

$$\hat{\boldsymbol{\nu}}(\boldsymbol{\theta}) = \operatorname{argmax}_{\boldsymbol{\nu}} L(\boldsymbol{\theta}, \boldsymbol{\nu})$$

$$\lambda(\boldsymbol{\theta}) = \frac{L(\boldsymbol{\theta}, \hat{\boldsymbol{\nu}}(\boldsymbol{\theta}))}{L(\hat{\boldsymbol{\theta}}, \hat{\boldsymbol{\nu}})}$$

$$q_{\boldsymbol{\theta}} = -2\ln \lambda(\boldsymbol{\theta})$$

$$\boldsymbol{\theta} = (\theta_1, \ldots, \theta_N)$$

$$\boldsymbol{\nu} = (\nu_1, \ldots, \nu_M)$$

$$Z=\Phi^{-1}(1-p)=3.6$$

$$P(n|s,b)=\frac{(s+b)^n}{n!}e^{-(s+b)}$$

$$m=\sqrt{2E_1E_2(1-\cos\theta_{12})}$$

$$P(n|s,b)=\frac{(s+b)^n}{n!}e^{-(s+b)}$$

$$Q=-2\ln\frac{L_{s+b}}{L_b}=2s-2\sum_{i=1}^n\ln\left[1+\frac{s}{b}\frac{f(\mathbf{x}_i|s)}{f(\mathbf{x}_i|b)}\right]$$

$$\lambda(\mu) = \frac{L(\mu)}{L(\hat{\mu})}$$

$$\hat{\mu} = \operatorname{argmax}_{\mu} L(\mu)$$

$$p=1-\Phi\left(\sqrt{q_0}\right)$$

$$Z=\sqrt{q_0}=\hat{\mu}/\sigma_{\hat{\mu}}$$

$$L(\mu) = \frac{(\mu s + b)^n}{n!} e^{-(\mu s + b)} \prod_{i=1}^n f(x_i|\mu)$$

$$f(x|\mu) = \frac{\mu s}{\mu s + b} f(x|s) + \frac{b}{\mu s + b} f(x|b)$$

$$x_1,\ldots,x_n$$

$$n \sim \text{Poisson}(\mu s + b)$$

$$y(\mathbf{x}) = \frac{p(\mathbf{x}|\mathbf{s})}{p(\mathbf{x}|\mathbf{b})}$$

$$\frac{p(\mathbf{x}|\mathbf{s})}{p(\mathbf{x}|\mathbf{b})} > c$$

$$t_{\text{p}}(x) = \frac{L_{\text{p}}(x|s)}{L_{\text{p}}(x|b)} = \frac{L(x|\hat{\nu}(s),s)}{L(x|\hat{\nu}(b),b)}$$

$$t_{\text{m}}(x) = \frac{L_{\text{m}}(x|s)}{L_{\text{m}}(x|b)} = \frac{\int L(x|\nu,s)\pi_{\nu}(\nu)\,d\nu}{\int L(x|\nu,b)\pi_{\nu}(\nu)\,d\nu}$$

$$P(\theta < \theta_{\text{up}}(x)|\theta) = ?$$

$$P(\theta < \theta_{\text{up}}|x) = \int_{-\infty}^{\theta_{\text{up}}} p(\theta|x)\,d\theta = 95\%$$

$$\pi_{\nu}(\nu) = \pi(\nu|y) \propto L(y|\nu)\pi_0(\nu)$$

$$p(\theta,\nu|x) \propto L(x|\theta,\nu)\pi_{\theta}(\theta)\pi_{\nu}(\nu)$$

$$p(\theta,\nu|x,y) \propto L(x|\theta,\nu)L(y|\nu)\pi_{\theta}(\theta)\pi_0(\nu)$$

$$p(\theta|x) = \int p(\theta,\nu|x)\,d\nu \propto \int L(x|\theta,\nu)\pi_{\nu}(\nu)\pi_{\theta}(\theta)\,d\nu = L_{\text{m}}(x|\theta)\pi_{\theta}(\theta)$$

