

第二届的R语言会议 2009年12月12-13日, 华东师范大学



R与WinBUGS

---贝叶斯统计分析

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Outline

1. Statistics and Statistical Inference
2. Statistical Softwares (packages or toolboxes)
3. What is **R** and why **R**?
4. Bayesian Inference and **BUGS**
5. **WinBUGS** and related packages
6. A start example using **DoodleBUGS**
7. Case Study --- Gamma-Poisson hierarchical model
8. **R2WinBUGS**: **WinBUGS** run in **R**
9. Summary
10. References

Statistics and Statistical Inference



- **Statistics** is the science that relates **data** to specific questions of interest. This includes devising
 - methods to gather data relevant to the question,
 - methods to summarize and display the data to shed light on the question,
 - methods that enable us to draw answers to the question that are supported by the data.
- Different from other disciplines:
data almost always contain uncertainty.
- **Statistics is a science about uncertainty**



- **Statistical inference** gives us **methods** and **tools** for data analysis.
 - there is a **probability model** explaining how the uncertainty gets into the data.
 - **data** are generated in accordance with some unknown probability distribution
 - By analyzing the data, we attempt
 - to learn about the unknown distribution,
 - To make some inferences about certain properties of the distribution, and
 - to determine the relative likelihood that each possible distribution is actually the correct one.



Statistical Softwares

- Some well-known statistics/math softwares
 - SAS
 - SPSS
 - STATA
 - Minitab
 - Matlab (a toolbox)
 - S-Plus
- All of them are **NOT** free



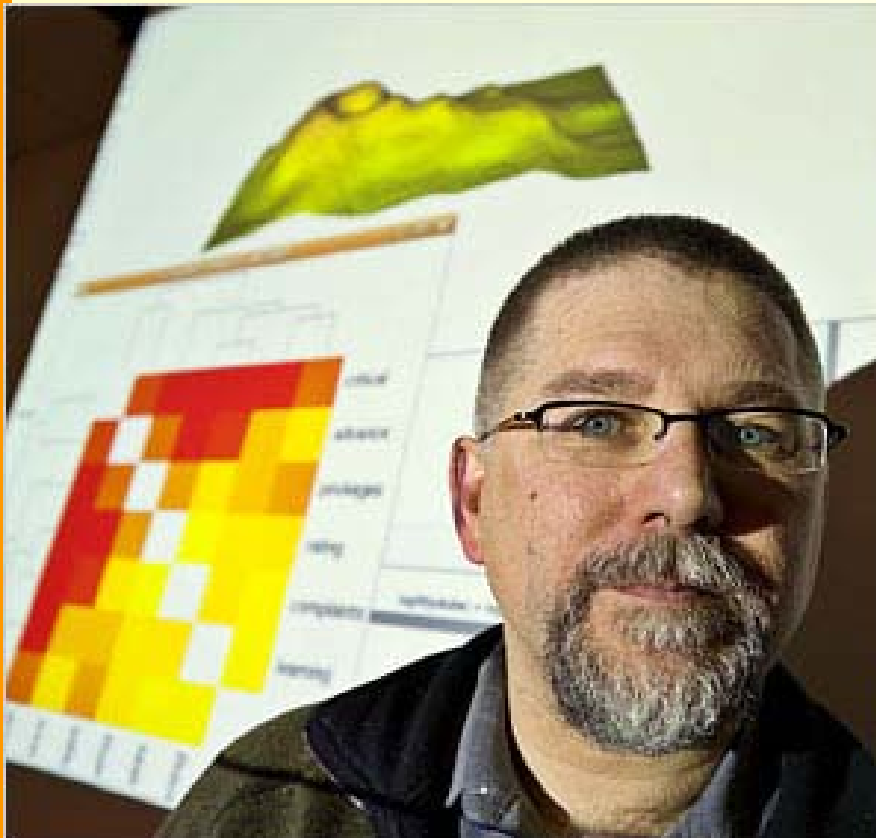
- Free statistics/math softwares
 - Maxima vs. maple/mathematica
 - Octave/scilab vs. Matlab
 - **R** vs. S-plus
- My points of view:
Free does not always mean “not safe or useful”
 - Maxima, Scilab and R are popular and reliable softwares!
 - Linux are better than Windows
 - **(La)TeX** are more popular and better than word/Power Point.



