

# CMPUT 466 Assignment 1

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## Part 1

$$\frac{1}{m} \sum_{i=1}^m (x_i^2)$$

Because we'd like to square each  $x_i$  and sum them, we can simply take the dot product of the  $\mathbf{x}$  vector transposed with itself not transposed, then multiply the resulting scalar (or, 1x1 matrix) by  $\frac{1}{m}$ . i.e.

$$= \frac{1}{m} \mathbf{x}^\top \cdot \mathbf{x}$$

## Part 2

$$\frac{1}{m} \sum_{i=1}^m \left[ (x_i - \mu)^2 \right]$$

Here, we want to subtract a bias term  $\mu$  from each  $x_i$   
Let's define a vector  $\mathbf{u} \in \mathbb{R}^m$  where each  $u_i = -\mu$ , i.e.

$$\mathbf{u} = (-\mu, -\mu, -\mu, \dots, -\mu)^\top \in \mathbb{R}^m$$

Let's define a second vector  $\mathbf{y} \in \mathbb{R}^m$ :

$$\mathbf{y} = \mathbf{x} + \mathbf{u} \in \mathbb{R}^m$$

This way, each  $y_i = x_i - \mu$ .

finally, we can represent the biased empirical estimate of the second-order central moment as:

$$= \frac{1}{m} \mathbf{y}^\top \cdot \mathbf{y}$$