$$f(x_1, \dots, x_n) = \prod_{i=1}^n f(x_i) \implies f(x_{n+1}, \dots, x_{n+m} | x_1, \dots, x_n) = \prod_{i=n+1}^m f(x_i)$$

$$\theta \sim f(\theta)$$

 Y_1 1 n



$$f(y_1,\ldots,y_n)$$

$$= \int f(y_1, \dots, y_n | \theta) dF(\theta)$$

$$= \int \prod_{i=1}^{n} f(y_i|\theta) dF(\theta)$$

$$= f\left(y_{\pi_1}, \dots, y_{\pi_n}\right)$$

$$p(y_1,\ldots,y_n)$$

$$= \int \prod_{i=1}^{n} p(y_i|\theta) dF(\theta)$$

$$F(\theta) = \lim_{n \to \infty} P(s_n \le \theta)$$

$$p(y_1 + \ldots + y_n = s_n) = \binom{n}{s_n} p(y_1, \ldots, y_n) = \binom{n}{s_n} p(y_{\pi_1}, \ldots, y_{\pi_n})$$

$$p(y_1 + \ldots + y_n = s_n) = \sum p(y_1 + \ldots + y_n = s_n | y_1 + \ldots + y_N = s_N) p(y_1 + \ldots + y_N = s_N)$$

$$B_n \sim Bin\left(n, \frac{b_0}{b_0 + w_0}\right)$$

 $\sim P(B_0 -$

 $= b_0$

$$P\left(B_1 = b_0 + 1\right)$$

$$=\frac{b_0}{b_0+w_0}$$

$$P\left(B_2 = b_0\right)$$

$$= \frac{w_0}{w_0 + b_0} \cdot \frac{w_0 + 1}{w_0 + b_0 + 1}$$

$$P\left(B_2 = b_0 + 1\right)$$

$$= \frac{w_0}{w_0 + b_0} \frac{b_0}{w_0 + b_0 + 1} + \frac{b_0}{w_0 + b_0} \frac{w_0}{w_0 + b_0 + 1}$$

$$P\left(B_2 = b_0 + 2\right)$$

$$= \frac{b_0}{w_0 + b_0} \cdot \frac{b_0 + 1}{w_0 + b_0 + 1}$$

 X_1,\ldots,X_n

$$Y_n = \frac{B_n}{B_n + W_n} = \frac{B_n}{b_0 + w_0 + n}$$

$$B_{n+1} = \begin{cases} B_n & \text{con probabilidad } 1 - Y_n \\ B_n + 1 & \text{con probabilidad } Y_n \end{cases}$$

$$E\left(Y_{n+1}|X_1,\ldots,X_n\right)$$

$$= E\left(\frac{B_{n+1}}{b_0 + w_0 + n} | X_1, \dots, X_n\right)$$

$$= \frac{1}{b_0 + w_0 + n} E(B_{n+1}|X_1, \dots, X_n)$$

$$= \frac{1}{b_0 + w_0 + n} \left(B_n \left(1 - Y_n \right) + \left(B_n + 1 \right) Y_n \right)$$

$$= \frac{1}{b_0 + w_0 + n} \left(B_n - B_n Y_n + B_n Y_n + Y_n \right)$$

$$= \frac{1}{b_0 + w_0 + n} \left(B_n + Y_n \right)$$

$$p(y = k) = \int_0^1 p(y = k|\theta) d\theta$$

 $k=1,\ldots,n$

$$p\left(y=k\right)$$

$$\int_{0}^{1} p(y=k,\theta) d\theta$$

$$= \int_{0}^{1} p(y=k|\theta) \underbrace{p(\theta)}_{1} d\theta$$

$$\int_0^1 p\left(y = k|\theta\right) d\theta$$

$$= \int_0^1 \binom{n}{k} \theta^k (1-\theta)^{n-k} d\theta$$

$$= \binom{n}{k} \int_0^1 \theta^k (1 - \theta)^{n-k} d\theta$$

$$= \binom{n}{k} \frac{\Gamma(k+1)\Gamma(n-k+1)}{\Gamma(n+2)}$$

$$= \binom{n}{k} \frac{k! (n-k)!}{(n+1)!}$$

$$=\frac{1}{n+1}$$

$$\int_0^1 \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \theta^{\alpha-1} (1-\theta)^{\beta-1} d\theta$$

