=1,b=1; and thus the prediction on x is simply x+1. The MSE of this predictor is:

$$\frac{1}{4}\left(1^2 + 1^2 + 1^2 + 1^2\right) = 1.$$

- 2. Optimality of the mean.
 - (a) $dL/ds = -2(x_1 + \cdots + x_n)/n + 2s$.
 - (b) Setting dL/ds = 0, we get $s = (x_1 + \cdots + x_n)/n$.
- 3. We would write the loss induced by a linear predictor $w \cdot x + b$ as

$$L(w,b) = \sum_{i=1}^{n} |y^{(i)} - (w \cdot x^{(i)} + b)|.$$

4. Inherent uncertainty. This is somewhat subjective, but (b), (d) seem pretty clear-cut cases where perfect predictions are not possible.

1

5. Logistic regression.

- (a) The decision boundary is the hyperplane given by c = 1/2.
- (b) c = 3/4 yields a hyperplane that is parallel to the decision boundary.
- (c) c = 1/4 yields a hyperplane parallel to the decision boundary. It is on the opposite side to (b) and the same distance from the decision boundary.
- 6. Discovering relevant features in regression.
 - (a) A sensible strategy is to do linear regression using the Lasso, and to choose a regularization constant λ that yields roughly 10 non-zero coefficients.
 - (b) First value of λ which gave nonzero coefficients only for 10 features is 0.4. This yielded the following features (numbering starting at 1): 2, 3, 5, 7, 11, 13, 17, 19, 23, 27.
- 7. Binary logistic regression. See the accompanying notebook heart-soln.ipynb. The results obtained depend on the random partition of the data into training and test sets. In one particular run, we got the following results.
 - (a) Coefficients:

```
[0.014, -0.928, 0.588, -0.010, -0.004, -0.269, 0.358, 0.027, -0.869, -0.685, 0.285, -0.736, -0.567]
```

The three features that were most influential in the model – the features with the highest absolute values – were 1 (sex), 8 (exang), and 11 (ca).

- (b) The test error was 18.45%.
- (c) The 5-fold cross validation error was 16.00%, which is fairly close to the test error.