$$L_{\text{m}}(x|\theta) = \int L(x|\theta,\nu)\,\pi_{\nu}(\nu)\,d\nu$$

$$p(\theta|x) = \int p(\theta,\nu|x)\,d\nu$$

$$p(\theta,\nu|x) \propto L(x|\theta,\nu)\pi(\theta,\nu)$$

$$\pi_{\theta}(\theta)$$

$$\pi_{\nu}(\nu)$$

$$\pi(\theta,\nu) = \pi_{\theta}(\theta)\pi_{\nu}(\nu)$$

$$\pi(\theta, \nu)$$

$$p_{\theta} = \int_{t_{\theta, \text{obs}}}^{\infty} f(t_{\theta} | \theta, \nu) dt_{\theta}$$

$$t_{\theta} = -2 \ln \frac{L(\theta, \hat{\nu}(\theta))}{L(\hat{\theta}, \hat{\nu})}$$

$$L_{\text{p}}(\theta) = L(\theta, \hat{\nu}(\theta))$$

$$\hat{\hat{\nu}}(\theta) = \operatorname{argmax}_{\nu} L(\theta, \nu)$$

$$L(\theta)$$

$$L(x|\theta)$$

$$L_{\text{p}}(\mu) = L(\mu, \hat{\hat{\theta}}(\mu))$$

$$L_{\text{m}}(\mu) = \int L(\mu, \theta) \pi_{\theta}(\theta) d\theta$$

$$L(L_1, L_2, T, \tilde{\tau}_0, \tilde{\alpha}_1, \tilde{\alpha}_2 | \lambda, \tau, \tau_0, \alpha_1, \alpha_2) =$$

$$\frac{1}{\sqrt{2\pi}\sigma_T}e^{-(T-\tau)^2/2\sigma_T^2}\prod_{i=1}^2\frac{1}{\sqrt{2\pi}\sigma_i}e^{-(L_i-\lambda+\alpha_i(\tau-\tau_0))^2/2\sigma_i^2}$$

$$\times \frac{1}{\sqrt{2\pi}\sigma_{\tilde{\tau}_0}}e^{-(\tilde{\tau}_0-\tau_0)^2/2\sigma_{\tilde{\tau}_0}^2}$$

$$\prod_{i=1}^2\frac{1}{\sqrt{2\pi}\sigma_{\tilde{\alpha}_i}}e^{-(\tilde{\alpha}_i-\alpha_i)^2/2\sigma_{\tilde{\alpha}_i}^2}$$

$$\text{cov}[y_1, y_2] = \alpha_1 \alpha_2 \sigma_T^2$$

$$\text{cov}[y_i, T] = \alpha_i \sigma_T^2$$

$$y_i = L_i + \alpha_i (T - \tau_0)$$

$$L(T, L_1, L_2 | \lambda, \tau) = \frac{1}{\sqrt{2\pi}\sigma_T}e^{-(T-\tau)^2/2\sigma_T^2}\prod_{i=1}^2\frac{1}{\sqrt{2\pi}\sigma_i}e^{-(L_i-\lambda+\alpha_i(\tau-\tau_0))^2/2\sigma_i^2}$$

$$L_i \sim \text{Gauss}(\lambda - \alpha_i(\tau - \tau_0), \sigma_i)$$

$$T \sim \text{Gauss}(\tau, \sigma_T)$$

$$E[L_i] = \lambda - \alpha_i(T - \tau_0), \qquad i = 1, 2$$

$$\text{cov}[\theta_i, \theta_j] = \int \theta_i \theta_j p(\boldsymbol{\theta} | x) \, d\boldsymbol{\theta} \, - \, \int \theta_i p(\boldsymbol{\theta} | x) \, d\boldsymbol{\theta} \, \int \theta_j p(\boldsymbol{\theta} | x) \, d\boldsymbol{\theta}$$

$$\rho_{xy} = \frac{\text{cov}[x,y]}{\sigma_x\sigma_y}$$

$$\text{cov}[x,y] = E[xy] - E[x]E[y]$$

$$t_\theta = -2\ln \frac{L(\theta)}{L(\hat{\theta})}$$

$$\widehat{V}_{ij}^{-1} \approx -\frac{\partial^2 \ln L}{\partial \theta_i \, \partial \theta_j} \Big|_{\theta = \hat{\theta}}$$

$$b=E[\hat{\theta}]-\theta$$

$$V_{ij} = \text{cov}[\hat{\theta}_i, \hat{\theta}_j]$$

$$\hat{\theta} = \operatorname{argmax}_{\theta} L(x|\theta)$$

$$p(\theta|x) = \frac{L(x|\theta)\pi(\theta)}{\int L(x|\theta)\pi(\theta)\,d\theta}$$