What is R and why R?

R first appeared in 1996, when the statistics professors **Robert Gentleman**, left, and **Ross Ihaka** released the code as a free software package (under GNU General Public License).

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Robert Gentleman



Ross Ihaka



- R is a **language** and **environment**
- R is an integrated suite of software facilities for data manipulation, calculation and graphical display. It includes
 - an effective data handling and storage facility,
 - a suite of operators for calculations on arrays/ matrices,
 - a large, coherent, integrated collection of intermediate **tools for data analysis**,
 - **graphical facilities** for data analysis and display
 - a well-developed, **simple and effective programming language** which includes conditionals, loops, user-defined recursive functions and input and output facilities.



- Why R?

- Data Analysts Captivated by R’s Power (Ashlee Vance, The New York Times, **Jan 6, 2009**) give the answer:
 - R is an **open-source program**, and its popularity reflects a shift in the type of software used inside corporations.
 - ... statisticians, engineers and scientists without computer programming skills find it **easy to use**
 - They can improve the software’s code or write variations for specific tasks.
 - “The **great beauty of R** is that you can modify it to do all sorts of things,” said Hal Varian, chief economist at Google.



- Now R is **surpassing** what Mr. Chambers had imagined possible with **S**.
- “R has really become the **second language** for people coming out of grad school now, and there’s an amazing amount of code being written for it,” said Max Kuhn, associate director of nonclinical statistics at Pfizer.
- while SAS plays down R’s corporate appeal, companies like Google and Pfizer say they **use the software for just about anything they can**
- “R is a real demonstration of the power of collaboration, and I don’t think you could construct something like this any other way,” Mr. Ihaka said. “We could have chosen to be commercial, and we would have sold five copies of the software.”



Bayesian Inference/Modeling

- Main Approaches to Statistics
 - Frequentist/classical approach, which is based on:
 - **Parameters**, the numerical characteristics of the population, are **fixed but unknown** constants.
 - Probabilities are always interpreted as long run relative frequency.
 - Statistical procedures are judged by how well they perform in the long run over an infinite number of hypothetical repetitions of the experiment.



- Bayesian approach, the ideas of which are:
 - **Parameters** are considered as **random**.
 - Probability statements about parameters must be interpreted as "**degree of belief**." The prior distribution must be subjective.
 - We revise our beliefs about parameters after getting the data by using **Bayes' theorem**.
 - The posterior distribution comes from two sources: the prior distribution and the observed data.
- The Bayesian approach forms the basis of science -- **learning process**: As we get more data, we add to our store of information by multiplying it by our current posterior distribution.



Notations:

$D = \{y_1, y_2, \dots, y_n\}$: data from sampling distribution $f(x | \theta)$

$L(\theta | D) = \prod_{i=1}^n f(y_i | \theta)$: Likelihood---sampling information

θ : parameter of interest

$\pi(\theta)$: prior distribution of θ --- prior information

$\pi(\theta | D) \propto \pi(\theta)L(\theta | D)$: posterior distribution



Prior info+sampling info=Posterior info

Bayes Theorem:

$$\pi(\theta | x) = \frac{\pi(\theta) f(x | \theta)}{\int \pi(\theta) f(x | \theta) d\theta} \propto \pi(\theta) f(x | \theta)$$

Formulation:

$$\begin{cases} \theta \sim \pi(\theta) \\ y_1, y_2, \dots, y_n \text{ iid} \sim f(y | \theta) \end{cases} \Rightarrow \theta | x \sim \pi(\theta | x)$$

