

Ejemplos

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Paper: *Modelos ocultos de Markov:*

una aplicación de estimación Bayesiana para series de tiempo financieras

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<https://github.com/lizbethna/HMMBayes.git>

Este archivo muestra las instrucciones para correr los códigos de R y Stan.

Distribución Gamma-Poisson

```
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 4.1.2
```

```
library(extraDistr)
library(rstan)
```

```
## Loading required package: StanHeaders
```

```
## rstan (Version 2.21.3, GitRev: 2e1f913d3ca3)
```

```
## For execution on a local, multicore CPU with excess RAM we recommend calling
```

```
## options(mc.cores = parallel::detectCores()).
```

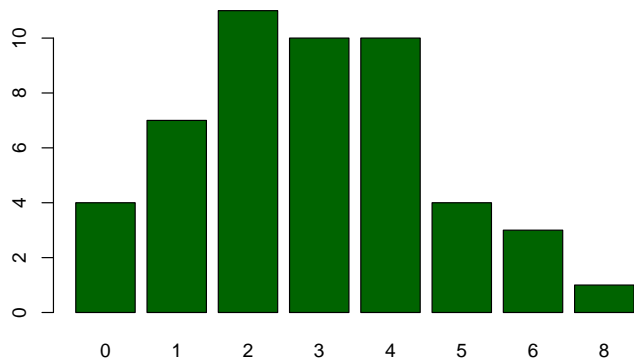
```
## To avoid recompilation of unchanged Stan programs, we recommend calling
```

```
## rstan_options(auto_write = TRUE)
```

Simular datos

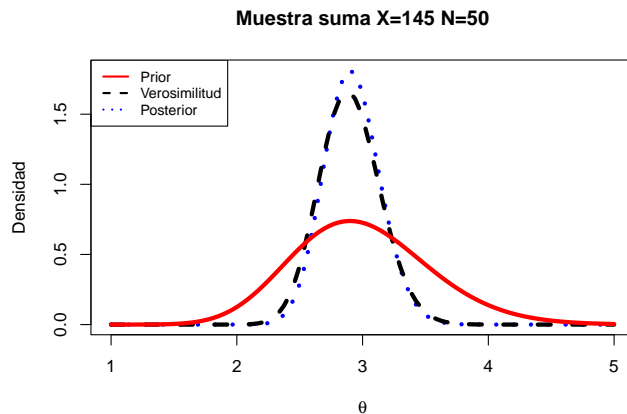
```
N = 50    # tamaño de muestra
theta = 3  # parametro de media
a0 = 30; b0 = 10 # hiperparametros de la distribucion inicial

set.seed(12345)
x = rpois(N, theta)    #  $x \sim \text{Poisson}(\theta)$ 
barplot(table(x), nclass=10, col="darkgreen")
```

