Please submitted by Saturday, 29 July 2022, 11am, right before the discussion hour.

- 1. Attempt all problems. There is no penalty for submitting incorrect attempts
- 2. However, plagiarism will result in serious penalties, such as an F grade.
- 1. Using concavity of the logarithm, show that  $x^{\theta}y^{1-\theta} \leq \theta x + (1-\theta)y$ .
- 2. Show that the harmonic mean  $f(x) = (\sum_{i=1}^{n} 1/x_i)^{-1}$  is concave.
- 3. Prove the reverse Jensen's inequality for a convex f with dom  $f = \mathbb{R}^n$ ,  $\lambda_i > 0$  and  $\lambda_1 \sum_{i=2}^n \lambda_i = 1$

$$f(\lambda_1 \mathbf{x}_1 - \lambda_2 \mathbf{x}_2 - \dots - \lambda_n \mathbf{x}_n) \ge \lambda_1 f(\mathbf{x}_1) - \lambda_2 f(\mathbf{x}_2) - \dots - \lambda_n f(\mathbf{x}_n)$$
 (2)

- 4. Give an example of a function  $f(\mathbf{x})$  whose epigraph is (a) half-space, (b) norm cone, and (c) polyhedron.
- 5. Let  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n_{++}$  be two vectors. We need to show that the Itakura-Saito distance, defined as

$$D_{IS}(\mathbf{x}, \mathbf{y}) = \sum_{i=1}^{n} \left( \frac{x_i}{y_i} - \log \left( \frac{x_i}{y_i} \right) - 1 \right)$$
 (5)

is always positive, using the following steps:

(a) Show that for a convex differentiable function f, the Bregman divergence,

$$D(\mathbf{x}, \mathbf{y}) = f(\mathbf{x}) - f(\mathbf{y}) - \nabla f(\mathbf{y})^{T} (\mathbf{x} - \mathbf{y})$$
(6)

is always non-negative.

- (b) Show that for the convex function  $f(\mathbf{x}) = -\sum_{i=1}^{n} \log(x_i)$ , it holds that  $D(\mathbf{x}, \mathbf{y}) = D_{IS}(\mathbf{x}, \mathbf{y})$ .
- (c) Along similar lines, prove that the generalized KL divergence

$$D_{KL}(\mathbf{x}, \mathbf{y}) = \sum_{i=1}^{n} \left( x_i \log \left( \frac{x_i}{y_i} \right) - x_i + y_i \right)$$
 (7)

is always positive.