

If form of posterior and prior are the same that is a conjugate prior.

Laplace's Prior: prior of indifference / uniform. In the Poisson model θ is in the set $(0, \infty)$, so we need a distribution which is uniform on that set. A distribution would look like $P(\theta) = c > 0$, $\int_0^\infty P(\theta) d\theta = \int_0^\infty c d\theta = c \int_0^\infty d\theta = \infty$ does not exist. There cannot be a proper Laplace prior, but there is an improper Laplace prior.

$$P(\theta|x) \propto P(x|\theta) P(\theta) = e^{-n\theta} \theta^{\sum x_i} p(\theta) \propto e^{-n\theta} \theta^{\sum x_i + 1 - 1} \propto \text{Gamma}(\sum x_i + 1, n)$$

$P(\theta) \propto 1 \Rightarrow$ Laplace's Idea \uparrow diverges

$$\propto \text{Gamma}(1 + \sum x_i, n) \Rightarrow P(\theta) = \text{Gamma}(1, 0), \text{ an improper prior.}$$

\Downarrow
1 success in no trials, nonsense!

Is the posterior proper? α, β must both be > 0 .

Yes always since $\sum x_i \geq 0$ its first parameter is always $\geq 1 > 0$ and since $n \geq 1$ its second parameter is always $\geq 1 > 0$.

Haldane's Prior of complete ignorance. Setting all pseudodata to be zero i.e. $x_0 = 0, n_0 = 0 \Rightarrow P(\theta) = \text{Gamma}(0, 0)$ obviously improper. \Rightarrow

$$P(\theta|x) = \text{Gamma}(\theta + \sum x_i, \theta + n) = \text{Gamma}(\sum x_i, n). \text{ Is the posterior}$$

proper? Only if $\sum x_i > 0$. If you don't get at least one success it blows up.

$$\hat{\theta}_{\text{MMSE}} = \frac{\sum x_i}{n} = \bar{x} = \hat{\theta}_{\text{MLE}}. \text{ Now on to Jeffrey's.}$$

Jeffrey's prior, we need Fisher Information. $P_J(\theta) \propto \sqrt{I(\theta)}$

From previous notes: $\ell'(\theta) = -n + \frac{\sum x_i}{\theta} \Rightarrow \ell''(\theta) = -\frac{\sum x_i}{\theta^2}$, then

$$I(\theta) = E_x[-\ell''(\theta)] = E\left[\frac{\sum x_i}{\theta^2}\right] \stackrel{\text{iid}}{=} \frac{n E[X_i]}{\theta^2}. \text{ Need expectation of Poisson}$$

$$X \sim \text{Poisson}(\theta). E[X] = \sum_{x \in \text{Supp}[X]} x p(x) = \sum_{x=0}^{\infty} x \frac{e^{-\theta} \theta^x}{x!} = e^{-\theta} \sum_{x=1}^{\infty} \frac{x \theta^x}{x!} =$$

re-index trick $y = x-1 \Leftrightarrow x = y+1$

first term $x=0$ is 0 in the sum (cancel)

$$= e^{-\theta} \sum_{y=0}^{\infty} \frac{\theta^{y+1}}{y!} = e^{-\theta} \theta \sum_{y=0}^{\infty} \frac{\theta^y}{y!} = \theta e^{-\theta} e^{\theta} = \theta$$

taylor series for e^{θ}

$$E[X] = \theta \Rightarrow I(\theta) = \frac{n\theta}{\theta^2} = \frac{n}{\theta}$$

Free variable change at will constant - you cannot change

$$P_J(\theta) \propto \sqrt{I(\theta)} = \sqrt{\frac{n}{\theta}} \propto \theta^{-\frac{1}{2}} \propto \text{Gamma}\left(\frac{1}{2}, \theta\right) \text{ improper}$$

\Downarrow Posterior

$$p(\theta|x) = \text{Gamma}\left(\sum x_i + \frac{1}{2}, \theta + n\right)$$

Is Jeffrey's prior proper, yes, always, same reason for Laplace n always > 0 .

