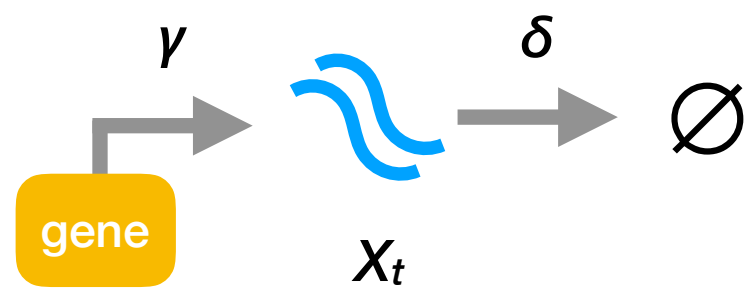


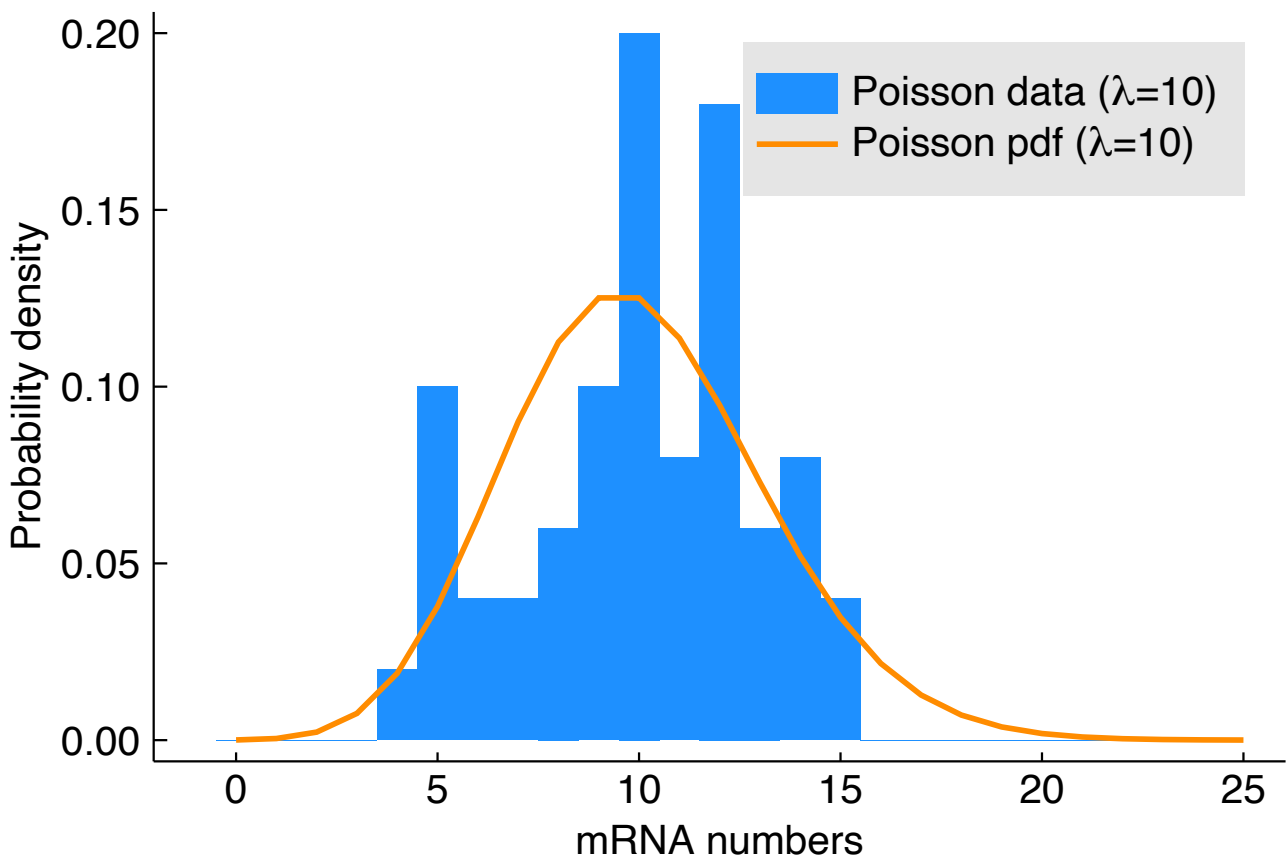
Poisson example



Constitutive gene expression

$X_{\text{steady state}} \sim \text{Poi}(\frac{\gamma}{\delta}) = \text{Poi}(\lambda)$
with $\lambda = \frac{\gamma}{\delta}$

