

LINEAR MODELS AND GLM: ASSIGNMENT 1

Exercise 1. (2 points) Linear models are called “linear” because they are linear in the parameters β , or more technically, the expected value $E(\mathbf{y} \mid \beta)$ as a function of β is linear in β . Make this statement more precise and prove it.

A model $\{P_{\theta}, \theta \in \Theta\}$ is said to be *identifiable* if for any $\theta_1, \theta_2 \in \Theta, \theta_1 \neq \theta_2 \implies P_{\theta_1} \neq P_{\theta_2}$. We have already seen that the one-way model is not identifiable.

Exercise 2. (2 points) Consider the model given by $\mathcal{N}(\theta, 1), \theta \in \mathbb{R}$. Show that the model is identifiable; that is, if $\theta_1 \neq \theta_2$, then the distributions defined by them are different.

SOME BASIC LINEAR ALGEBRA CONCEPTS REVIEWED

A set of vectors $\mathbf{x}_1, \dots, \mathbf{x}_n$ is *linearly dependent* if there exist coefficients c_1, \dots, c_n , not all zero, such that $\sum_i c_i \mathbf{x}_i = \mathbf{0}$. This can be rephrased in matrix notation as follows: the columns of a matrix \mathbf{X} are said to be *linearly dependent* if there exists $\mathbf{c} \neq \mathbf{0}$ such that $\mathbf{X}\mathbf{c} = \mathbf{0}$.

A set of vectors $\mathbf{x}_1, \dots, \mathbf{x}_n$ (the columns of matrix \mathbf{X}) is *linearly independent* if

$$\sum_i c_i \mathbf{x}_i = \mathbf{X}\mathbf{c} = \mathbf{0} \implies \mathbf{c} = \mathbf{0}$$

In \mathbb{R}^d , no more than d vectors can be linearly independent.

Two vectors are *orthogonal* if their inner product is zero, i.e.,

$$\mathbf{x}'\mathbf{y} = \mathbf{y}'\mathbf{x} = \sum_i x_i y_i = 0$$

A set of vectors is *mutually orthogonal* if they are pairwise orthogonal. If the columns of a matrix \mathbf{Q} are orthogonal, then $\mathbf{Q}'\mathbf{Q}$ is a diagonal matrix. In addition, if the columns of \mathbf{Q} are of unit length ($\|\mathbf{x}\| = \sqrt{\mathbf{x}'\mathbf{y}} = 1$), then $\mathbf{Q}'\mathbf{Q} = \mathbf{I}$, and \mathbf{Q} is said to be *orthonormal*.

Exercise 3. (2 points) Show that a set of mutually orthogonal vectors is linearly independent.

A *vector space* is a set of vectors closed under addition and scalar multiplication.

The *span* of a set of vectors $\mathbf{x}_1, \dots, \mathbf{x}_n$ is the set of all linear combinations

$$\text{span}\{\mathbf{x}_1, \dots, \mathbf{x}_n\} = \left\{ \mathbf{x} \mid \mathbf{x} = \sum_i c_i \mathbf{x}_i \text{ for some } c_i, i = 1, \dots, n \right\}$$

$\text{span}\{\mathbf{x}_1, \dots, \mathbf{x}_n\}$ is a vector space, and is said to be generated by $\mathbf{x}_1, \dots, \mathbf{x}_n$.

A *basis* for a vector space \mathcal{V} is a set of linearly independent vectors $\mathbf{x}_1, \dots, \mathbf{x}_n$ that generate \mathcal{V} (i.e., $\mathcal{V} = \text{span}\{\mathbf{x}_1, \dots, \mathbf{x}_n\}$).

The *dimension* of a vector space is the number of vectors in a basis for the vector space.

Dimension is unique, although basis is not. The set of elementary vectors $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ is an orthonormal basis for \mathbb{R}^n , where $\mathbf{e}_i \in \mathbb{R}^n$ has i^{th} element 1 and all other elements 0.

The *rank* of a matrix \mathbf{X} , denoted by $\text{rank}(\mathbf{X})$, is the number of linearly independent columns (or rows). \mathbf{X} has full column rank if rank equals the number of columns (similarly full row rank).

The *column space* of a matrix \mathbf{X} , denoted by $\mathcal{C}(\mathbf{X})$, is the span of the columns of \mathbf{X} ; that is, $\mathcal{C}(\mathbf{X}) = \{\mathbf{z} \mid \mathbf{z} = \mathbf{X}\mathbf{y} \text{ for some } \mathbf{y}\}$.