Posterior Predictive Distribution. You see n observations and want to know the distribution of n^* future observations.

$$\underbrace{x_1 \ x_2 \ x_3 \ \dots \ x_n}_{n \text{ i.i.d. Poisson}} \mid \underbrace{x_{n+1} \ \dots \ x_{n+n^*}}_{n^* = 1} \mid X \sim ?$$

past future

$$P(X_n | X) = \int_0^\infty \underbrace{P(X_n | \theta)}_{\text{Poisson}} \underbrace{P(\theta | X)}_{\text{Gamma}} d\theta = \int_0^\infty \frac{e^{-\theta} \theta^{x_n}}{x_n!} \frac{(\beta+n)^{\alpha+\sum x_i}}{\Gamma(\alpha+\sum x_i)} \theta^{\alpha+\sum x_i-1} e^{-(\beta+n)\theta} d\theta$$

$$= \frac{(\beta+n)^{\alpha+\sum x_i}}{x_n! \Gamma(\alpha+\sum x_i)} \int_0^\infty e^{-\theta} \theta^{x_n} \theta^{\alpha+\sum x_i-1} e^{-(\beta+n)\theta} d\theta$$

$$= \int_0^\infty \theta^{x_n + \alpha + \sum x_i - 1} e^{-(\beta+n+1)\theta} d\theta$$

now we do a substitution. let $\frac{t}{\beta+n+1} = (\beta+n+1)\theta \Rightarrow \theta = \frac{t}{\beta+n+1} \Rightarrow$

$$\frac{d\theta}{dt} = \frac{1}{\beta+n+1} \Rightarrow d\theta = \frac{dt}{\beta+n+1}. \text{ If } \theta=0 \Rightarrow t=0 \text{ if } \theta=\infty \Rightarrow t=\infty$$

So nothing changes in bounds of integral.

$$= \frac{(\beta+n)^{\alpha+\sum x_i}}{x_n! \Gamma(\alpha+\sum x_i)} \int_0^\infty \frac{t^{x_n + \alpha + \sum x_i - 1}}{(\beta+n+1)^{x_n + \alpha + \sum x_i - 1}} \cdot \frac{1}{(\beta+n+1)} e^{-t} dt$$

cancels

$$= \frac{(\beta+n)^{\alpha+\sum x_i}}{x_n! \Gamma(\alpha+\sum x_i) (\beta+n+1)^{x_n + \alpha + \sum x_i}} \int_0^\infty t^{(x_n + \alpha + \sum x_i) - 1} e^{-t} dt$$

gamma integral

$$= \frac{(\beta+n)^{\alpha+\sum x_i} \Gamma(x_n + \alpha + \sum x_i)}{x_n! \Gamma(\alpha+\sum x_i) (\beta+n+1)^{x_n + \alpha + \sum x_i}}$$

$$= \frac{(\beta+n)^{\alpha + \sum x_i}}{(\beta+n+1)^{\alpha + \sum x_i}} \cdot \frac{1}{(\beta+n+1)^{x_n}} \cdot \frac{\Gamma(x_n + \alpha + \sum x_i)}{x_n! \Gamma(\alpha + \sum x_i)}$$

$$= \left(\frac{\beta+n}{\beta+n+1} \right)^{\alpha + \sum x_i} \left(\frac{1}{\beta+n+1} \right)^{x_n} \frac{\Gamma(x_n + \alpha + \sum x_i)}{x_n! \Gamma(\alpha + \sum x_i)}$$

$$= p^r (1-p)^{x_n} \frac{\Gamma(x_n + r)}{x_n! \Gamma(r)} =$$

Extended Negative Binomial

= Ext Neg Bin (r, p) as r.v.~~or~~

rewrite using this

$$\text{Let } p = \frac{\beta+n}{\beta+n+1} \in (0,1)$$

Because $\beta > 0, n \geq 0$

$$1-p = \frac{1}{\beta+n+1} \in (0,1)$$

$$\text{Let } r = \sum x_i + \alpha > 0$$

If $\alpha \in \{0, 1, 2, \dots\}$ then

$$= \binom{x_n + r - 1}{r} p^r (1-p)^{x_n} = \text{Neg Bin}(r, p)$$

From 368 the neg binomial is the sum of iid geometric random variables. Since the expectation of the geometric rv is

$\frac{(1-p)}{p}$, the expectation of the neg bin by linearity is

$$p(x_n | x) = \text{Ext Neg Bin}(r, p) \Rightarrow E[x_n | x] = r \left(\frac{1-p}{p} \right)$$

$$= \frac{(\sum x_i + \alpha)}{n + \beta} = \hat{\theta}_{\text{MMSE}}$$