Bayesian Model Specification:

- 1) prior distribution (often conjugate)
- 2) sampling distribution



Examples:

1) Beta-Binomial model for proportion:

$$\begin{cases} \theta \sim \text{Beta}(\alpha, \beta) \\ y | \theta \sim \text{Bin}(n, \theta) \end{cases} \Rightarrow \theta | y \sim \text{Beta}(\alpha + y, n - y + \beta)$$

2) Normal-normal model for mean (given variance)

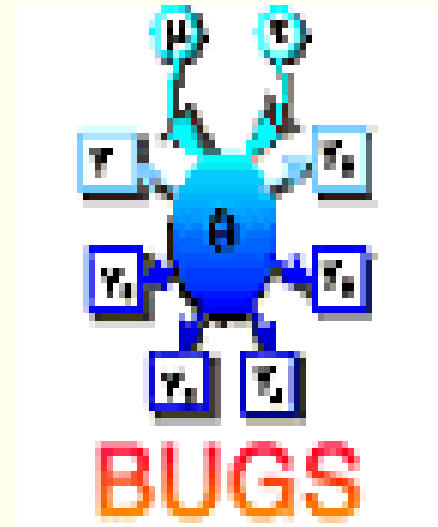
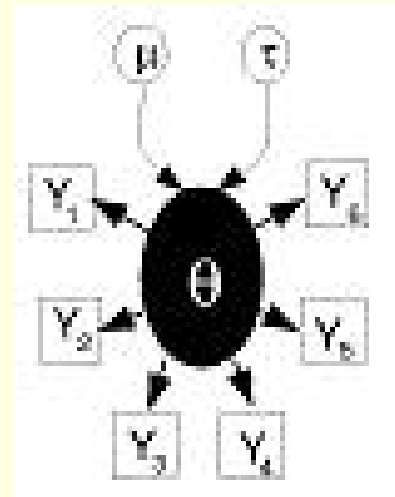
$$\begin{cases} \theta \sim N(\mu, \tau^2) \\ y_1, \dots, y_n | \theta \sim N(\theta, \sigma^2) \end{cases} \Rightarrow \theta | y \sim N(\mu_n, \sigma_n^2), \text{ where}$$

$$\mu_n = \frac{\frac{1}{\sigma^2/n} \bar{y} + \frac{1}{\tau^2} \mu}{\frac{1}{\sigma^2/n} + \frac{1}{\tau^2}}, \quad \frac{1}{\sigma_n^2} = \frac{1}{\sigma^2/n} + \frac{1}{\tau^2}$$



What is BUGS/WinBUGS?

