

Orthonormal Matrices

A square matrix Q is orthonormal if $Q'Q = I$. Let \mathbf{v}_j be the j th column of Q , then, the definition implies that $\mathbf{v}_i'\mathbf{v}_i = 1, i = 1, \dots, n$ and $\mathbf{v}_i'\mathbf{v}_j = 0, i \neq j$, that is that the columns of Q are all orthogonal to each other, i.e. the inner product of any two columns is 0.

Say we have data (X, y) . Rotating the data by Q does not affect our least squares estimator $\hat{\beta}$. Note that

$$\hat{\beta} = (X'X)^{-1}X'y = (X'Q'QX)^{-1}X'Q'Qy = (\tilde{X}'\tilde{X})^{-1}\tilde{X}'\tilde{y},$$

where $\tilde{X} = QX$ and $\tilde{y} = Qy$. More is true. Note that

$$Q'Q = I \implies Q'QQ^{-1} = Q^{-1} \implies Q' = Q^{-1}$$

thus

$$\hat{\beta} = (X'X)^{-1}X'y = (X'QQ^{-1}X)^{-1}X'QQ^{-1}y = (X'QQ'X)^{-1}X'QQ'y = (\hat{X}'\hat{X})^{-1}\hat{X}'\hat{y},$$

Let's see a practical example now.

```
data <- cars
Q <- randortho(length(cars$speed))
data$speedQ <- Q %*% data$speed
data$distQ <- Q %*% data$dist
data$speedQtr <- t(Q) %*% data$speed
data$distQtr <- t(Q) %*% data$dist
```

Regression with raw data.

```
lm(dist ~ speed - 1, data = cars)
```

```
##
## Call:
## lm(formula = dist ~ speed - 1, data = cars)
##
## Coefficients:
## speed
## 2.909
```

Regression with data pre-multiplied by Q .

```
lm(distQ ~ speedQ - 1, data = data)
```

```
##
## Call:
## lm(formula = distQ ~ speedQ - 1, data = data)
##
## Coefficients:
## speedQ
## 2.909
```

Regression with data pre-multiplied by Q' .

```
lm(distQtr ~ speedQtr - 1 , data = data)
```

```
##  
## Call:  
## lm(formula = distQtr ~ speedQtr - 1, data = data)  
##  
## Coefficients:  
## speedQtr  
##      2.909
```