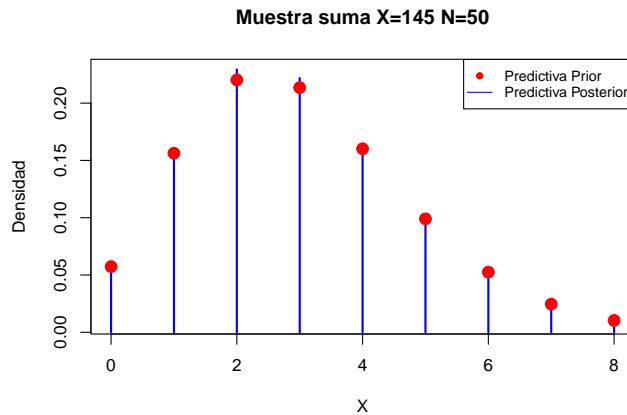


Graficas de las distribuciones final y predictiva final

```
### Posterior
a1 = a0+sum(x)
b1 = b0+N
the0 = seq(quantile(x,0.1),quantile(x,0.9),length.out=100)
plot(the0, dgamma(the0,a1,b1), main=paste0("Muestra suma X=",sum(x)," N=",N),
      xlab=expression(theta), ylab="Densidad", lty=3, lwd=4, col="blue", type="l")
lines(the0, dgamma(the0,sum(x),N), lty=2, lwd=4)
lines(the0,dgamma(the0,a0,b0), lty=1, lwd=4, col="red")
legend("topleft", legend=c("Prior","Verosimilitud","Posterior"),
      lty=c(1,2,3), col=c("red","black","blue"), lwd=2, cex=0.8)
```



```
### Predictiva Prior & Posterior
x0 = (0:max(x))
plot(x0,dgpois(x0, a1, b1), main=paste0("Muestra suma X=",sum(x), " N=",N),
      xlab="X", ylab="Densidad", type="h", lwd=2, col="blue")
points(x0,dgpois(x0, a0, b0), col="red", pch=19, cex=1.5)
legend("topright", legend=c("Predictiva Prior","Predictiva Posterior"),
      lty=c(NA,1), col=c("red","blue"), pch=c(19,NA), cex=0.8)
```



Código Stan

```
datos <- list( "x"=x, "N"=N, # muestra
               "a0"=a0,"b0"=b0) # valores iniciales de la distribucion inicial
param = c("theta","x_star") # parametros a estimar
fit_dist <- stan("dist_poisson_gamma.stan", data=datos,
                chains=2, warmup=1000, iter=2000, thin=2)
```

Running /Library/Frameworks/R.framework/Resources/bin/R CMD SHLIB foo.c

clang -mmacosx-version-min=10.13 -I"/Library/Frameworks/R.framework/Resources/include" -DNDEBUG -I"/L

In file included from <built-in>:1:

In file included from /Library/Frameworks/R.framework/Versions/4.1/Resources/library/StanHeaders/include

In file included from /Library/Frameworks/R.framework/Versions/4.1/Resources/library/RcppEigen/include/Eigen

In file included from /Library/Frameworks/R.framework/Versions/4.1/Resources/library/RcppEigen/include/Eigen

/Library/Frameworks/R.framework/Versions/4.1/Resources/library/RcppEigen/include/Eigen/src/Core/util/Mac

namespace Eigen {

~

/Library/Frameworks/R.framework/Versions/4.1/Resources/library/RcppEigen/include/Eigen/src/Core/util/Mac

namespace Eigen {

~

;

In file included from <built-in>:1:

In file included from /Library/Frameworks/R.framework/Versions/4.1/Resources/library/StanHeaders/include

In file included from /Library/Frameworks/R.framework/Versions/4.1/Resources/library/RcppEigen/include/Eigen

/Library/Frameworks/R.framework/Versions/4.1/Resources/library/RcppEigen/include/Eigen/Core:96:10: fatal

#include <complex>

~~~~~

3 errors generated.

make: \*\*\* [foo.o] Error 1

SAMPLING FOR MODEL 'dist\_poisson\_gamma' NOW (CHAIN 1).

Chain 1:

Chain 1: Gradient evaluation took 1.4e-05 seconds

Chain 1: 1000 transitions using 10 leapfrog steps per transition would take 0.14 seconds.

Chain 1: Adjust your expectations accordingly!

Chain 1:

Chain 1:

Chain 1: Iteration: 1 / 2000 [ 0%] (Warmup)

