

FISH TALES

If you give a man a fish he will eat for a day. If you teach a man to fish he will eat for a lifetime.

POISSON TALES

If you give a man Cash he will analyze for a day. If you teach a man Poisson, he will analyze for a lifetime.

It is all about Probabilities!

- The Poisson Likelihood
- Probability Calculus – Bayes' Theorem
- Priors – the gamma distribution
- Source intensity and background marginalization
- Hardness Ratios (BEHR)

Deriving the Poisson Likelihood

- N counts uniformly distributed in a duration τ (rate $R = N/\tau$)
- what is the probability of finding k counts in an interval δt ?
- probability of “success” (choose an interval with a count)

$$\rho = \frac{\delta t}{\tau} \equiv \frac{R\delta t}{N}$$

- probability of finding k events in this interval

$${}^N C_k \rho^k (1 - \rho)^{N-k}$$

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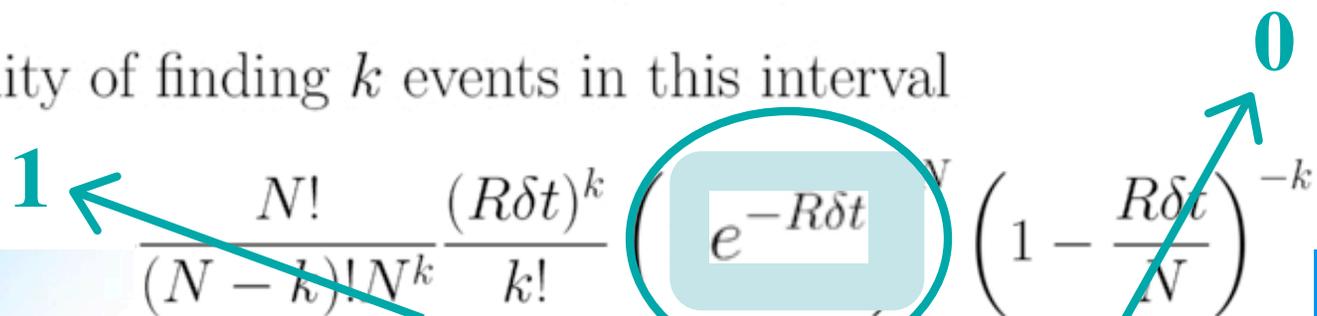
$N, \tau \rightarrow \infty$

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$$\frac{N!}{(N-k)!N^k} \frac{(R\delta t)^k}{k!} \left(e^{-R\delta t} \right)^N \left(1 - \frac{R\delta t}{N} \right)^{-k}$$


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$$p(k|R\delta t) = \frac{(R\delta t)^k e^{-R\delta t}}{k!}$$

Probability Calculus

$$\textit{A or B} :: p(A+B) = p(A) + p(B) - p(AB)$$

$$\textit{A and B} :: p(AB) = p(A|B) p(B) = p(B|A) p(A)$$

$$\textit{Bayes' Theorem} :: p(B|A) = p(A|B) p(B) / p(A)$$

$$p(\text{Model} | \text{Data}) = p(\text{Model}) p(\text{Data} | \text{Model}) / p(\text{Data})$$

posterior
distribution

prior
distribution

likelihood normalization

$$\textit{marginalization} :: p(a|D) = \int p(ab|D) db$$

Priors



Priors

- Incorporate known information
- Forced acknowledgement of bias
- Non-informative priors
 - flat
 - range
 - least informative (Jeffrey's)
 - gamma

