

QR Decomposition

- The columns of Q consists of an orthonormal basis for the column space of X .
- Q is orthogonal matrix of $n \times d$, satisfying $Q^T Q = I$.
- R is upper triangular matrix of $d \times d$, full-ranked.
- $X^T X = (QR)^T (QR) = R^T Q^T QR = R^T R$

The least square solutions are

$$\begin{aligned}\hat{\beta} &= (X^T X)^{-1} X^T y \\ &= R^{-1} R^{-T} R^T Q^T y = R^{-1} Q^T y \\ \hat{y} &= X \hat{\beta} \\ &= (QR)(R^{-1} Q^T y) \\ &= QQ^T y.\end{aligned}$$

QR Algorithm for Normal Equations

Regard $\hat{\beta}$ as the solution for linear equations system:

$$R\beta = Q^T y.$$

- 1 Conduct QR decomposition of $X = QR$. (Gram-Schmidt Orthogonalization)
- 2 Compute $Q^T y$.
- 3 Solve the triangular system $R\beta = Q^T y$.

The computational complexity: nd^2

Cholesky Decomposition Algorithm

For any positive definite square matrix A , we have

$$A = RR^T,$$

where R is a lower triangular matrix of full rank.

- 1 Compute $X^T X$ and $X^T \mathbf{y}$.
- 2 Factoring $X^T X = RR^T$, then $\hat{\beta} = (R^T)^{-1} R^{-1} X^T \mathbf{y}$
- 3 Solve the triangular system $R\mathbf{w} = X^T \mathbf{y}$ for \mathbf{w} .
- 4 Solve the triangular system $R^T \beta = \mathbf{w}$ for β .

The computational complexity: $d^3 + nd^2/2$ (can be faster than QR for small d , but can be less stable)

$$\text{Var}(\hat{\mathbf{y}}_0) = \text{Var}(\mathbf{x}_0^T \hat{\beta}) = \sigma^2 (\mathbf{x}_0^T (R^T)^{-1} R^{-1} \mathbf{x}_0).$$