

# CSC311 A3 Non-programming

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## Question 1(e).

Suppose we have that

$$x^{(i)} - \mu = M^{(i)}$$

then we can say from the empirical covariance matrix, we have,

$$\frac{1}{N} \sum_{i=1}^N (x^{(i)} - \mu)(x^{(i)} - \mu)^T$$

Using what we suggested above:

$$= \frac{1}{N} \sum_{i=1}^N M^{(i)} M^{(i)T}$$

Using definition of matrix

$$= \frac{1}{N} \sum_{i=1}^N M_{ij} M_{ij}^T$$

Using definition of transpose

$$\begin{aligned} &= \frac{1}{N} \sum_{i=1}^N M_{ij} M_{ji} \\ &= \frac{1}{N} \sum_{i=1}^N M_{ji} M_{ij} \\ &= \frac{1}{N} [M^T M]_{jj} \end{aligned}$$

This proves that this is a vectorized form of the empirical covariance formula and that it is symmetric

## Question 2(c).

Overfitting occurs when the regularization parameter is  $< 10^{-4}$ . In the graph, you can see the training accuracy vs. validation accuracy to drastically spread out. Underfitting occurs when the regularization parameter is  $> 10^{-2}$ . You can see if the figure that the training vs validation accuracy are too close together as the regularization parameter gets greater.

## Question 2(e).

Overfitting occurs when K is  $> 10$ . In the graph, you can see the training accuracy vs. validation accuracy to drastically spread out. Underfitting occurs in when  $K < 5$ . You can see if the figure that the training vs validation accuracy are too close together as K gets greater.