# Convergence diagnostics for a ill-posed model

#### Chapter 3.4: Diagnosing and improving convergence

In this example the chains **do not converge** and we show how the convergence diagnostics flag non-convergence. The model is contrived to give poor convergence. It is \[Y\sim\mbox{Poisson}(\exp[\mu\_1+\mu\_2]) \mbox{ where }\mu\_1,\mu\_2\sim\mbox{Normal}(0,1000).\] This is a silly model because there is only one observation, \(Y=1\), and two parameters. Further, the two parameter give the same likelihood for all combinations of \(\mu\_1\) and \(\mu\_2\) that give the same sum.

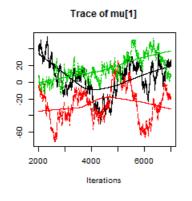
### Define the model as a string

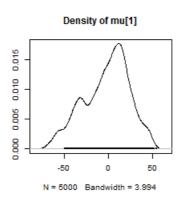
#### Generate posterior samples

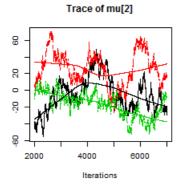
#### **Graphical diagnostics**

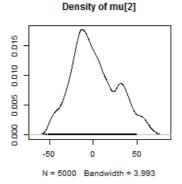
The trace plots are meandering, the chains (each chain is a different color) give different estimates, and the autocorrelation is high even for lag 35. All of these indicate poor convergence.

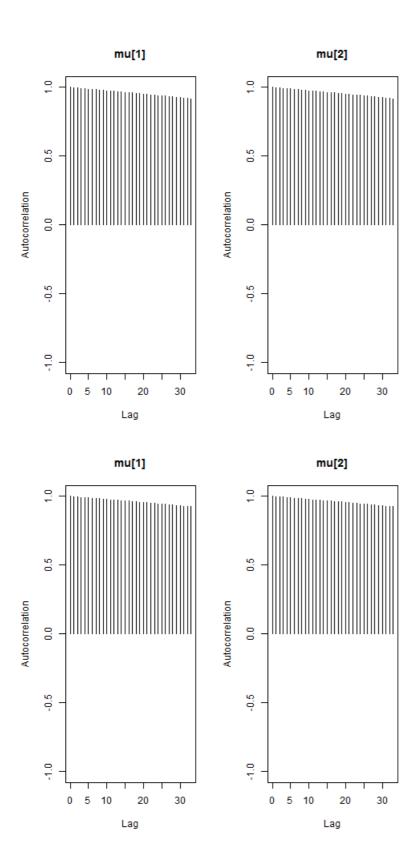
```
plot(samples)
```

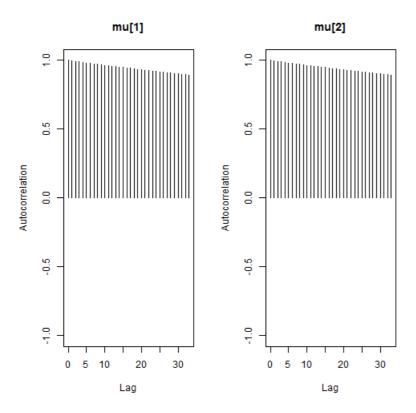












## **Numerical diagnostics**

## 1.8

# |z| greater than 2 indicates poor convergence

geweke.diag(samples[[1]])

```
# Autocorrelation near 1 indicates poor convergence
  \verb"autocorr(samples[[1]],lag=1")"
\mbox{\tt ##} , , \mbox{\tt mu[1]}
##
##
             mu[1]
## Lag 1 0.9969476 -0.9957215
##
## , , mu[2]
##
              mu[1]
                         mu[2]
## Lag 1 -0.9975801 0.9970195
  # Low ESS indicates poor convergence
  effectiveSize(samples)
      mu[1]
                mu[2]
## 21.18113 20.71919
 # R greater than 1.1 indicates poor convergence
  gelman.diag(samples)
## Potential scale reduction factors:
##
         Point est. Upper C.I.
## mu[1]
                2.01
                            3.5
                             3.5
## mu[2]
                2.01
## Multivariate psrf
##
```

```
##
## Fraction in 1st window = 0.1
## Fraction in 2nd window = 0.5
##
## mu[1] mu[2]
## 3.239 -3.347
```