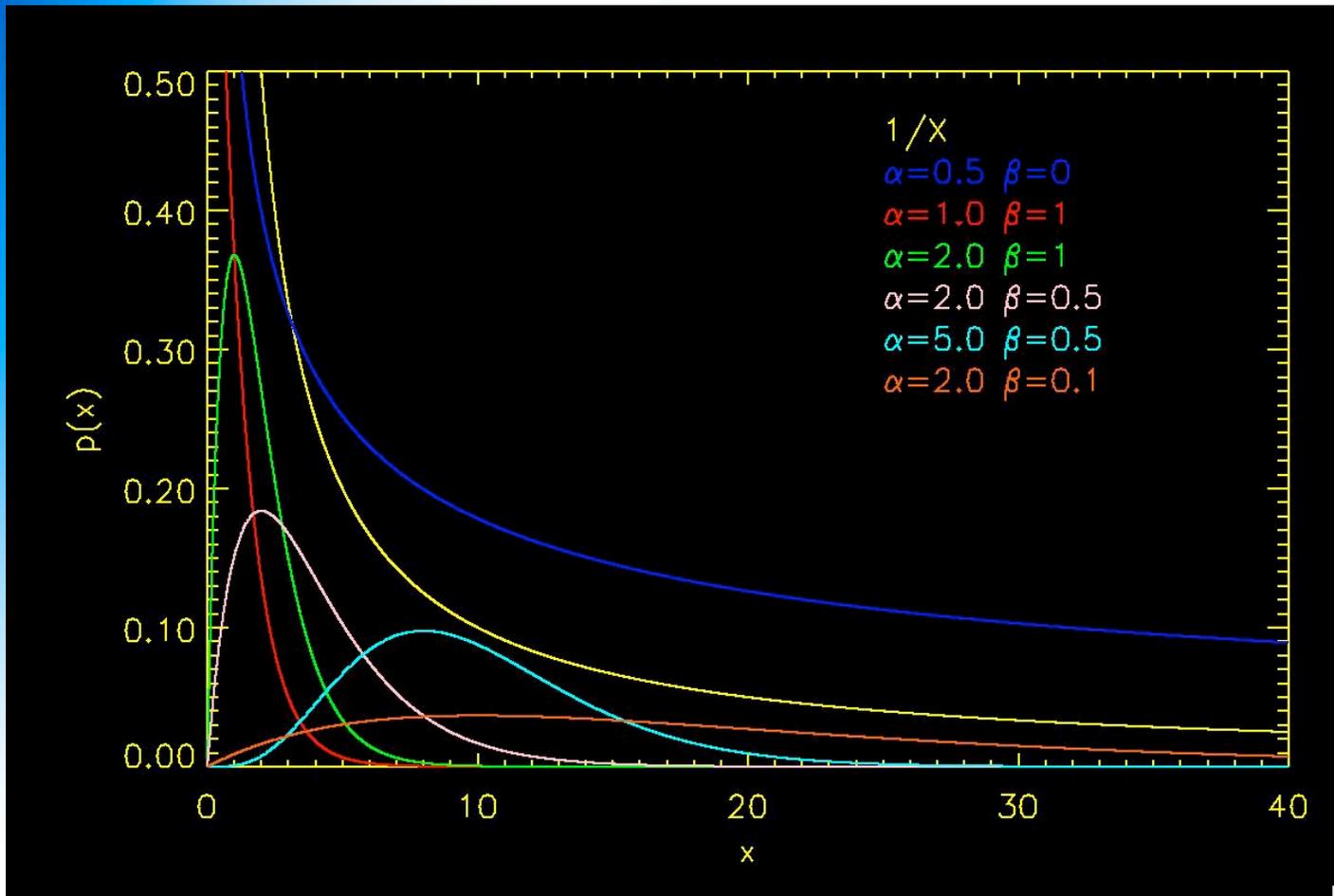
Priors

the gamma distribution $\gamma(x; \alpha, \beta) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x}$

$$\bar{x} = \frac{\alpha}{\beta}, \quad \overline{x^2} - \bar{x}^2 = \frac{\alpha}{\beta^2}$$

Priors

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Panning for Gold: Source and Background

$$p(b|N_B I) = \frac{p(b|I)p(N_B|bI)}{p(N_B|I)}$$

$$p(b|N_B I) = \frac{1}{p(N_B|I)} \frac{\beta_B^{\alpha_B} \left(\frac{a_B}{a_S}\right)^{N_B}}{\Gamma(\alpha_B)\Gamma(N_B + 1)} b^{N_B + \alpha_B - 1} e^{-b\left(\beta_B + \frac{a_B}{a_S}\right)}$$