Data (N=50)



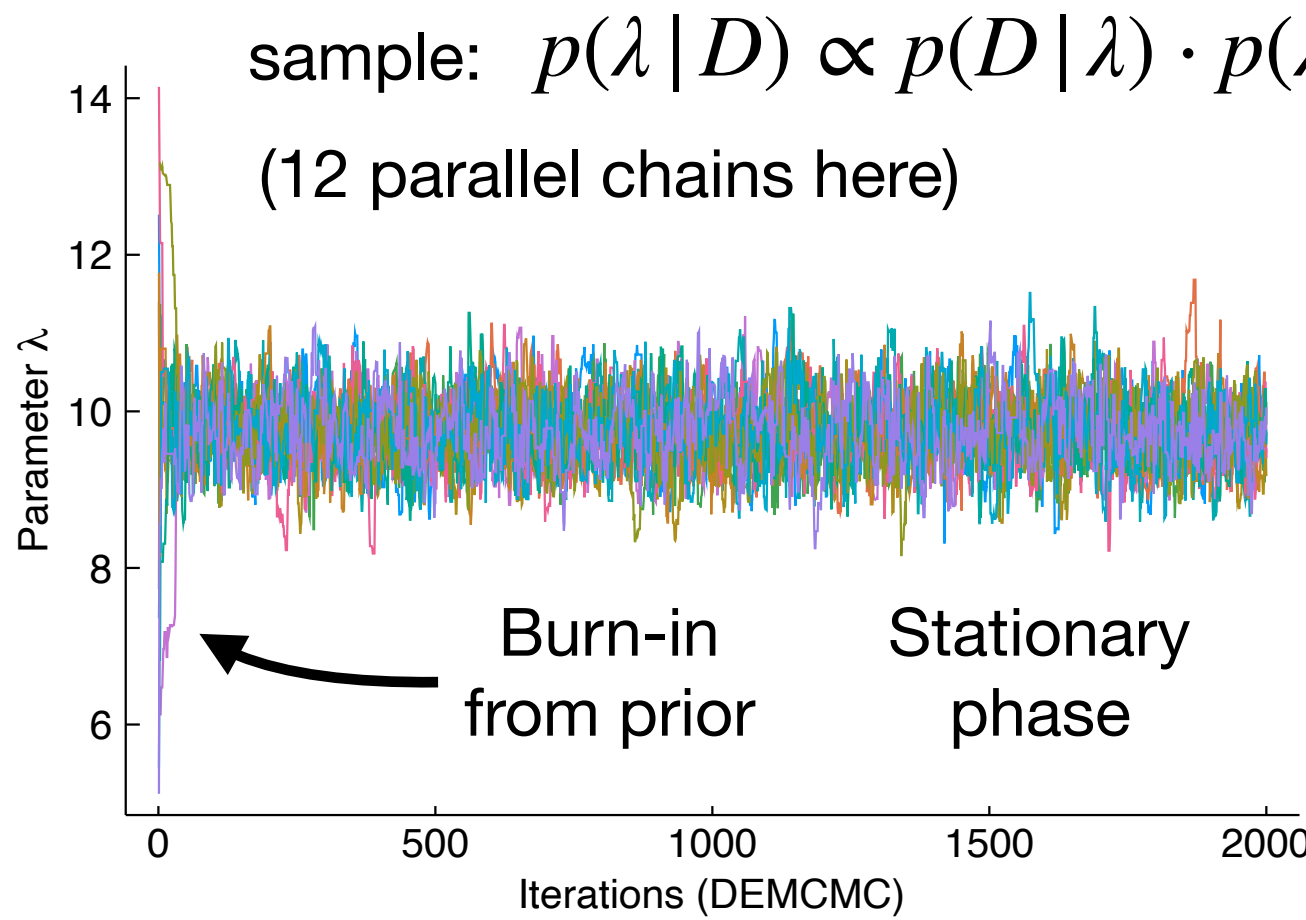
Analytical result

Prior (conjugated)
 $\lambda \sim \Gamma(\alpha = 10, \beta = 1) \iff p(\lambda) = f_{\Gamma(10,1)}(\lambda)$