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Logo of BUGS

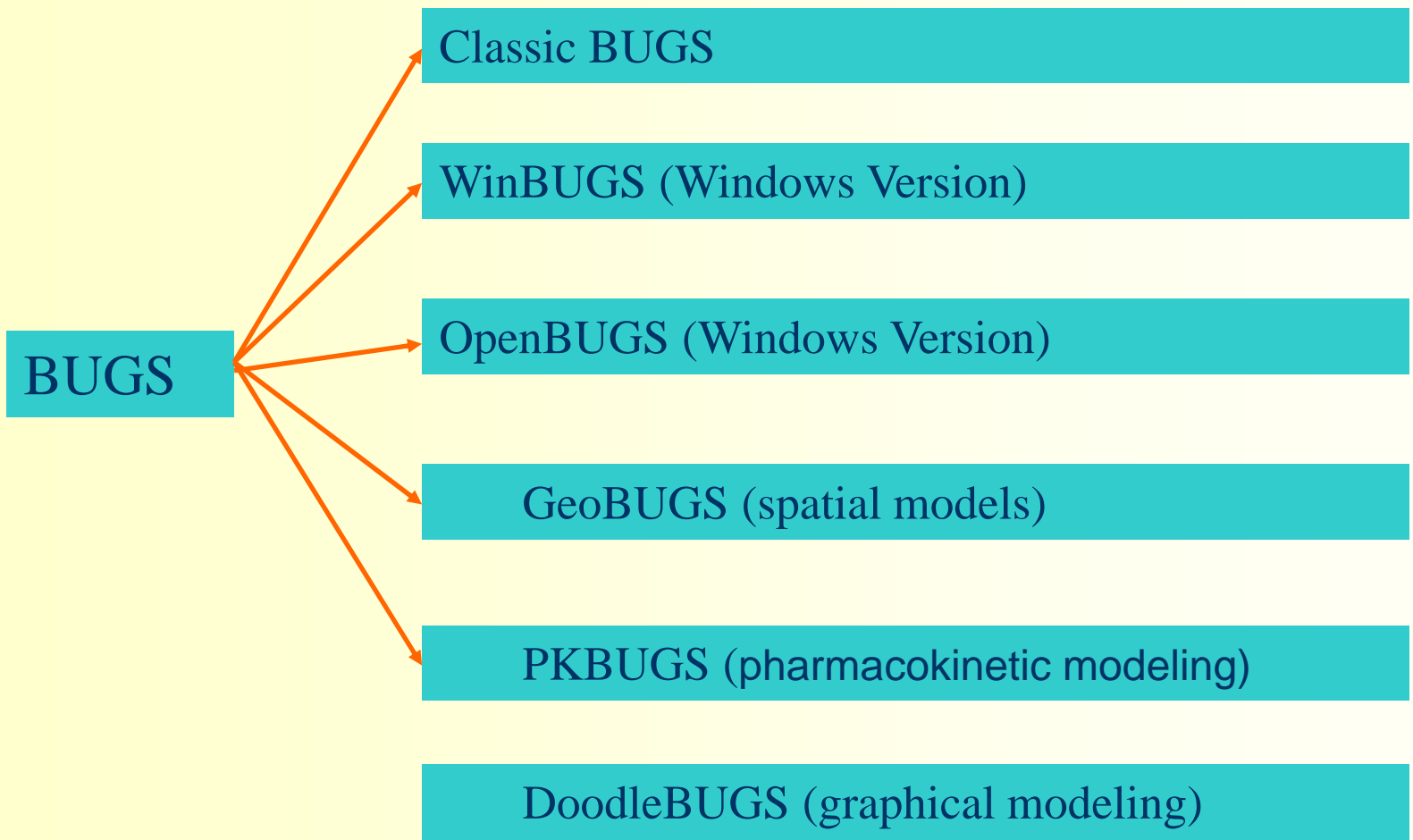


What is BUGS?

- BUGS (**B**ayesian inference **U**sing **G**ibbs **S**ampling) is a popular software for analyzing complex statistical models using Markov Chain Monte Carlo (MCMC) methods
- It uses
 - Gibbs sampling
 - Metropolis algorithm to generate a Markov chain by sampling from full conditional distributions.



- **History** (<http://www.mrc-bsu.cam.ac.uk/bugs/>)
 - a command-line interface. Not further developed since 1996.
 - Originate at the MRC BIostatistical Unit in Cambridge
 - Later developed into **WinBUGS**, jointly with the Imperial College School of Medicine at St Mary's, London.
 - **OpenBUGS**, completely revised version of WinBUGS, developed in the University of Helsinki, Finland.
 - Add-on or interfaces: **GeoBUGS, PKBUGS,.....**





What is WinBUGS?

- **WinBUGS**, a windows program with an option of a graphical user interface, the standard 'point-and-click' windows interface, and on-line monitoring and convergence diagnostics. It also supports Batch-mode running (version 1.4).
- **GeoBUGS**, an add-on to **WinBUGS** that fits spatial models and produces a range of maps as output.
- **PKBUGS**, an efficient and user-friendly interface for specifying complex population pharmacokinetic pharmacokinetic and pharmacodynamic (PK/PD) models within **WinBUGS** software.



