Machine Learning - Homework 2

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Problem 1

(a)
$$\mu_{y,j} = \sum_{i=1}^{n} \frac{\mathbb{1}\{y_i = y\}x_{i,j}}{\mathbb{1}\{y_i = y\}}$$

(b) Source code: hw2_p1b.m

Training Error Rate: 21.6257% Test Error Rate: 37.6016% Running Time: 30 secs

(c) Source Code: hw2_p1c.m

Training Error Rate: 5.7794%Test Error Rate: 13.1383%

Running Time: 1.15 secs

(d) Source Code: hw2_p1d.m $\alpha_0 = -20.7849$

20 most positive words (corresponding to 20 largest values of α_j)

- 1: 'firearms'
- 2: 'occupied'
- 3: 'israelis'
- 4: 'serdar'
- 5: 'argic'
- 6: 'ohanus'
- 7: 'appressian'
- 8: 'sahak'
- 9: 'melkonian'
- 10: 'villages'
- 11: 'cramer'
- 12: 'armenia'
- 13: 'cpr'
- 14: 'sdpa'
- 15: 'handgun'
- 16: 'optilink'
- 17:'palestine'

- 18: 'firearm'
- 19: 'budget'
- 20: 'arabs'

20 most negative words (corresponding to 20 smallest values of

- α_j)
- 1: 'athos'
- 2: 'atheism'
- 3: 'atheists'
- 4: 'clh'
- 5: 'teachings'
- 6: 'revelation'
- 7: 'testament'
- 8: 'livesey'
- 9: 'atheist'
- 10: 'wpd'
- 11: 'solntze'
- 12: 'scriptures'
- 13: 'theology'
- 14: 'believers'
- 15: 'ksand'
- 16: 'alink'
- 17: 'benedikt'
- 18: 'jesus'
- 19: 'prophet'
- 20: 'mozumder'

Problem 2

(a) The classifier f^* predicts 1 when $c_{\frac{3}{2}}^2N(0,1) \leq \frac{1}{3}N(2,\frac{1}{4})$

Thus, the range in which f^* predicts 1 can be given by the roots of:

$$c_{\frac{3}{3}}N(0,1) = \frac{1}{3}N(2,\frac{1}{4})$$

Thus, we get the quadratic equation

$$3x^2 - 16x + 16 + 2log(c) = 0$$

which give the roots:

$$x = \frac{16 \pm \sqrt{256 - 12(16 + 2log(c))}}{6}$$
$$x = \frac{16 \pm \sqrt{64 - 24log(c)}}{6}$$

For $1 \le c \le 14$,

 f^* predicts 1 when x is in the interval $[\frac{16-\sqrt{64-24log(c)}}{6},\frac{16+\sqrt{64-24log(c)}}{6}]$

(b) For $c \ge 15$, the roots are imaginary i.e. they don't exist. Thus, the classifier will always predict 0.

Problem 3

(a) Covariance matrix can be written as: $\Sigma = U\Lambda U^T \text{ where } U \text{ is orthonormal}$ Thus, $\Sigma + \sigma^2 I = U\Lambda U^T + \sigma^2 U U^T$ $\Sigma + \sigma^2 I = U(\Lambda + \sigma^2 I) U^T$ Thus, the eigenvalues of $\Sigma + \sigma^2 I \text{ are}$

$$\lambda_1 + \sigma^2, \lambda_2 + \sigma^2, \dots, \lambda_d + \sigma^2$$

(b) Similarly, we get the eigenvalues of $(\Sigma + \sigma^2 I)^{-2}$ as

$$(\lambda_1 + \sigma^2)^{-2}, (\lambda_2 + \sigma^2)^{-2}, \dots, (\lambda_d + \sigma^2)^{-2}$$

References

- [1] https://www.cs.ubc.ca/~murphyk/Teaching/Stat406-Spring08/Lectures/linalg1.pdf
- [2] http://cs229.stanford.edu/section/gaussians.pdf
- [3] https://ocw.mit.edu/courses/mathematics/
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