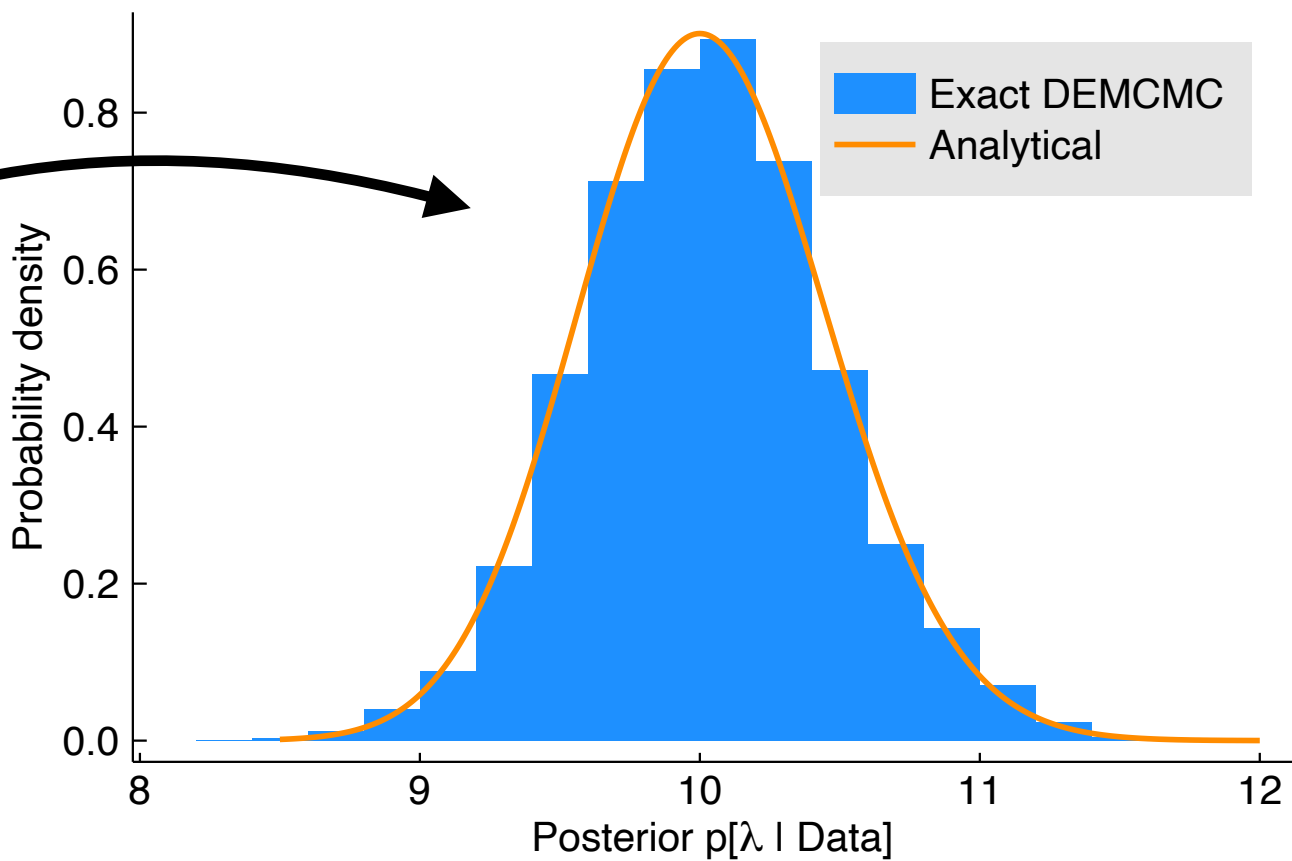
Likelihood (exact)
 $p(D | \lambda) = \prod_{i=1}^N p(y_i | \lambda) = \prod_{i=1}^N f_{\text{Poi}(\lambda)}(y_i)$

\Rightarrow Posterior (analytical)
 $\lambda | D \sim \Gamma(10 + \sum_{i=1}^N y_i, N + 1) \iff p(\lambda | D) = f_{\Gamma}(\lambda)$

MCMC algorithms



Posterior samples (as histogram)



Julia implementation

```
λtrue = 10
priors = (λ=(Gamma(λtrue, 1),),)

function loglike(λ, data)
    return sum(logpdf.(Poisson(λ), data))
end

model = DEModel(priors=priors, model=loglike)
de = DE(burnin=1000, priors=priors)

chains_exact = psample(model, de, 2000)
```