Why use WinBUGS?

- Its **ability** to fit complex statistical models using MCMC methods.
- Its **flexibility** to program, two different ways to specify model
 - DoodleBUGS: *Direct graphics*
 - *BUGS language*
- **Free** to download:
<http://www.mrc-bsu.cam.ac.uk/bugs>.
- A lot of **examples** in WinBUGS
- A lot of **online resources**
<http://www.mrc-bsu.cam.ac.uk/bugs/weblinks/webresource.shtml>

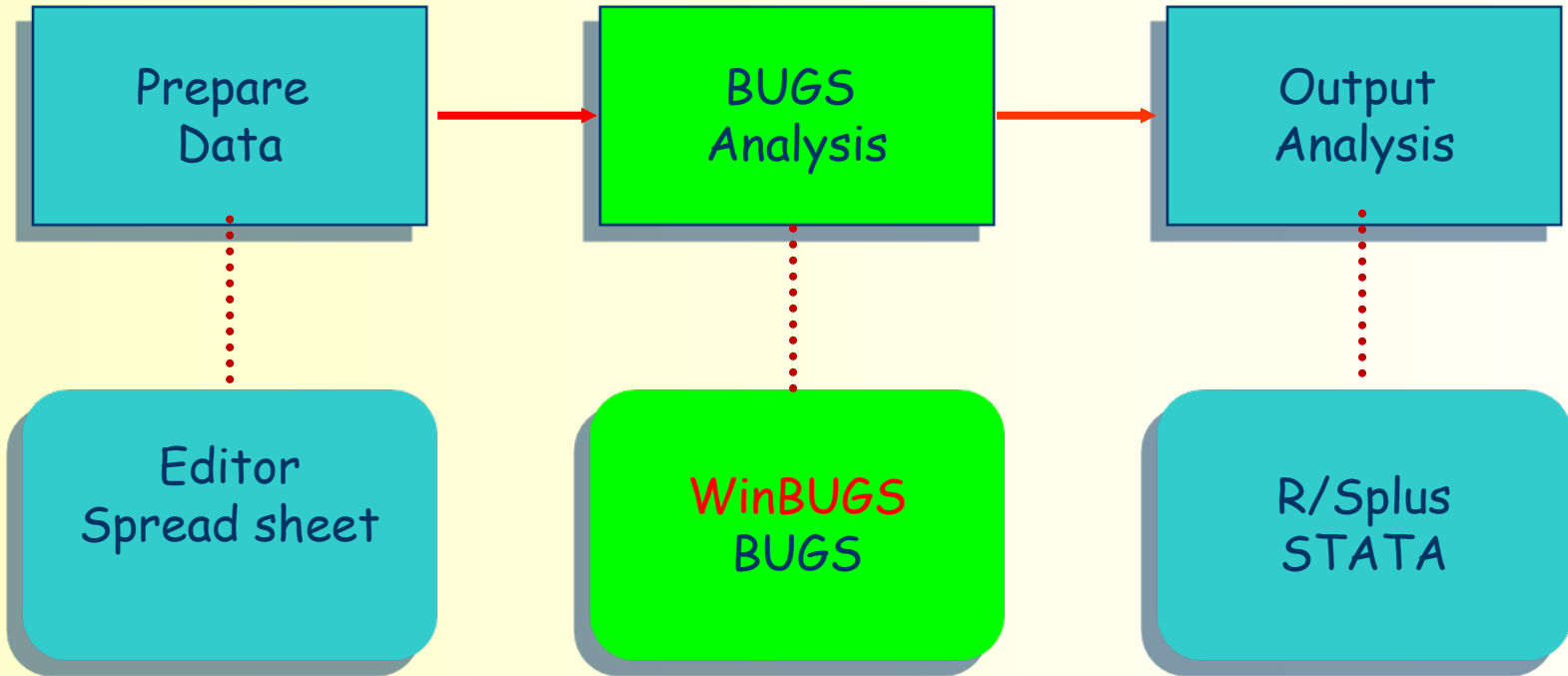


Related packages

- Package needs to be downloaded
 - WinBUGS14.exe
- Potential useful packages
 - **CODA** (Convergence Diagnostic and Output Analysis)
 - **BOA** (Bayesian Output Analysis)
 - **JAGS** (Just Another Gibbs Sampler)---for analysis of Bayesian hierarchical models.
 - **R2WinBUGS**---A package for Running WinBUGS from R



