

MCMC - Stats and jAGS

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<https://harmony-net.eu/>

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Markov chain

- ▶ Definition: A random process that undergoes transitions from one state to another on a state space

<https://setosa.io/blog/2014/07/26/markov-chains/>

Markov chain

S S S R R R S S S S R R R R R R R R R R R R S S S S S S S S

One way to simulate this weather would be to just say "Half of the days are rainy. Therefore, every day in our simulation will have a fifty percent chance of rain." This rule would generate the following sequence in simulation:

S R R S R S R R S R S S S R S R S S R S S R S S R S S S R R S S S R

Markov chain

- ▶ Markov property: A Markov chain possesses a property that is characterised as “memoryless”: the probability of the next state depends only on the current state and not on the sequence of events that preceded it.

- ▶ Markov property defined as:

$$\begin{aligned}\Pr(X_{n+1} = x | X_1 = x_1, X_2 = x_2, \dots, X_n = x_n) \\ = \Pr(X_{n+1} = x | X_n = x_n)\end{aligned}$$

Ergodicity

- ▶ This means that during numerous iterations, the chain will explore every point (or possible state) and will do so proportionally to its probability.
- ▶ To be considered ergodic, the Markov chain must be
 - ▶ irreducible: for every state there is a positive probability of moving to any other state
 - ▶ aperiodic: the chain must not get trapped in cycles

Reversibility: reversible chains

- ▶ For a stationary distribution π and a transition matrix P , the reversibility condition can be written as
$$\pi(x)P(x, y) = \pi(y)P(y, x), \text{ for all } (x, y) \in S$$

Link to Metropolis-Hastings and Gibbs

In his famous paper from 1953, Metropolis showed how to construct a Markov chain with stationary distribution π such that

$$\pi(x) = p_x, x \in S$$

Here the first step to obtaining the stationary distribution of a Markov chain is to prove that the probabilities of a distribution satisfy the reversibility condition.

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JAGS

- ▶ JAGS uses the BUGS language
 - ▶ This is a declarative (non-procedural) language
 - ▶ The order of statements does not matter
 - ▶ The compiler converts our model syntax into an MCMC algorithm with appropriately defined likelihood and prior
 - ▶ You can only define each variable once!!!

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 - ▶ You can only define each variable once!!!
- ▶ Different ways to run JAGS from R:
 - ▶ `rjags`, `runjags`, `R2jags`, `jagsUI`
- ▶ See <http://runjags.sourceforge.net/quickjags.html>

A simple JAGS model might look like this:

```
model_definition <- "model{  
  # Likelihood part:  
  Positives ~ dbinom(prevalence, TotalTests)  
  
  # Prior part:  
  prevalence ~ dbeta(2, 2)  
  
  # Hooks for automatic integration with R:  
  #data# Positives, TotalTests  
  #monitor# prevalence  
  #inits# prevalence  
}  
"  
cat(model_definition, file='basicjags.bug')
```

There are two model statements:

```
# Likelihood part:  
Positives ~ dbinom(prevalence, TotalTests)
```

- ▶ states that the number of *Positive* test samples is Binomially distributed with probability parameter *prevalence* and total trials *TotalTests*

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```
# Prior part:  
prevalence ~ dbeta(2,2)
```

- ▶ states that our prior probability distribution for the parameter *prevalence* is Beta(2,2)

The other lines in this model:

```
#data# Positives, TotalTests  
#monitor# prevalence  
#inits# prevalence
```

are automated hooks that are only used by runjags.

To run this model, copy/paste the code above into a new text file called “basicjags.bug” in the same folder as your current working directory. Then run:

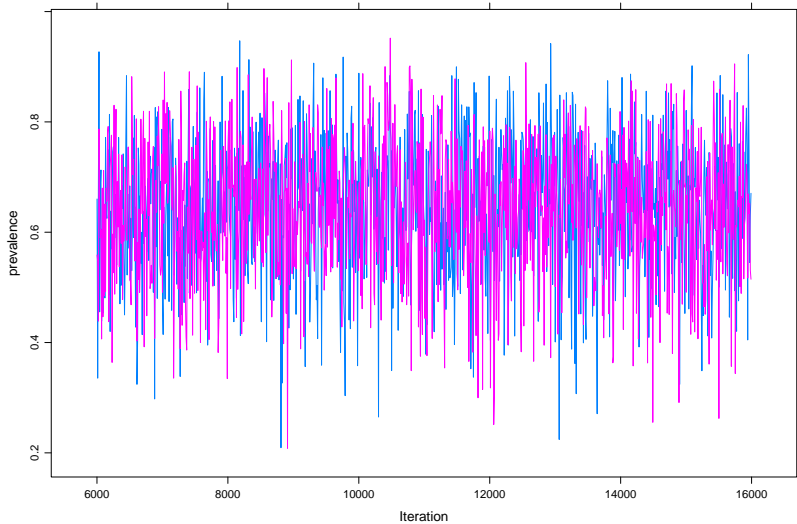
```
library('runjags')  
  