Panning for Gold: Source and Background

$$p(b|N_B I) = \frac{p(b|I)p(N_B|bI)}{p(N_B|I)}$$

$$p(s|N_S I) = \int db p(sb|N_S I)$$

$$p(sb|N_S I) = \frac{p(b|N_B I)p(s|I)p(N_S|sbI)}{p(N_S|I)}$$

$$p(sb|N_S I) = \frac{(\beta_B + r)^{\alpha_B + N_B} \beta_S^{\alpha_S}}{\Gamma(\alpha_B + N_B)\Gamma(\alpha_S)} \frac{b^{\alpha_B + N_B - 1} s^{\alpha_S - 1} (s + b)^{N_S} e^{-b(1 + (\beta_B + r))} e^{-s(1 + \beta_S)}}{\Gamma(N_S + 1) p(N_S|I)}$$

Panning for Gold: Source and Background

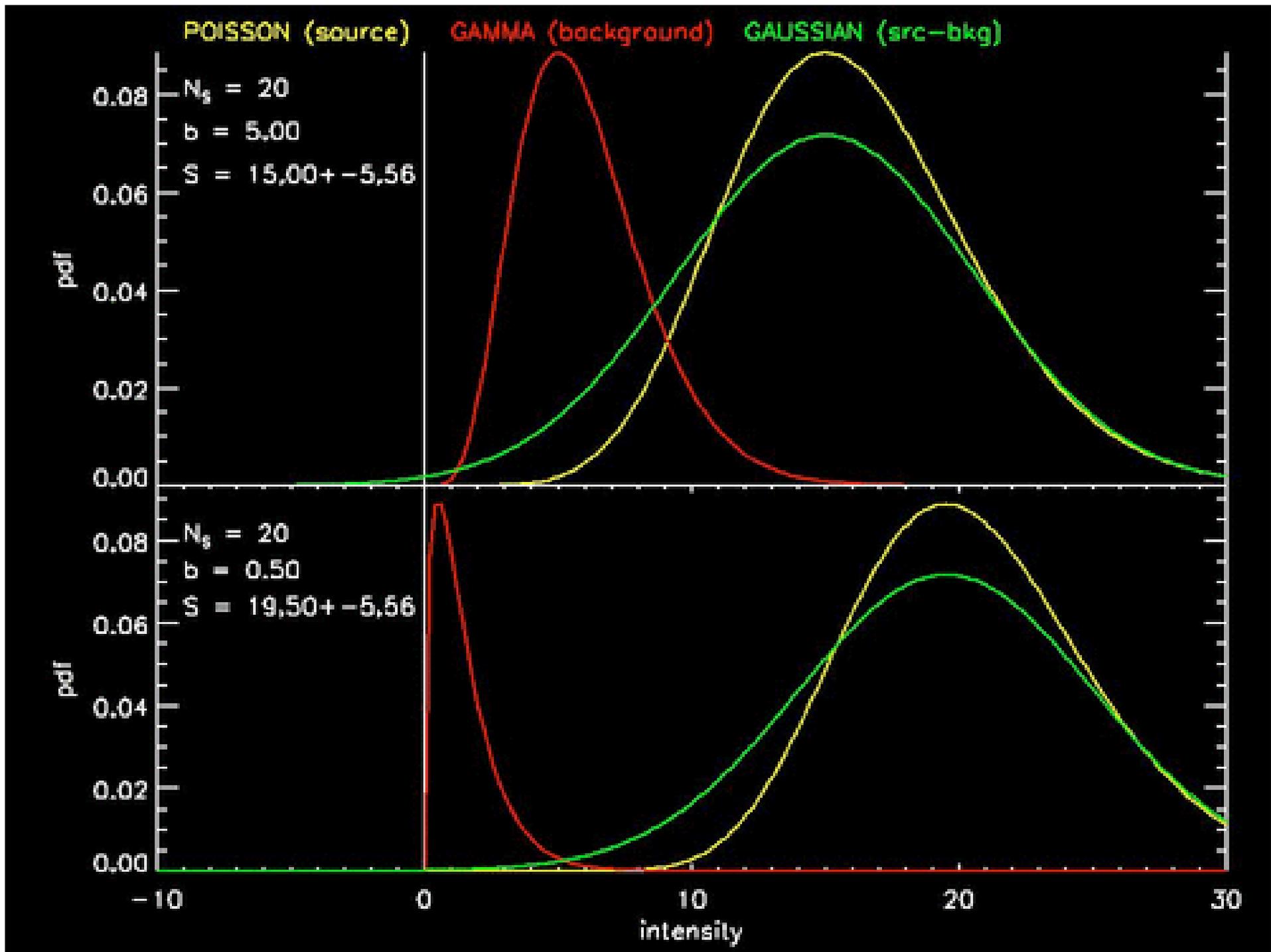
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$$p(s|N_S I) = \left(\frac{1}{\sum_{k=0}^{N_S} \mathcal{I}_k^{bs}} \right) e^{-s(1+\beta)} \sum_{k=0}^{N_S} \frac{\Gamma(\alpha_B + N_B + N_S - k)}{\Gamma(k+1)\Gamma(N_S - k + 1)} \frac{s^{k+\alpha_S-1}}{(\beta_B + r + 1)^{\alpha_B + N_B + N_S - k}}$$