Working with BUGS





How to use WinBUGS

- **Step 1:** Specify the model to run
- **Step 2:** Load the data and initial values for a specified number of chains
- **Step 3:** Run WinBugs to get the Markov chains and save the results for the parameters for later use.



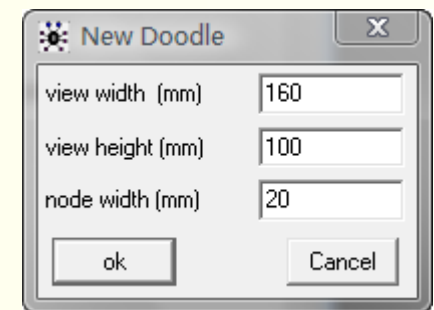
A start example via DoodleBUGS

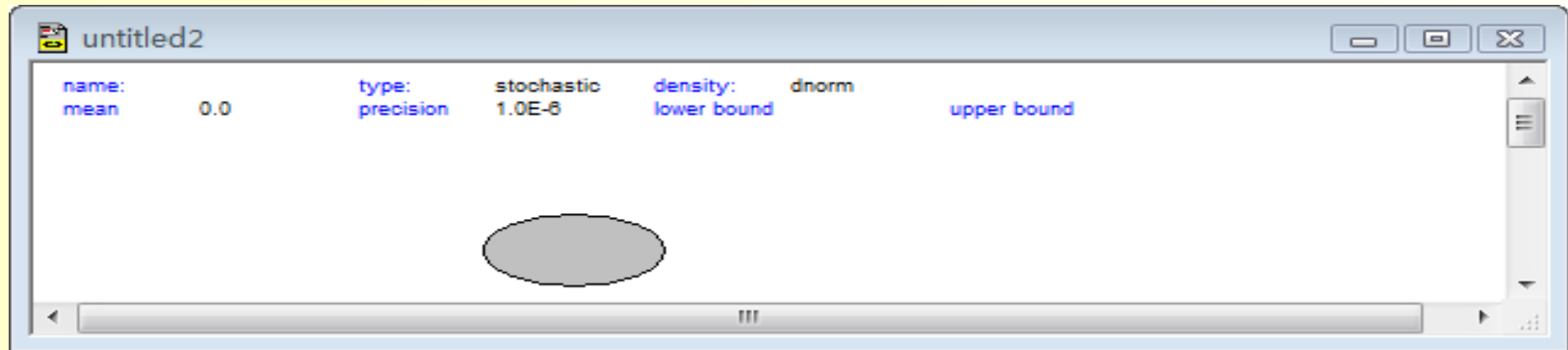
- unemployed proportion p in the population?
 - p is in $[0, 1]$, where 0 means no one is unemployed and 1 means everyone is.
 - prior information of p (from newspaper reports, economic theory, previous surveys, etc.) can be used.
 - choose a beta prior distribution restricted between 0.1 and 0.6.
 - Data: $n = 14$ people were surveyed and $r = 4$ of them were unemployed.



