

# Homework 0

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Warm-up

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**Exercise 1:** An  $n \times n$  symmetric real matrix  $M$  is said to be positive-definite (p.d) if

$$x^\top Mx > 0, \forall 0 \neq x \in \mathbb{R}^n.$$

It is said to be positive-semidefinite (p.s.d) if

$$x^\top Mx \geq 0, \forall x \in \mathbb{R}^n$$

1. Show that if  $M$  is p.s.d and  $\varepsilon > 0$ , then  $M + \varepsilon I$  is p.d.
2. Deduce that  $M^\top M + \varepsilon I$  is p.d for any matrix  $M$  and any  $\varepsilon > 0$
3. Let  $\text{Null}(M) = \{x \in \mathbb{R}^n : Mx = 0\}$ . Show that if  $\text{Null}(M) = \{0\}$  then  $M^\top M$  is p.d.

**Exercise 2:** Let  $\Omega$  be a finite set and  $X : \Omega \mapsto \mathbb{R}, w \in \Omega \mapsto X(w)$  be a real valued function. We further suppose that  $X$  has a discrete number of output given by  $X(\Omega) = \{x_1, \dots, x_n\}$ . We refer to this  $X$  as a random variable (rv). We define its expectation as

$$E(X) = \sum_{i=1}^n x_i P[X = x_i]$$

. Let  $Y$  be another r.v. We define the conditional expectation of  $X$  on  $Y = y_j$  as

$$E[X/Y = y_j] = \sum_{i=1}^n x_i P[X = x_i / Y = y_j]$$

. It is the expected value of  $X$  but given that  $Y = y_j$  has occurred.

1. Show that  $E[E[X/Y]] = E[X]$
2. Let  $V = (V_1, \dots, V_n)^\top, W = (W_1, \dots, W_n)^\top$  be two independent r.v. Show that

$$E[\langle V, W \rangle / W] = \langle E[V], W \rangle$$

**Exercise 3:** All the algorithm we will discuss in the course generate a sequence of random vectors  $x_t$  that converge to a desired  $x^*$  in some sense. In particular, we focus on two forms of convergence, either showing that the difference of function values converges

$$f(x_t) - f^* \rightarrow 0$$

or norm difference of the iterates converges

$$\|x_t - x^*\| \rightarrow 0$$

Two important questions:

- How fast is this convergence ?
  - given an  $\varepsilon$  accuracy how many iterations  $t$  are needed before  $f(x_t) - f^* < \varepsilon$  or  $\|x_t - x^*\| < \varepsilon$ .
1. Consider a sequence  $(u_t)_t \in \mathbb{R}_+$  of positive scalars that converge to zero according to

$$u_t \leq \frac{C}{t}$$

where  $C > 0$ . Given  $\varepsilon > 0$ , show that

$$t \geq \frac{C}{\varepsilon} \Rightarrow u_t < \varepsilon$$

We refer to this result as a  $\mathcal{O}(1/\varepsilon)$  iteration complexity. This type of convergence is known as sublinear convergence.

2. Consider a sequence  $(u_t)_t \in \mathbb{R}_+$  of positive scalars that converge to zero according to

$$u_t \leq q^t u_0$$

where  $0 \leq q < 1$ . Given  $1 > \varepsilon > 0$ , and using the fact that

$$\frac{1}{1-q} \cdot \log\left(\frac{1}{q}\right) \geq 1,$$

show that

$$t \geq \frac{1}{1-q} \cdot \log\left(\frac{1}{\varepsilon}\right) \Rightarrow u_t < \varepsilon u_0$$

We refer to this result as a  $\mathcal{O}(\log(1/\varepsilon))$  iteration complexity. This type of convergence is known as linear convergence at a rate  $q^t$ .

**Exercise 1:** Compute  $\nabla f$  and  $\nabla^2 f$  for the following functions  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ .

1.  $f(x) = a^\top x$
2.  $f(x) = x^\top M x$
3.  $f(x) = x^\top M x + b^\top x + c$