

**Homework 3***Due date: Wednesday, Nov. 28, 2018 (at the beginning of class)*

You are allowed to drop 1 subproblem without penalty. In addition, up to 1 bonus point will be awarded to each subproblem for clean, well-organized, and elegant solutions.

**1. Gaussian graphical models (20 points)**

- (a) Consider a  $p$ -dimensional Gaussian vector  $\mathbf{x} \sim \mathcal{N}(\mathbf{0}, \mathbf{\Sigma})$ . For any  $1 \leq u, v \leq p$ , show that

$$x_u \perp\!\!\!\perp x_v \mid \mathbf{x}_{\mathcal{V} \setminus \{u, v\}}$$

(namely,  $x_u$  and  $x_v$  are conditionally independent given all other variables) if and only if  $\Theta_{u,v} = 0$ . Here,  $\mathbf{\Theta} = \mathbf{\Sigma}^{-1}$ .

- (b) In graphical lasso, the objective function includes a term  $\log \det \mathbf{\Theta}$ . Show that  $g(\mathbf{\Theta}) := \log \det(\mathbf{\Theta})$  ( $\mathbf{\Theta} \succ \mathbf{0}$ ) is a concave function.

Hint: A function  $g(\mathbf{\Theta})$  is concave if  $h(t) := g(\mathbf{\Theta} + t\mathbf{V})$  is concave for all  $t$  and  $\mathbf{V}$  obeying  $\mathbf{\Theta} + t\mathbf{V} \succ \mathbf{0}$ .

**2. Restricted isometry properties (30 points)**

Recall that the restricted isometry constant  $\delta_s \geq 0$  of  $\mathbf{A}$  is the smallest constant such that

$$(1 - \delta_s) \|\mathbf{x}\|_2^2 \leq \|\mathbf{A}\mathbf{x}\|_2^2 \leq (1 + \delta_s) \|\mathbf{x}\|_2^2 \quad (1)$$

holds for all  $s$ -sparse vector  $\mathbf{x} \in \mathbb{R}^p$ .

- (a) Show that

$$|\langle \mathbf{A}\mathbf{x}_1, \mathbf{A}\mathbf{x}_2 \rangle| \leq \delta_{s_1+s_2} \|\mathbf{x}_1\|_2 \|\mathbf{x}_2\|_2$$

for all pairs of  $\mathbf{x}_1$  and  $\mathbf{x}_2$  that are supported on disjoint subsets  $S_1, S_2 \subset \{1, \dots, n\}$  with  $|S_1| \leq s_1$  and  $|S_2| \leq s_2$ .

- (b) For any  $\mathbf{u}$  and  $\mathbf{v}$ , show that

$$|\langle \mathbf{u}, (\mathbf{I} - \mathbf{A}^\top \mathbf{A})\mathbf{v} \rangle| \leq \delta_s \|\mathbf{u}\|_2 \cdot \|\mathbf{v}\|_2,$$

where  $s$  is the cardinality of  $\text{support}(\mathbf{u}) \cup \text{support}(\mathbf{v})$ .

- (c) Suppose that each column of  $\mathbf{A}$  has unit norm. Show that  $\delta_2 = \mu(\mathbf{A})$ , where  $\mu(\mathbf{A})$  is the mutual coherence of  $\mathbf{A}$ .

**3. Statistical dimension (10 points)** Recall that for any convex cone  $\mathcal{K}$ , its statistical dimension and Gaussian width are defined respectively as

$$\text{stat-dim}(\mathcal{K}) := \mathbb{E}[\|\mathcal{P}_{\mathcal{K}}(\mathbf{g})\|_2^2]$$

and

$$w(\mathcal{K}) := \mathbb{E} \left[ \sup_{\mathbf{z} \in \mathcal{K}, \|\mathbf{z}\|_2=1} \langle \mathbf{z}, \mathbf{g} \rangle \right],$$

where  $\mathbf{g} \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$  and  $\mathcal{P}_{\mathcal{K}}$  denotes the projection to  $\mathcal{K}$  as

$$\mathcal{P}_{\mathcal{K}}(\mathbf{g}) = \arg \min_{\mathbf{z} \in \mathcal{K}} \|\mathbf{g} - \mathbf{z}\|_2.$$

(a) Prove that  $w^2(\mathcal{K}) \leq \text{stat-dim}(\mathcal{K})$ .

(b) (Optional (10 bonus points)) Prove the reverse inequality  $\text{stat-dim}(\mathcal{K}) \leq w^2(\mathcal{K}) + 1$ .

*hint:* Let  $f(\cdot)$  be a function that is Lipschitz with respect to the Euclidean norm:

$$|f(\mathbf{u}) - f(\mathbf{v})| \leq M \|\mathbf{u} - \mathbf{v}\|_2 \quad \forall \mathbf{u}, \mathbf{v}.$$

Then,  $\text{Var}(f(\mathbf{g})) \leq M^2$ .