$$Z = \sqrt{2\left((s+b)\ln\left(1+\frac{s}{b}\right)-s\right)}$$

$$t_{\text{float}} = -2\ln \frac{L(0)}{L(\hat{\mu}, \hat{m})} \equiv -2\ln \lambda(0)$$

$$\chi^2(\boldsymbol{\mu}) + \tau \sum_i [(\mu_{i+1} - \mu_i) - (\mu_i - \mu_{i-1})^2]$$

$$\chi^2 = \sum_{i=1}^N \frac{(\nu_{0i} - n_i)^2}{\nu_{0i}}$$

$$\nu_{0i} = \sum_{j=1}^M R_{ij} \mu_{0j} + \beta_i$$

$$\Phi(\boldsymbol{\mu}) = \alpha \left(\ln L(\boldsymbol{\mu}) + \ln L_{\theta}(\theta) \right) + S(\boldsymbol{\mu})$$

$$C_i=0.1$$

$$n_i=100$$

$$\beta_i=0$$

$$\hat{\mu}_i = C_i n_i = 10$$

$$\sigma_{\hat{\mu}_i} = C_i \sqrt{n_i} = 1.0$$

$$E[\hat{\boldsymbol{\mu}}] = R^{-1}(E[\mathbf{n}] - \boldsymbol{\beta}) = \boldsymbol{\mu}$$

$$U_{ij} = \text{cov}[\hat{\mu}_i, \hat{\mu}_j] = \sum_{k,l=1}^N (R^{-1})_{ik} (R^{-1})_{jl} \text{cov}[n_k, n_l]$$

$$= \sum_{k=1}^N (R^{-1})_{ik} (R^{-1})_{jk} \, \nu_k$$

$$\frac{\partial^2 \ln L}{\partial \mu^2} = -\frac{n}{\mu^2}$$

$$\ln L(\boldsymbol{\mu}) = \sum_{i=1}^N (n_i \ln \nu_i - \nu_i)$$

$$\hat{\boldsymbol{\nu}} = \mathbf{n}$$

$$\hat{\boldsymbol{\mu}} = R^{-1}(\mathbf{n} - \boldsymbol{\beta})$$

$$P(n_i;\nu_i)=\frac{\nu_i^{n_i}}{n_i!}e^{-\nu_i}$$

$$\boldsymbol{\mu} = R^{-1}(\boldsymbol{\nu} - \boldsymbol{\beta})$$

$$\boldsymbol{\mu} = (\mu_1, \ldots, \mu_M), \quad \mu_{\text{tot}} = \sum_{j=1}^M \mu_j$$

$$\mathbf{p} = (p_1, \ldots, p_M) = \boldsymbol{\mu}/\mu_{\text{tot}}$$

$$\boldsymbol{\nu} = (\nu_1, \dots, \nu_N)$$

$$\varepsilon_j = \sum_{i=1}^N R_{ij}$$

$$\boldsymbol{\beta} = (\beta_1, \dots, \beta_N)$$

$$E[\mathbf{n}] = \boldsymbol{\nu} = R\boldsymbol{\mu} + \boldsymbol{\beta}$$

$$\sum_{i=1}^N R_{ij} = \sum_{i=1}^N P(\text{observed in bin } i \mid \text{true value in bin } j)$$

$$= P(\text{observed anywhere} \mid \text{true value in bin } j)$$

$$= \varepsilon_j$$

$$\nu_i = \sum_{j=1}^M R_{ij} \mu_j + \beta_i$$

$$\mathbf{n} = (n_1, \dots, n_N)$$

$$\nu_i = E[n_i] = \sum_{j=1}^M R_{ij} \mu_j \; , \quad i = 1, \dots, N$$

$$R_{ij} = P(\text{observed in bin } i \mid \text{true in bin } j)$$

$$f_{\mathrm{meas}}(x) = \int R(x|y) f_{\mathrm{true}}(y) \, dy$$

$$p_j = \int_{\mathrm{bin}\,j} f(y) \, dy$$

$$\mu_j = \mu_{\mathrm{tot}} p_j$$

$$\pi(b) = \frac{1}{\sqrt{2\pi}\sigma_b} e^{-(b-b_{\mathrm{meas}})^2/2\sigma_b^2}$$