```

Chain 1: Iteration: 200 / 2000 [ 10%] (Warmup)
Chain 1: Iteration: 400 / 2000 [ 20%] (Warmup)
Chain 1: Iteration: 600 / 2000 [ 30%] (Warmup)
Chain 1: Iteration: 800 / 2000 [ 40%] (Warmup)
Chain 1: Iteration: 1000 / 2000 [ 50%] (Warmup)
Chain 1: Iteration: 1001 / 2000 [ 50%] (Sampling)
Chain 1: Iteration: 1200 / 2000 [ 60%] (Sampling)
Chain 1: Iteration: 1400 / 2000 [ 70%] (Sampling)
Chain 1: Iteration: 1600 / 2000 [ 80%] (Sampling)
Chain 1: Iteration: 1800 / 2000 [ 90%] (Sampling)
Chain 1: Iteration: 2000 / 2000 [100%] (Sampling)
Chain 1:
Chain 1: Elapsed Time: 0.009626 seconds (Warm-up)
Chain 1: 0.008987 seconds (Sampling)
Chain 1: 0.018613 seconds (Total)
Chain 1:

SAMPLING FOR MODEL 'dist_poisson_gamma' NOW (CHAIN 2).
Chain 2:
Chain 2: Gradient evaluation took 3e-06 seconds
Chain 2: 1000 transitions using 10 leapfrog steps per transition would take 0.03 seconds.
Chain 2: Adjust your expectations accordingly!
Chain 2:
Chain 2:
Chain 2: Iteration: 1 / 2000 [ 0%] (Warmup)
Chain 2: Iteration: 200 / 2000 [ 10%] (Warmup)
Chain 2: Iteration: 400 / 2000 [ 20%] (Warmup)
Chain 2: Iteration: 600 / 2000 [ 30%] (Warmup)
Chain 2: Iteration: 800 / 2000 [ 40%] (Warmup)
Chain 2: Iteration: 1000 / 2000 [ 50%] (Warmup)
Chain 2: Iteration: 1001 / 2000 [ 50%] (Sampling)
Chain 2: Iteration: 1200 / 2000 [ 60%] (Sampling)
Chain 2: Iteration: 1400 / 2000 [ 70%] (Sampling)
Chain 2: Iteration: 1600 / 2000 [ 80%] (Sampling)
Chain 2: Iteration: 1800 / 2000 [ 90%] (Sampling)
Chain 2: Iteration: 2000 / 2000 [100%] (Sampling)
Chain 2:
Chain 2: Elapsed Time: 0.010003 seconds (Warm-up)
Chain 2: 0.008384 seconds (Sampling)
Chain 2: 0.018387 seconds (Total)
Chain 2:

```

## Resultados

```
print(fit_dist, pars=param)
```

```

Inference for Stan model: dist_poisson_gamma.
2 chains, each with iter=2000; warmup=1000; thin=2;
post-warmup draws per chain=500, total post-warmup draws=1000.

