

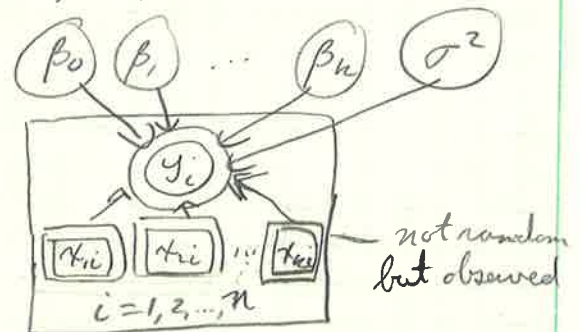
$$y_i = \underbrace{\beta_0 + \beta_1 x_{1i} + \dots + \beta_n x_{ni}}_{\text{mean}} + \underbrace{\varepsilon_i}_{\text{error}}, \quad \varepsilon_i \stackrel{\text{i.i.d.}}{\sim} N(0, \sigma^2), i=1, \dots, n$$

$$y_i | x_i, \beta, \sigma^2 \stackrel{\text{i.i.d.}}{\sim} N(\beta_0 + \beta_1 x_{1i} + \dots + \beta_n x_{ni}, \sigma^2)$$

$$\beta_0 \sim p(\beta_0), \quad \beta_i \sim p(\beta_i), \dots$$

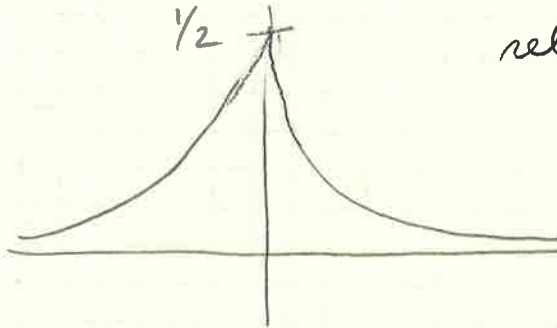
$$\sigma^2 \sim p(\sigma^2) \text{ (maybe IG)}$$

usually normal or multivariate normal



Laplace prior

$$p(\beta) = \frac{1}{2} e^{-|\beta|}$$



related to the LASSO

Factors: sound

Levels: music.
no music

font size

small
medium
large

$g_i \sim$ group of subject i
 $\underline{\mu} \sim$ vector of G group means

$$y_i | g_i, \underline{\mu}, \sigma^2 \stackrel{\text{ind}}{\sim} N(\mu_{g_i}, \sigma^2)$$

$$g_i \in \{1, \dots, G\}, i = 1, \dots, n$$

Alternative

$$E(y_i) = \beta_0 + \beta_1 \underset{\substack{\uparrow \\ I(g_i=1)}}{X_{1,i}} + \dots + \beta_{G-1} \underset{\substack{\uparrow \\ I(g_i=G-1)}}{X_{G-1,i}}$$

X 's are indicators or dummy variables

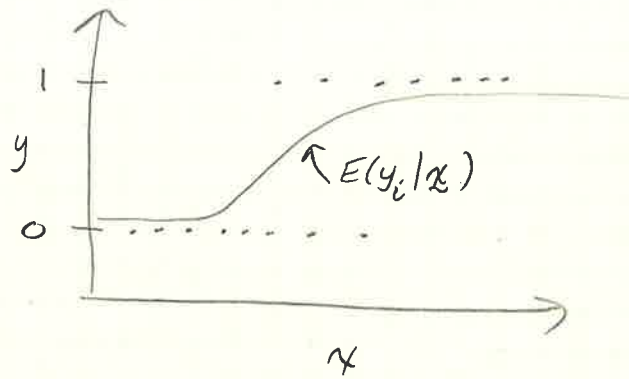
		B		
		1	2	3
A	1	$\mu_{1,1}$	$\mu_{1,2}$	$\mu_{1,3}$
	2	$\mu_{2,1}$	$\mu_{2,2}$	$\mu_{2,3}$

$A, B \sim$ Factors

6 treatment groups

Additive model: (good for no interactions between factors)

$$E(y_i) = \mu + \alpha_2 I(a_i = 2) + \beta_2 I(b_i = 2) + \beta_3 I(b_i = 3)$$



$y_i | \phi_i \overset{\text{ind}}{\sim} \text{Bernoulli}(\phi_i)$, $i=1, \dots, n$
 In linear regression: $E(y_i) = \beta_0 + \beta_1 x_{1i}$

ϕ is probability of success

$$\frac{\phi}{1-\phi} \Leftrightarrow \text{odds}$$

$$\log\left(\frac{\phi}{1-\phi}\right) \Leftrightarrow \text{logit link}$$

$$\text{logit}(\phi_i) = \log\left(\frac{\phi_i}{1-\phi_i}\right) = \beta_0 + \beta_1 x_{1i}$$

$$E(y_i) = \phi_i = \frac{\exp(\beta_0 + \beta_1 x_{1i})}{1 + \exp(\beta_0 + \beta_1 x_{1i})} = \frac{1}{1 + e^{-(\beta_0 + \beta_1 x_{1i})}}$$

$$y_i | x_i, \beta \overset{\text{ind}}{\sim} \text{Bernoulli}\left(\frac{1}{1 + \exp[-\beta_0 + \beta_1 x_{1i} + \dots + \beta_k x_{ki}]}\right)$$

$$y_i | \lambda_i \stackrel{\text{ind}}{\sim} \text{Pois}(\lambda_i), \quad i=1, \dots, n$$

linear regression: $E(y_i) = \beta_0 + \beta_1 x_{1i}$ is a problem

$$\text{log link: } \log(\lambda_i) = \beta_0 + \beta_1 x_{1i}$$

$$\Rightarrow E(y_i) = \lambda_i = e^{\beta_0 + \beta_1 x_{1i}}$$

λ_i is the mean
and variance
of the poisson distribution

$$E(\log(y)) \neq \log(E(y))$$

$$y_i | x_i, \beta \stackrel{\text{ind}}{\sim} \text{Pois}(\exp[\beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \dots + \beta_k x_{ki}])$$