

## Quiz 3

# 03 Support Vector Machines, Convex Optimization

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## 1 Convexity

Prove that if  $f(x)$  is convex,  $f(\alpha x + \beta)$  is convex for scalars  $\alpha, \beta$ . Hint: If you're stuck, take  $g(x) = \alpha x + \beta$ .

**Solution:** Recall that a function is convex if

$$\forall x_1, x_2 \in \mathbb{R}, t \in [0, 1], f((1-t)x_1 + tx_2) \leq (1-t)f(x_1) + tf(x_2)$$

Take  $g(x) = \alpha x + \beta$  and the following  $f(g(x))$ ; we first prove a lemma.

$$\begin{aligned} & f(g((1-t)x_1 + tx_2)) \\ &= f(\alpha((1-t)x_1 + tx_2) + \beta) \\ &= f((1-t)\alpha x_1 + (1-t)\beta + t\alpha x_2 + t\beta) \\ &= f((1-t)(\alpha x_1 + \beta) + t(\alpha x_2 + \beta)) \\ &= f((1-t)g(x_1) + tg(x_2)) \end{aligned}$$

Then, simply apply the convexity of  $f$ .

$$\begin{aligned} f(g((1-t)x_1 + tx_2)) &= f((1-t)g(x_1) + tg(x_2)) \\ &\leq (1-t)f(g(x_1)) + tf(g(x_2)) \end{aligned}$$

## 2 Linear Algebra

Compute the variance of  $u \in \mathbb{R}^n$ , where  $u \sim (0, I)$ . This notation simply means that  $u$  is sampled from some distribution with mean 0, where the covariance matrix of  $u$  is  $I$ . Consider  $A \in \mathbb{R}^{n \times n}$ . Compute variance of  $y = Au$ .

**Solution:**

$$\begin{aligned} & E[(Au - \mu)^T(Au - \mu)] \\ &= E[(Au)^T Au] \\ &= E[u^T A^T Au] \\ &= E[\text{Tr}(A^T A u u^T)] \\ &= \text{Tr}(E[A^T A u u^T]) \\ &= \text{Tr}(A^T A) \\ &= \|A\|_F^2 \end{aligned}$$

As it turns out, this is precisely the Frobenius norm.