 $\implies \int^{1} \theta^{\alpha-1} \left(1-\theta\right)^{\beta-1} d\theta$

$$=\frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)}$$

 $\int_{-\infty}^{\infty} \theta^k \left(1 - \theta\right)^{n-k} d\theta$

$$=\frac{\Gamma\left(k+1\right)\Gamma\left(n-k+1\right)}{\Gamma\left(k+1+n-k+1\right)}=\frac{\Gamma\left(k+1\right)\Gamma\left(n-k+1\right)}{\Gamma\left(n+2\right)}$$

$$\propto p(y|\theta) p(\theta)$$

$$= \binom{n}{y} \theta^{y} (1 - \theta)^{n-y} \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha) \Gamma(\beta)} \theta^{\alpha - 1} (1 - \theta)^{\beta - 1}$$

$$\propto \theta^{\alpha+y-1} \left(1-\theta\right)^{\beta+n-y-1}$$

$$(\alpha + y, \beta + n - y)$$

$$= \frac{\alpha + y}{\alpha + y + \beta + n - y}$$

$$= \frac{\alpha + y}{\alpha + \beta + n}$$

$$= \frac{\alpha}{\alpha + \beta + n} + \frac{y}{\alpha + \beta + n}$$

$$= \frac{\alpha + \beta}{\alpha + \beta} \frac{\alpha}{\alpha + \beta + n} + \frac{n}{n} \frac{y}{\alpha + \beta + n}$$

$$= \frac{\alpha + \beta}{\alpha + \beta + n} \underbrace{\frac{\alpha}{\alpha + \beta}}_{E(\theta)} + \frac{n}{\alpha + \beta + n} \frac{y}{n}$$

$$=\frac{\left(\alpha+y\right)\left(\beta+n-y\right)}{\left(\alpha+y+\beta+n-y\right)^{2}\left(\alpha+y+\beta+n-y\right)}$$

$$=\frac{\left(\alpha+y\right)\left(\beta+n-y\right)}{\left(\alpha+\beta+n\right)^{2}\left(\alpha+\beta+n+1\right)}$$

$$= \frac{(1+y)(1+n-y)}{(2+n)^2(2+n+1)} = \frac{(1+y)(1+n-y)}{(2+n)^2(n+3)} \le \frac{1}{12} =$$

$$\frac{(\alpha+y)(\beta+n-y)}{(\alpha+\beta+n)^{2}(\alpha+\beta+n+1)} \le \frac{\alpha\beta}{(\alpha+\beta)^{2}(\alpha+\beta+1)}$$

$$p(\theta) \propto \theta^{-1} (1 - \theta)^{-1}$$

$$\phi = \log\left(\frac{\theta}{1-\theta}\right) = h\left(\theta\right)$$

$$=\underbrace{p\left(\phi\right)}_{\propto 1}\left|\frac{d\phi}{d\theta}\right|$$

$$\propto \left| \frac{d}{d\theta} \log \left(\frac{\theta}{1-\theta} \right) \right|$$

$$\frac{1}{\theta} + \frac{1}{1}$$

$$=\theta^{-1}\left(1-\theta\right)^{-1}$$

$$= \binom{n}{y} \theta^y (1-\theta)^{n-y} \theta^{-1} (1-\theta)^{-1}$$





$$\sim N\left(\theta, \sigma^2\right)$$

$$\sim N\left(\mu_0, \tau_0^2\right)$$

$$= \prod_{i=1}^{n} \left[\left(2\pi \sigma^2 \right)^{-1/2} \exp \left(-\frac{1}{2\sigma^2} \left(y_i - \theta \right)^2 \right) \right] \left(2\pi \tau_0^2 \right)^{-1/2} \exp \left(-\frac{1}{2\tau_0^2} \left(\theta - \mu_0 \right)^2 \right)$$

$$\propto \exp\left(\frac{1}{2\sigma^2}\sum_{i=1}^n (y_i - \theta)^2\right) \exp\left(\frac{1}{2\tau_0^2}(\theta - \mu_0)^2\right)$$

