

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 $\mathbf{P}_\mathbf{A} = \mathbf{A}(\mathbf{A}^T\mathbf{A})^- \mathbf{A}^T$. Show that $\mathbf{P}_\mathbf{A}$ 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

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

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 $\mathbf{A} + \mathbf{U}\mathbf{U}^\top$.