$$\mathcal{I}_k^{bs} = \frac{\Gamma(\alpha_B + N_B + N_S - k)\Gamma(k + \alpha_S)}{\Gamma(k+1)\Gamma(N_S - k + 1)} \frac{1}{(\beta_B + r + 1)^{\alpha_B + N_B + N_S - k} (1 + \beta_S)^{k + \alpha_S}}$$



Hardness Ratios

Simple Ratio, $R \equiv \frac{S}{H}$

Color, $C \equiv \log_{10} \left(\frac{S}{H} \right)$

Fractional Difference, $HR \equiv \frac{H - S}{H + S}$

Simple, robust, intuitive summary

Proxy for spectral fitting

Useful for large samples

Most needed for low counts

Hardness Ratios

$$\text{Simple Ratio, } R = \frac{\theta_S}{\theta_H}$$

$$\text{Color, } C = \log_{10} \left(\frac{\theta_S}{\theta_H} \right)$$

$$\text{Fractional Difference, } HR = \frac{\theta_H - \theta_S}{\theta_H + \theta_S}$$

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Proxy for spectral fitting

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$$R = \theta_S / \theta_H,$$

$$\begin{aligned} & p(R, \theta_H | S, H, B_S, B_H) dR d\theta_H \\ &= p(\theta_S, \theta_H | S, H, B_S, B_H) \left| \frac{\partial(\theta_S, \theta_H)}{\partial(R, \theta_H)} \right| d\theta_S d\theta_H \\ &= p(R\theta_H, \theta_H | S, H, B_S, B_H) \theta_H dR d\theta_H \end{aligned}$$

$$C = \log_{10}(\theta_S / \theta_H),$$

$$\begin{aligned} & p(C, \theta_H | S, H, B_S, B_H) dC d\theta_H \\ &= p(\theta_S, \theta_H | S, H, B_S, B_H) \left| \frac{\partial(\theta_S, \theta_H)}{\partial(C, \theta_H)} \right| d\theta_S d\theta_H \\ &= p(10^C \theta_H, \theta_H | S, H, B_S, B_H) 10^C \ln(10) \theta_H dC d\theta_H \end{aligned}$$

$$HR = (\theta_H - \theta_S) / \omega,$$

$$\omega = \theta_S + \theta_H,$$

$$\begin{aligned} & p(HR, \omega | S, H, B_S, B_H) dHR d\omega \\ &= p(\theta_S, \theta_H | S, H, B_S, B_H) \left| \frac{\partial(\theta_S, \theta_H)}{\partial(HR, \omega)} \right| d\theta_S d\theta_H \\ &= p\left(\frac{(1 - HR)\omega}{2}, \frac{(1 + HR)\omega}{2} \middle| S, H, B_S, B_H\right) \frac{\omega}{2} dHR d\omega \end{aligned}$$

BEHR

<http://hea-www.harvard.edu/AstroStat/BEHR/>