$$= \exp\left(-\frac{1}{2\sigma^2} \sum_{i=1}^{n} (y_i - \theta)^2 - \frac{1}{2\tau_0^2} (\theta - \mu_0)^2\right)$$

$$\propto \exp\left(-\frac{1}{2\sigma^2}\sum_{i=1}^n (y_i^2 - 2y_i\theta + \theta^2) - \frac{1}{2\tau_0^2} (\theta^2 - 2\theta\mu_0 + \mu^2)\right)$$

$$\propto \exp\left(-\frac{1}{2\sigma^2}\sum_{i=1}^n\left(\theta^2-2y_i\theta\right)-\frac{1}{2\tau_0^2}\left(\theta^2-2\theta\mu_0\right)\right)$$

$$\theta|y \sim N\left(\frac{\frac{1}{\sigma^2}\mu_0 + \frac{n}{\tau_0^2}\overline{y}}{\frac{1}{\sigma^2} + \frac{n}{\tau_0^2}}, \frac{1}{\frac{1}{\sigma^2} + \frac{n}{\tau_0^2}}\right)$$

$$\theta|y \sim N\left(\frac{\frac{1}{20^2}180 + \frac{n}{40^2}150}{\frac{1}{20^2} + \frac{n}{40^2}}, \frac{1}{\frac{1}{20^2} + \frac{n}{40^2}}\right)$$

$$\widetilde{y}|y \sim N\left(\widetilde{\mu}, \widetilde{\sigma}^2\right)$$

$$=E\left(\tilde{y}|y\right)$$

$$= E\left(E\left(\tilde{y}|\theta,y\right)|y\right)$$

$$= E\left(E\left(\tilde{y}|\theta\right)|y\right)$$

$$=E\left(\theta|y\right)$$

$$= \frac{\frac{1}{\sigma^2}\mu_0 + \frac{n}{\tau_0^2}\overline{y}}{\frac{1}{\sigma^2} + \frac{n}{\tau_0^2}}$$

$$= var\left(\tilde{y}|y\right)$$

$$= var \left(E\left(\tilde{y}|\theta \right)|y \right) + E\left(var\left(\tilde{y}|\theta \right)|y \right)$$

$$= var(\theta|y) + E(\sigma^2|y)$$

$$= \frac{1}{\frac{1}{\sigma^2} + \frac{n}{\tau_0^2}} + \sigma^2$$

$$\tilde{y}|y \sim N\left(\frac{\frac{1}{20^2}180 + \frac{n}{40^2}150}{\frac{1}{20^2} + \frac{n}{40^2}}, \frac{1}{\frac{1}{\sigma^2} + \frac{n}{\tau_0^2}} + \sigma^2\right)$$

 y_1 . $., y_5$.

$$p(y_i|\theta) \propto \frac{1}{1+(y_i-\theta)^2}$$

$$l(\theta|y) = \prod_{i=1}^{n} \frac{1}{1 + (y_i - \theta)^2}$$

$$p\left(\mu,\sigma^2\right) \propto \left(\sigma^2\right)^{-3/2}$$

$$p\left(\mu, \sigma^2 | \mathbf{x}\right)$$

$$\propto p\left(\mu, \sigma^2\right) p\left(\mathbf{x}|\mu, \sigma^2\right)$$

$$= (\sigma^{2})^{-3/2} \sigma^{-n/2} \exp \left(-\frac{1}{2\sigma^{2}} \sum_{i} (x_{i} - \mu)^{2}\right)$$

$$= (\sigma^2)^{-3/2} \sigma^{-n/2} \exp\left(-\frac{1}{2\sigma^2} \sum_{i} (x_i - \overline{x})^2 + n(\overline{x} - \mu)^2\right)$$

$$= (\sigma^2)^{-3/2} \sigma^{-n/2} \exp\left(-\frac{1}{2\sigma^2} \left(\sum (x_i - \overline{x})^2 + n(\overline{x} - \mu)^2\right)\right)$$

$$= (\sigma^{2})^{-3/2 - n/2} \exp\left(-\frac{1}{2\sigma^{2}} \left((n-1) s^{2} + n (\overline{x} - \mu)^{2} \right) \right)$$

