

$$(y|\beta) \sim N(X\beta, \Sigma) \quad \Sigma = \sigma^2 I, \text{diag}(\sigma_1^2, \dots, \sigma_N^2) \quad \beta \sim N(m, V)$$

$$\begin{pmatrix} y \\ \beta \end{pmatrix} \sim N \left( \begin{pmatrix} X_m \\ m \end{pmatrix}, \begin{bmatrix} \Sigma + XVX^T & XV \\ VX^T & V \end{bmatrix} \right)$$

$$(\beta|y) \sim N(\hat{\beta}, \hat{\Sigma})$$

$$\hat{\Sigma} = V - VX^T(XVX^T + \Sigma)^{-1}XV \quad \star$$

$$\hat{\beta} = m + VX^T[XVX^T + \Sigma]^{-1}(y - X_m)$$

Woodbury Matrix Identity

$$\star (A + LTR)^{-1} = A^{-1} - A^{-1}L(T^{-1} + RA^{-1}L)^{-1}RA^{-1}$$

$$\hat{\Sigma} = V - \underbrace{VX^T(XVX^T + \Sigma)^{-1}XV}_{n \times n \text{ matrix}} = \underbrace{(V^{-1} + X^T \Sigma^{-1} X)^{-1}}_{p \times p \text{ matrix}} \quad \text{easier to invert a } p \times p \text{ matrix}$$

$$\hat{\beta} = m + VX^T[XVX^T + \Sigma]^{-1}(y - X_m) = \hat{\Sigma} [\underbrace{V^{-1}m}_{\text{prior mean}} + \underbrace{X^T \Sigma^{-1} y}_{\text{data}}]$$

$$\hat{\Sigma} \left[ \frac{1}{\tau^2} I_m + \frac{1}{\sigma^2} X^T X \right] \quad V = \tau^2 I$$

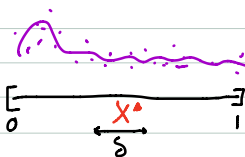
$$\hat{\Sigma} = \left( \frac{1}{\tau^2} I + X^T \frac{1}{\sigma^2} I X \right)^{-1} \quad \Sigma = \sigma^2 I$$

$$\begin{aligned} E(\beta|y) &= \left( \frac{1}{\tau^2} I + \frac{1}{\sigma^2} X^T X \right)^{-1} \left( \frac{1}{\tau^2} m + \frac{1}{\sigma^2} X^T y \right) \\ &= \frac{1}{\sigma^2} \left( \frac{1}{\tau^2} I + \frac{1}{\sigma^2} X^T X \right)^{-1} \left( \frac{\sigma^2}{\tau^2} m + X^T y \right) \\ &= \left( \frac{\sigma^2}{\tau^2} I + X^T X \right)^{-1} \left( \frac{\sigma^2}{\tau^2} m + X^T y \right) \\ &\quad \downarrow \lambda \end{aligned}$$

$$\begin{aligned} \text{Ridge Regression} &= (\lambda I + X^T X)^{-1} (\lambda m + X^T y) \quad \lambda = \frac{\sigma^2}{\tau^2} = \frac{1/\sigma^2}{1/\tau^2} \quad \text{Signal-to-noise ratio} \\ \text{Estimation /} & \\ \text{Tychonov regularization} & \end{aligned}$$

Folk Theorem for Multivariate Stats

- To maintain given MSE,  $n$  must grow exponentially w/ dimension



$X = [0, 1]$  What # of points fall w/  $\delta/2$  of  $x^*$ ?  $n \delta$



$$X = [0,1] \times [0,1]$$

$$f(x_1, x_2)$$

$$x^* = (x_1^*, x_2^*)$$

$$f(x^*)$$

$$\text{w/in } \frac{\delta}{2} \text{ of } x^*$$

$$n \delta^2$$

$$n \delta^2$$

$$n \delta^{10}$$

$$n \delta^{1000}$$

$$MSE \approx \left[ \frac{c}{n^{1/p}} \right] \quad p = \text{dimension}$$

$$n \propto \left[ \frac{c}{MSE} \right]^{p/4}$$

data needs to grow exponentially