$$p_{\mathrm{float}} \approx p_{\mathrm{fix}} + \langle N(c) \rangle$$

$$c = t_{\mathrm{fix}} = Z_{\mathrm{fix}}^2$$

$$Z_{\text{fix}} = \Phi^{-1}(1 - p_{\text{fix}})$$

$$\mathcal{N} = \frac{\langle N(c) \rangle}{1 - F_{\chi^2_2}(c)}$$

$$\langle N(c) \rangle \approx \langle N(c_0) \rangle e^{-(c-c_0)/2}$$

$$F_{\text{trials}} \equiv \frac{p_{\text{float}}}{p_{\text{fix}}} \approx \sqrt{\frac{\pi}{2}} \frac{\langle N(t_{\text{float,obs}}) \rangle}{1 - F_{\chi^2_2}(t_{\text{float,obs}})} Z_{\text{fix}}$$

$$F_{\text{trials}} \equiv \frac{p_{\text{float}}}{p_{\text{fix}}} \approx \sqrt{\frac{\pi}{2}} \mathcal{N} Z_{\text{fix}}$$

$$P(t_{\text{float}} > t_{\text{float,obs}} | H_0) \leq 1 - F_{\chi^2_1}(t_{\text{float,obs}}) + \langle N(t_{\text{float,obs}}) \rangle$$

$$-2\ln\lambda(\mu)=-2\ln(L(\mu)/L(\hat{\mu}))<1\text{ i.e., }\ln L(\mu)>\ln L(\hat{\mu})-\frac{1}{2}$$

$$\frac{P(\mathbf{x}|H_1)}{P(\mathbf{x}|H_0)} \geq k$$

$$x \sim \text{Gauss}(\mu, \sigma)$$

$$\alpha = 1 - \Phi\left(\frac{x_{\text{c}} - \mu_0}{\sigma}\right)$$

$$\text{power} = 1 - \beta = P(x > x_{\text{c}} | \mu) = 1 - \Phi\left(\frac{\mu_0 - \mu}{\sigma} + \Phi^{-1}(1 - \alpha)\right)$$

$$x_{\text{c}} = \mu_0 + \sigma \Phi^{-1}(1 - \alpha)$$

$$L(x,y|\theta,\nu)=L(x|\theta,\nu)L(y|\nu)$$

$$\pi(\nu|y) \propto L(y|\nu)\pi_0(\nu)$$

$$L_{\text{m}}(x|\theta) = \int L(x|\theta,\nu)\pi(\nu)\,d\nu$$

$$\hat{\nu}(\theta)$$

$$p_{\theta} = \int_{q_{\theta,\text{obs}}}^{\infty} f(q_{\theta}|\theta,\nu)\,dq_{\theta}$$

$$L(x|\theta) \rightarrow L(x|\theta, \nu)$$

$$L(x|\theta)=\theta x$$

$$L(x|\theta)=\theta x+\alpha x^2+\beta x^3+\cdots$$

$$p_\theta=\int_{q_{\theta,\text{obs}}}^\infty f(q_\theta|\theta)\,dq_\theta$$

$$p(s|n)=\int p(s,b|n)\,db$$

$$p(\mathbf{x}|H_0), p(\mathbf{x}|H_1)$$

$$s_{\text{up}} = \tfrac{1}{2} F_{\chi^2}^{-1}(1-\alpha; 2(n+1)) - b$$

$$\alpha$$

$$s_{\text{up}} = -\ln \alpha \approx 3.00$$

$$P(\text{D}) \quad = \quad 0.001$$

$$P(\text{no D}) \quad = \quad 0.999$$

$$P(+|\text{D}) \quad = \quad 0.98$$

$$P(-|\text{D}) \quad = \quad 0.02$$

$$P(+|\text{no D}) \quad = \quad 0.03$$

$$P(-|\text{no D}) \quad = \quad 0.97$$

$$\begin{aligned} p(\text{D}|+) &= \frac{P(+|\text{D})P(\text{D})}{P(+|\text{D})P(\text{D}) + P(+|\text{no D})P(\text{no D})} \\ &= \frac{0.98 \times 0.001}{0.98 \times 0.001 + 0.03 \times 0.999} \\ &= 0.032 \end{aligned}$$