$$\propto \exp\left(-\frac{1}{2\sigma^2}\left(n\left(\overline{x}-\mu\right)^2\right)\right)$$

$$\sim N\left(\overline{x}, \sigma^2/n\right)$$

$$\propto (\sigma^2)^{-3/2-n/2} \exp\left(-\frac{1}{2\sigma^2}\left((n-1)s^2+n(\overline{x}-\mu)^2\right)\right)$$

$$\frac{-3/2 - n/2}{\exp\left(-\frac{1}{2\sigma^2}\left(n - 1\right)s^2\right)} \underbrace{\exp\left(-\frac{1}{2\sigma^2}n\left(\overline{x} - \mu\right)^2\right)\left(\frac{\sigma}{n}\right)^{-1/2}}_{\text{kernel de una normal}} \left(\frac{\sigma}{n}\right)^{1/2}$$

$$\propto (\sigma^2)^{-(n/2+1)} \exp\left(-\frac{1}{2\sigma^2}(n-1)s^2\right)$$

$$\sim \chi^2 - \mathbf{inv}\left(n, s^2\right)$$

$$p(\mu|\mathbf{x})$$

$$= \int_0^\infty p\left(\mu, \sigma^2 | \mathbf{x}\right) d\sigma^2$$

 $\propto \int_{0}^{\infty} (\sigma^{2})^{-3/2} \sigma^{-n/2} \exp\left(-\frac{1}{2\sigma^{2}}A\right) d\sigma^{2}$

$$= \sum (x_i - \overline{x})^2 + n(\overline{x} - \mu)^2$$

$$z = A/\left(2\sigma^2\right)$$

4/

$$d\sigma^2 = -\frac{A}{2z^2}dz$$

$$\propto \int_0^\infty \left(\frac{A}{2z}\right)^{-3/2-n/2} \exp\left(-z\right) \left(-\frac{A}{2z^2}\right) dz$$

$$\propto \left(\frac{A}{2z}\right)^{-1/2-n/2}$$

S כו

$$\mu^{(s)} \sim p\left(\mu|\mathbf{y}\right)$$

$$\sigma^{2(s)} \sim p\left(\sigma^2|\mathbf{y}\right)$$

$$y^s \sim p\left(x|\mu^{(s)}, \sigma^{2(s)}\right)$$

$$p(\mu, \sigma^2) = p(\mu|\sigma^2) p(\sigma^2)$$

$$= p(\mu|\sigma^2) p(\sigma^2) p(\mathbf{x}|\mu,\sigma^2)$$

$$p\left(\beta,\sigma^2\right) \propto \sigma^{-2}$$

$$y|\beta,\sigma^2,X$$

$$\sim N\left(X\beta,\sigma^2I\right)$$

$$p(y|\beta,\sigma^2,X)$$

$$\propto \det (\sigma^2 I)^{-1/2} \exp \left(-\frac{1}{2\sigma^2} (y - X\beta)^t (y - X\beta)\right)$$

$$= \left(\sigma^2\right)^{-1/2} \exp\left(-\frac{1}{2\sigma^2} \left(y - X\beta\right)^t \left(y - X\beta\right)\right)$$

 $\implies p\left(\beta, \sigma^2 | y, X\right)$

 $\propto (\sigma^2)^{-n/2} \exp\left(-\frac{1}{2\sigma^2} (y - X\beta)^t (y - X\beta)\right) \sigma^{-2}$

$$= \left(\sigma^2\right)^{-n/2-2} \exp\left(-\frac{1}{2\sigma^2} \left(y - X\beta\right)^t \left(y - X\beta\right)\right)$$

$$\hat{\beta}, \sigma^2 \left(X^t X \right)^{-1}$$

$$p\left(\beta,\sigma^2|y\right)$$

$$= p\left(\sigma^2|\beta, y\right) p\left(\beta|y\right)$$

$$p(\beta|y) = \frac{p(\beta, \sigma^{2}|y)}{p(\sigma^{2}|\beta, y)} = \frac{p(y|\beta, \sigma^{2}) p(\beta, \sigma^{2})}{p(\sigma^{2}|\beta, y)}$$

