

# Statistical modelling

## #2.c Geometry of least squares

**Dr. Léo Belzile**  
**HEC Montréal**

# Linear algebra reminders

For an  $n \times p$  matrix, the column space of  $\mathbf{X}$  is

$$S(\mathbf{X}) = \{\mathbf{X}\mathbf{a}, \mathbf{a} \in \mathbb{R}^p\}$$

The linear model equation

$$Y = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$$

corresponds to an (unknown) element of the span of  $\mathbf{X}$  plus a disturbance.

# Ordinary least squares

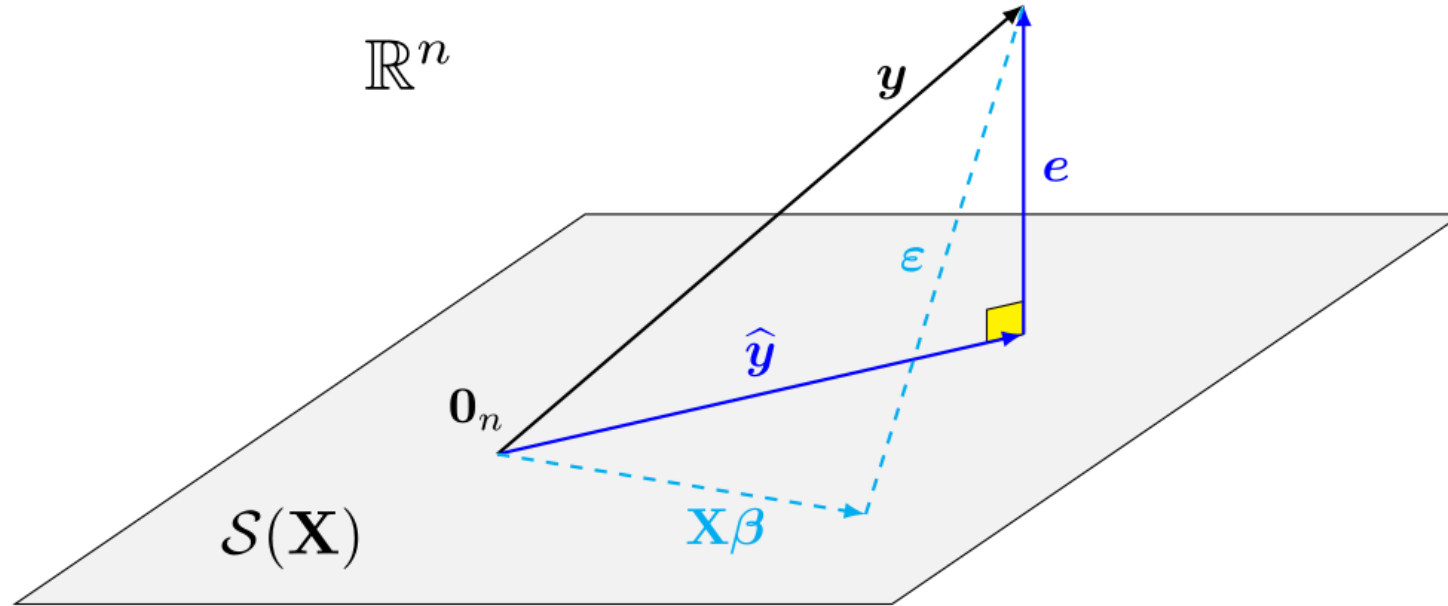
Find the element of  $S(\mathbf{X})$  with the minimum distance from  $\mathbf{y}$  (ordinary least squares), i.e.

$$\hat{\boldsymbol{\beta}} = \min_{\boldsymbol{\beta} \in \mathbb{R}^p} \|\mathbf{y} - \mathbf{X}\boldsymbol{\beta}\|^2$$

Intuition:  $\varepsilon_1, \dots, \varepsilon_n$  and  $\beta_0, \dots, \beta_{p-1}$  are unknown, but we cannot retrieve them ( $n$  observations,  $n + p$  unknowns).

# Column geometry

We try to find the best  $p$ -dimensional approximation onto  $s(\mathbf{X})$ .



The solution to the least square problem is the projection of  $\mathbf{y}$  onto  $\mathcal{S}(\mathbf{X})$ , i.e.,  $\mathbf{H}\mathbf{y}$ , where  $\mathbf{H} = \mathbf{X}(\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{X}^\top$ .

# Orthogonal decomposition

Write

$$\begin{aligned}\mathbf{y} &= \mathbf{H}\mathbf{y} + (\mathbf{I}_n - \mathbf{H})\mathbf{y} \\ &= \mathbf{X}\hat{\boldsymbol{\beta}} + \mathbf{e}\end{aligned}$$

The residuals  $\mathbf{e}$  are orthogonal to the columns of  $\mathbf{X}$  and the fitted values  $\hat{\mathbf{y}} = \mathbf{X}\hat{\boldsymbol{\beta}}$ .

✦ By Pythagoras' theorem,  $\|\mathbf{y}\|^2 = \|\hat{\mathbf{y}}\|^2 + \|\mathbf{e}\|^2$ .

Assuming  $\mathbf{1}_n \in \mathcal{S}(\mathbf{X})$  (intercept included)

✦ The sample mean of  $\mathbf{e}$  is zero.

✦ A linear regression of  $\hat{\mathbf{y}}$  onto  $\mathbf{e}$  has zero intercept and slope (they are uncorrelated).