

# ST552, Homework 1

Due Wednesday, Sep 4, 2013

Please make your proofs self-contained, without citing any corresponding theorem in the textbook or lecture notes

1. Show that for an arbitrary matrix  $\mathbf{A}$ , the maximum number of linearly independent rows equals the maximum number of linearly independent columns. Therefore the rank can be defined either way.
2. Assume  $\mathbf{A} \in \mathbb{R}^{m \times n}$  and  $\mathbf{B} \in \mathbb{R}^{n \times p}$ . Show the following facts about the effect of matrix multiplication on the rank.
  - (a)  $\text{rank}(\mathbf{AB}) \leq \min\{\text{rank}(\mathbf{A}), \text{rank}(\mathbf{B})\}$  for any  $\mathbf{B}$ .
  - (b)  $\text{rank}(\mathbf{AB}) = \text{rank}(\mathbf{B})$  for any  $\mathbf{A}$  of full column rank.
  - (c)  $\text{rank}(\mathbf{AB}) = \text{rank}(\mathbf{A})$  for any  $\mathbf{B}$  of full row rank.
  - (d)  $\text{rank}(\mathbf{AA}^T) = \text{rank}(\mathbf{A}^T \mathbf{A}) = \text{rank}(\mathbf{A})$ . This is same as 3(c) below.
  - (e)  $\text{rank}(\mathbf{AA}^-) = \text{rank}(\mathbf{A}^- \mathbf{A}) = \text{rank}(\mathbf{A})$ .
3. Show the following facts about the *Gramian* matrix  $\mathbf{A}^T \mathbf{A}$ .
  - (a)  $\mathbf{A}^T \mathbf{A}$  is symmetric and positive semidefinite.
  - (b)  $\mathcal{C}(\mathbf{A}) = \mathcal{C}(\mathbf{AA}^T)$ .
  - (c)  $\text{rank}(\mathbf{A}) = \text{rank}(\mathbf{A}^T) = \text{rank}(\mathbf{A}^T \mathbf{A}) = \text{rank}(\mathbf{AA}^T)$ .
  - (d)  $\mathbf{A}^T \mathbf{A} = \mathbf{0}$  if and only if  $\mathbf{A} = \mathbf{0}$ .
  - (e)  $\mathbf{BA}^T \mathbf{A} = \mathbf{CA}^T \mathbf{A}$  if and only if  $\mathbf{BA}^T = \mathbf{CA}^T$ .
  - (f)  $\mathbf{A}^T \mathbf{AB} = \mathbf{A}^T \mathbf{AC}$  if and only if  $\mathbf{AB} = \mathbf{AC}$ .
  - (g) For any generalized inverse  $(\mathbf{A}^T \mathbf{A})^-$ ,  $[(\mathbf{A}^T \mathbf{A})^-]^T$  is also a generalized inverse of  $\mathbf{A}^T \mathbf{A}$ . Note  $(\mathbf{A}^T \mathbf{A})^-$  is not necessarily symmetric.
  - (h)  $(\mathbf{A}^T \mathbf{A})^- \mathbf{A}^T$  is a generalized inverse of  $\mathbf{A}$ .
  - (i)  $\mathbf{AA}^+ = \mathbf{A}(\mathbf{A}^T \mathbf{A})^- \mathbf{A}^T$ , where  $\mathbf{A}^+$  is the Moore-Penrose inverse of  $\mathbf{A}$ .
  - (j) Let  $\mathbf{P}_A = \mathbf{A}(\mathbf{A}^T \mathbf{A})^- \mathbf{A}^T$ . Show that  $\mathbf{P}_A$  is symmetric, idempotent, invariant to the choice of generalized inverse  $(\mathbf{A}^T \mathbf{A})^-$ , and projects onto  $\mathcal{C}(\mathbf{A})$ .
4. (a) Show the Sherman-Morrison formula

$$(\mathbf{A} + \mathbf{u}\mathbf{u}^T)^{-1} = \mathbf{A}^{-1} - \frac{1}{1 + \mathbf{u}^T \mathbf{A}^{-1} \mathbf{u}} \mathbf{A}^{-1} \mathbf{u} \mathbf{u}^T \mathbf{A}^{-1},$$

where  $\mathbf{A} \in \mathbb{R}^{n \times n}$  is a nonsingular matrix and  $\mathbf{u} \in \mathbb{R}^n$ . This formula supplies the inverse of the symmetric, rank-one perturbation of  $\mathbf{A}$ .

(b) Show the Woodbury formula

$$(\mathbf{A} + \mathbf{U}\mathbf{V}^\top)^{-1} = \mathbf{A}^{-1} - \mathbf{A}^{-1}\mathbf{U}(\mathbf{I}_m + \mathbf{V}^\top\mathbf{A}^{-1}\mathbf{U})^{-1}\mathbf{V}^\top\mathbf{A}^{-1},$$

where  $\mathbf{A} \in \mathbb{R}^{n \times n}$  is nonsingular,  $\mathbf{U}, \mathbf{V} \in \mathbb{R}^{n \times m}$ , and  $\mathbf{I}_m$  is the  $m \times m$  identity matrix. In many applications  $m$  is much smaller than  $n$ . Woodbury formula generalizes Sherman-Morrison and is valuable because the smaller matrix  $\mathbf{I}_m + \mathbf{V}^\top\mathbf{A}^{-1}\mathbf{U}$  is typically much easier to invert than the larger matrix  $\mathbf{A} + \mathbf{U}\mathbf{V}^\top$ .

(c) Show the binomial inversion formula

$$(\mathbf{A} + \mathbf{U}\mathbf{B}\mathbf{V}^\top)^{-1} = \mathbf{A}^{-1} - \mathbf{A}^{-1}\mathbf{U}\mathbf{B}^{-1}(\mathbf{B}^{-1} + \mathbf{V}^\top\mathbf{A}^{-1}\mathbf{U})^{-1}\mathbf{B}\mathbf{V}^\top\mathbf{A}^{-1},$$

where  $\mathbf{A}$  and  $\mathbf{B}$  are nonsingular.

(d) Show the identity

$$\det(\mathbf{A} + \mathbf{U}\mathbf{V}^\top) = \det(\mathbf{A})\det(\mathbf{I}_m + \mathbf{V}^\top\mathbf{A}^{-1}\mathbf{U}).$$

This formula is useful for evaluating the density of a multivariate normal with covariance matrix  $\mathbf{A} + \mathbf{U}\mathbf{U}^\top$ .