$$= p\left(\beta|\sigma^2, y\right) p\left(\sigma^2|y\right)$$

$$p\left(\sigma^{2}|y\right) = \frac{p\left(y|\beta,\sigma^{2}\right)p\left(\beta,\sigma^{2}\right)}{p\left(\beta|\sigma^{2},y\right)}$$

$$p\left(\beta|\sigma^2,y\right)$$

$$= \frac{p\left(\beta, \sigma^2|, y\right)}{p\left(\sigma^2|y\right)}$$

$$\propto p\left(\beta, \sigma^2|, y\right)$$

$$\propto \exp\left(-\frac{1}{2\sigma^2}(y-X\beta)^t(y-X\beta)\right)$$

$$= \exp\left(-\frac{1}{2\sigma^2}\left(y - X\hat{\beta}\right)^t \left(y - X\hat{\beta}\right) - \frac{1}{2\sigma^2}\left(\beta - \hat{\beta}\right)^t X^t X \left(\beta - \hat{\beta}\right)\right)$$

$$\propto \exp\left(-\frac{1}{2\sigma^2}\left(\beta - \hat{\beta}\right)^t X^t X\left(\beta - \hat{\beta}\right)\right)$$

 $\beta | \sigma$ Ч

$$\sim N\left(\hat{\beta}, \sigma^2 \left(X^t X\right)^{-1}\right)$$

$$p\left(\sigma^2|y\right)$$

$$= \frac{p(\beta, \sigma^2|, y)}{p(\beta|\sigma^2, y)}$$

$$\frac{\left(\sigma^{2}\right)^{-n/2-2} \exp\left(-\frac{1}{2\sigma^{2}} \left(y - X\beta\right)^{t} \left(y - X\beta\right)\right)}{\det\left(\sigma^{2} \left(X^{t} X\right)^{-1}\right)^{-1/2} \exp\left(-\frac{1}{2\sigma^{2}} \left(\beta - \hat{\beta}\right)^{t} X^{t} X \left(\beta - \hat{\beta}\right)\right)}$$

$$=\frac{\left(\sigma^{2}\right)^{-n/2-2}\exp\left(-\frac{1}{2\sigma^{2}}\left(y-X\hat{\beta}\right)^{t}\left(y-X\hat{\beta}\right)-\frac{1}{2\sigma^{2}}\left(\beta-\hat{\beta}\right)^{t}X^{t}X\left(\beta-\hat{\beta}\right)\right)}{\det\left(\sigma^{2}\left(X^{t}X\right)^{-1}\right)^{-1/2}\exp\left(-\frac{1}{2\sigma^{2}}\left(\beta-\hat{\beta}\right)^{t}X^{t}X\left(\beta-\hat{\beta}\right)\right)}$$

$$=\frac{\left(\sigma^{2}\right)^{-n/2-2}\exp\left(-\frac{1}{2\sigma^{2}}\left(y-X\hat{\beta}\right)^{t}\left(y-X\hat{\beta}\right)\right)}{\det\left(\sigma^{2}\left(X^{t}X\right)^{-1}\right)^{-1/2}\exp\left(0\right)}$$

$$= \frac{\left(\sigma^{2}\right)^{-n/2-2} \exp\left(-\frac{1}{2\sigma^{2}}\left(y - X\hat{\beta}\right)^{t} \left(y - X\hat{\beta}\right)\right)}{\left(\sigma^{2k}\right)^{-1/2} \det\left(\left(X^{t}X\right)^{-1}\right)^{-1/2} \exp\left(0\right)}$$

$$\propto (\sigma^2)^{-n/2-2+k/2} \exp\left(-\frac{1}{2\sigma^2}\left(y - X\hat{\beta}\right)^t \left(y - X\hat{\beta}\right)\right)$$

$$\sim Inv - \chi^2 \left(n - k, s^2 \right)$$

$$\pi(x) = \Phi(\beta_0 + \beta_1 x | \mu, \sigma^2)$$

$$(\pi(x_i))$$

$$= \Phi \left(\beta_0 + \beta_1 x | \mu, \sigma^2 \right)$$

$$=\Phi_{0,1}\left(\frac{\beta_0+\beta_1x-\mu}{\sigma}\right)$$

$$= \Phi_{0,1} \left(\beta_0^* + \beta_1^* x \right)$$