$\dim(\mathcal{C}(\mathbf{X})) = \text{rank}(\mathbf{X})$, the number of linearly independent columns.

Exercise 4. (2 points) Show that $\text{rank}(\mathbf{AB}) \leq \min(\text{rank}(\mathbf{A}), \text{rank}(\mathbf{B}))$

Exercise 5. (3 points) Show that $\mathcal{C}(\mathbf{A}) \subseteq \mathcal{C}(\mathbf{B}) \iff \mathbf{A} = \mathbf{BC}$ for some \mathbf{C} .

The *null space* or *kernel* of a matrix \mathbf{A} is $\mathcal{N}(\mathbf{A}) = \{\mathbf{x} \mid \mathbf{Ax} = \mathbf{0}\}$

Exercise 6. (0 points) Show that if \mathbf{A} has full column rank, then $\mathcal{N}(\mathbf{A}) = \{\mathbf{0}\}$.

Vector spaces $\mathcal{U}, \mathcal{V} \subseteq \mathbb{R}^n$ are said to form *orthogonal complements* if (1) $\mathcal{U} \cap \mathcal{V} = \{\mathbf{0}\}$, (2) $\dim(\mathcal{U}) + \dim(\mathcal{V}) = n$, and (3) $\mathbf{u} \perp \mathbf{v}$ for all $\mathbf{u} \in \mathcal{U}, \mathbf{v} \in \mathcal{V}$. Then, we write $\mathcal{U} \oplus^\perp \mathcal{V} = \mathbb{R}^n$.

Exercise 7. (4 points) Show that if $\mathcal{U} \oplus^\perp \mathcal{V} = \mathbb{R}^n$, then any $\mathbf{x} \in \mathbb{R}^n$ can be written as $\mathbf{x} = \mathbf{u} + \mathbf{v}$ where $\mathbf{u} \in \mathcal{U}, \mathbf{v} \in \mathcal{V}$, and the decomposition is unique. Note that for such a decomposition, $\|\mathbf{x}\|^2 = \|\mathbf{u}\|^2 + \|\mathbf{v}\|^2$.

Exercise 8. (6 points) Prove that for any matrix $\mathbf{A}_{m \times n}$, $\dim(\mathcal{N}(\mathbf{A})) + \dim(\mathcal{C}(\mathbf{A})) = n$.

Exercise 9. (3 points) Show that for any matrix $\mathbf{A}_{m \times n}$, $\mathcal{C}(\mathbf{A})$ and $\mathcal{N}(\mathbf{A}')$ are orthogonal complements in \mathbb{R}^m .

Exercise 10. (3 points) Prove the following results:

- \mathcal{U}, \mathcal{V} vector spaces, $\mathcal{U} \subseteq \mathcal{V}, \dim(\mathcal{U}) = \dim(\mathcal{V}) \implies \mathcal{U} = \mathcal{V}$.
- $\mathbf{Ax} + \mathbf{b} = \mathbf{0} \forall \mathbf{x} \in \mathbb{R}^n \implies \mathbf{A} = \mathbf{0}, \mathbf{b} = \mathbf{0}$.
- $\mathbf{Ax} = \mathbf{Bx} \forall \mathbf{x} \implies \mathbf{A} = \mathbf{B}$.
- If \mathbf{A} has full column rank, then $\mathbf{AB} = \mathbf{AC} \implies \mathbf{B} = \mathbf{C}$.

Exercise 11. (3 points) Show that $\mathcal{N}(\mathbf{X}'\mathbf{X}) = \mathcal{N}(\mathbf{X})$, and as a corollary, that $\mathcal{C}(\mathbf{X}'\mathbf{X}) = \mathcal{C}(\mathbf{X}')$.

Exercise 12. (3 points) The goal of this exercise is to prove the existence of a *g-inverse* for an arbitrary matrix \mathbf{A} by constructing one. Let \mathbf{B} and \mathbf{C} be nonsingular matrices such that $\mathbf{BAC} = \mathbf{\Delta}$ is a diagonal matrix (not necessarily square). Find a candidate $\mathbf{\Delta}^-$. Prove that $\mathbf{C}\mathbf{\Delta}^-\mathbf{B}$ is a *g-inverse* of \mathbf{A} .