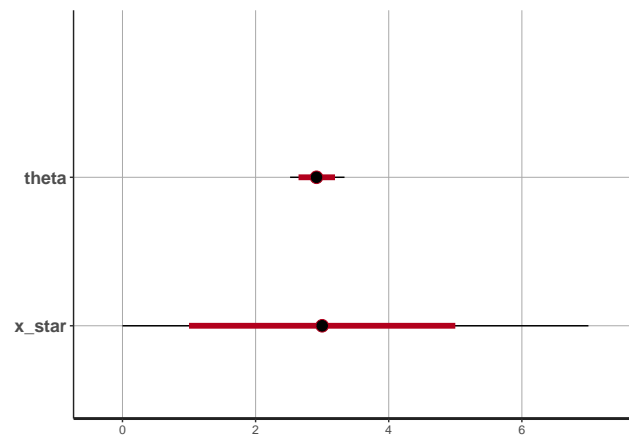
```

|       | mean | se_mean | sd   | 2.5% | 25%  | 50%  | 75%  | 97.5% | n_eff | Rhat |
|-------|------|---------|------|------|------|------|------|-------|-------|------|
| theta | 2.92 | 0.01    | 0.22 | 2.51 | 2.77 | 2.91 | 3.05 | 3.34  | 630   | 1    |

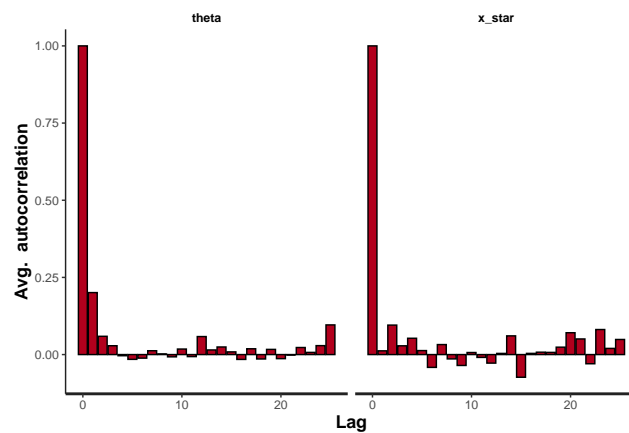
```
x_star 2.83    0.07 1.72 0.00 2.00 3.00 4.00  7.00   697    1
```

Samples were drawn using NUTS(diag\_e) at Tue Mar 14 18:47:04 2023.  
For each parameter, n\_eff is a crude measure of effective sample size,  
and Rhat is the potential scale reduction factor on split chains (at  
convergence, Rhat=1).

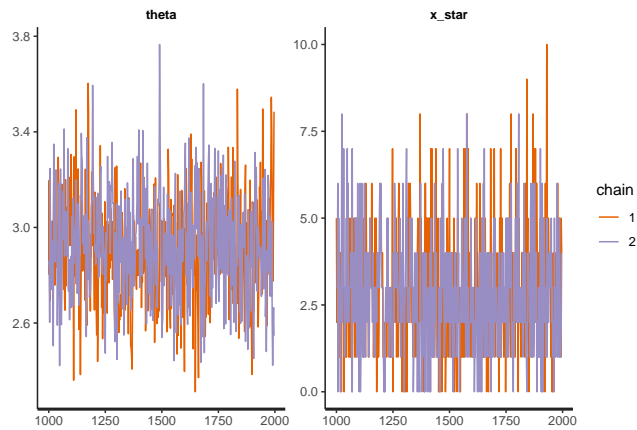
```
stan_plot(fit_dist,pars=param)
```



```
stan_ac(fit_dist,pars=param)
```

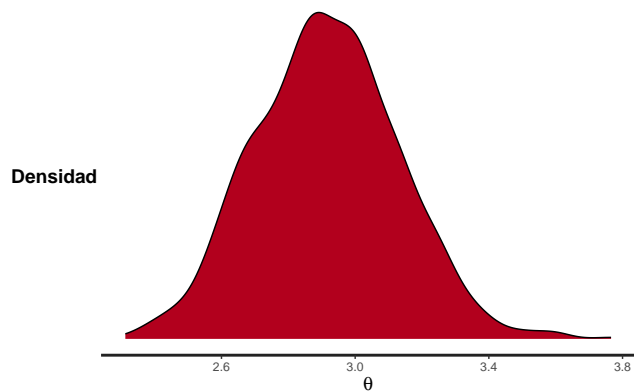


```
stan_trace(fit_dist,pars=param)
```



```
stan_dens(fit_dist, pars="theta", point_est = "mean", show_density = TRUE) +
  ggtitle(expression(paste("Distribución final de ", theta))) +
  ylab("Densidad") + xlab(expression(theta)) +
  theme(axis.title.x=element_text(size=14), axis.title.y=element_text(size=14),
        plot.title = element_text(size=16))
```

Distribución final de  $\theta$



```
stan_hist(fit_dist, pars="x_star", point_est = "mean", show_density = TRUE) +
  ggtitle(expression(paste("Distribución predictiva final de ", x^{star}))) +
  ylab("Densidad") + xlab(expression(x^{star})) +
  theme(axis.title.x=element_text(size=14), axis.title.y=element_text(size=14),
        plot.title = element_text(size=16))
```

Distribución predictiva final de  $x^{\text{star}}$