# data to be retrieved by runjags:  
Positives <- 7  
TotalTests <- 10  
  
# initial values to be retrieved by runjags:  
prevalence <- list(chain1=0.05, chain2=0.95)
```

```
results <- run.jags('basicjags.bug', n.chains=2, burnin=5000, sample=10000)
```

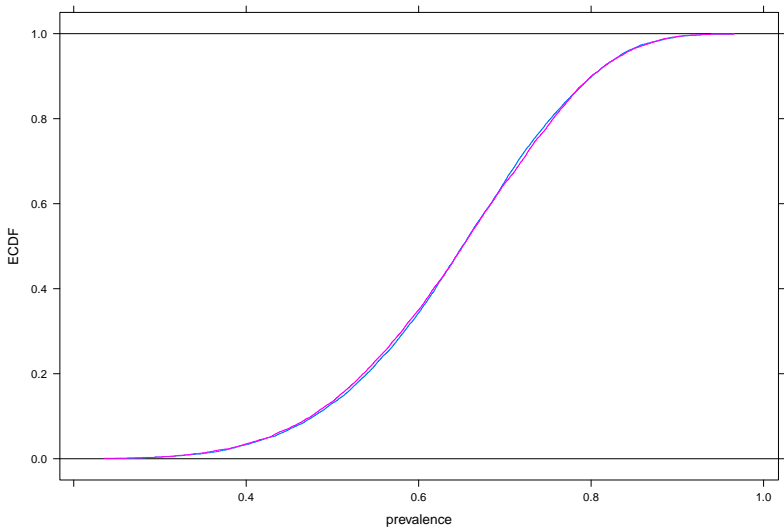
First check the plots for convergence:

```
plot(results)
```

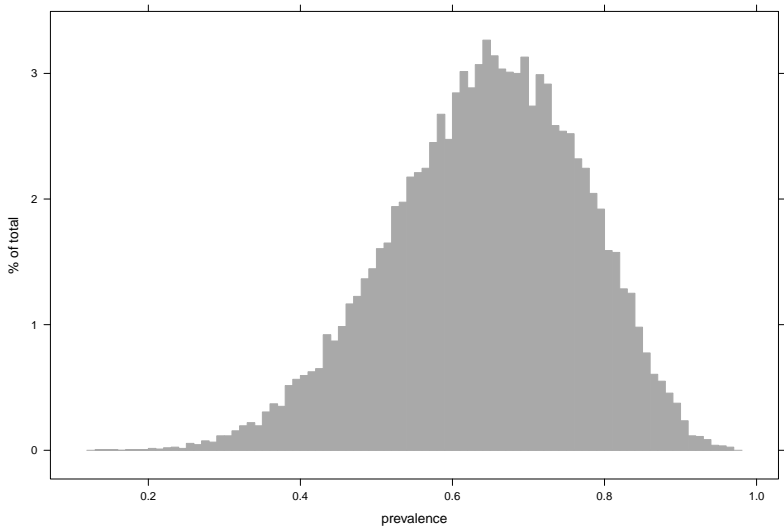

Trace plots: the two chains should be stationary:



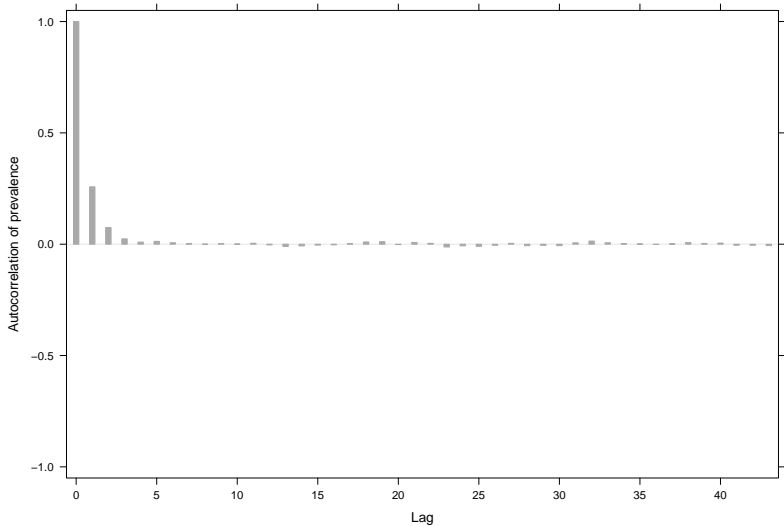
ECDF plots: the two chains should be very close to each other:



Histogram of the combined chains should appear smooth:



Autocorrelation plot tells you how well behaved the model is:



Then check the effective sample size (S_{Seff}) and Gelman-Rubin statistic (psrf):

```
results
##
## JAGS model summary statistics from 20000 samples (chains = 2; adapt+burnin =
##
##           Lower95 Median Upper95      Mean      SD Mode      MCerr MC%ofSD SSeff
## prevalence 0.40257 0.6508 0.88056 0.64373 0.12469  -- 0.0011478      0.9 1180
##
##           AC.10 psrf
## prevalence 0.00086095 1
##
## Total time taken: 0.2 seconds
```

Reminder: we want $\text{psrf} < 1.05$ and $\text{S}_{\text{Seff}} > 1000$