Model Specification in WinBUGS

- Method 1: directly write a WinBUGS document(in BUGS language)
- Method 2: Draw a directed graph in DoodleBUGS and then transfer into a WinBUGS document.
 - Start WinBUGS
 - Select “**Doodle**” from menu bar
 - Select “**New...**”
 - Press “**ok**”
 - You have a window to “Doodle” in.
 - Creating a node
 - **Mouse click** in Doodle Window





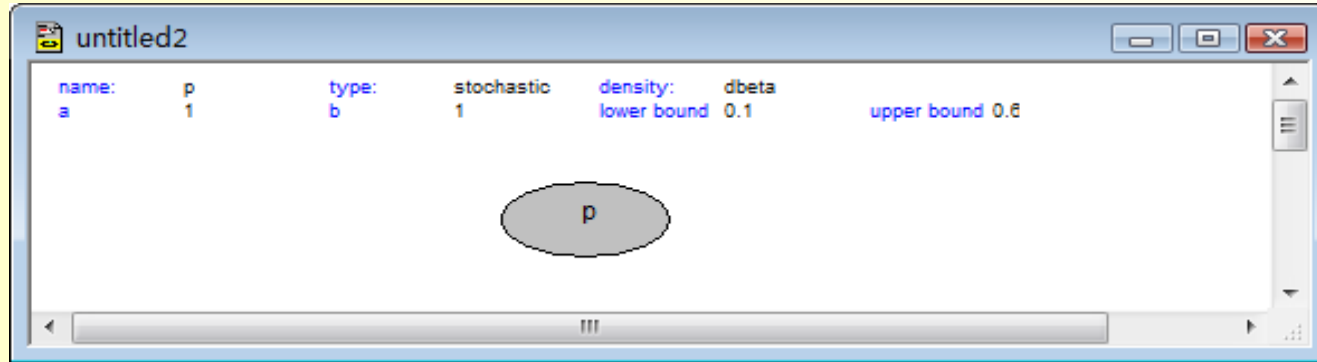
- A node has a name and type along with other characteristics and parameters depending on its type.
- Delete a node:
 - ⌘ highlight it
 - ⌘ **CTRL + Del**



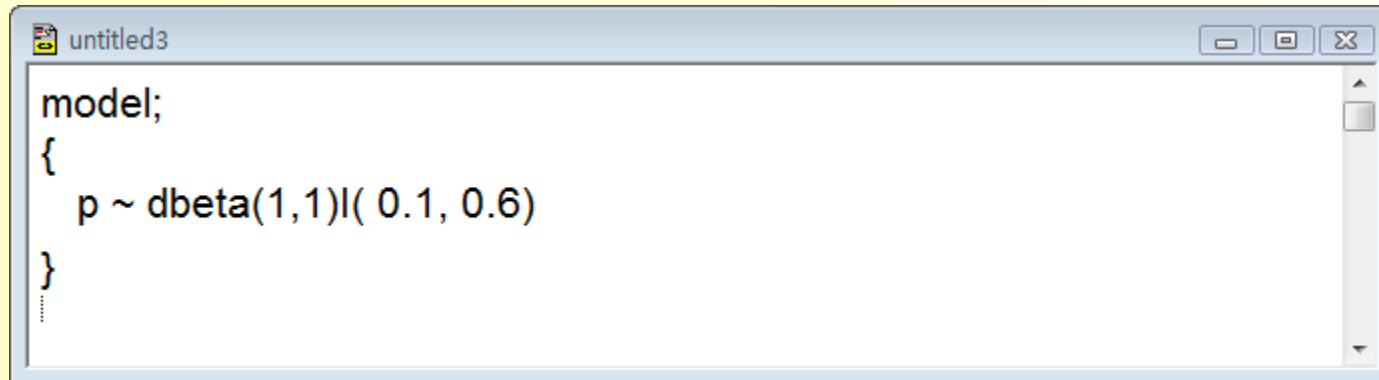
- Node Types
 - Stochastic
 - Logical
 - Constant (rectangle)
- Create node **p**, click on and type/choose
 - **Name:** p
 - **Type:** stochastic
 - **Density:** dbeta
 - **A:**1; **b:**1 (beta parameters)
 - **lower bound:**0.1; **upper bound:** 0.6



WinBUGS document



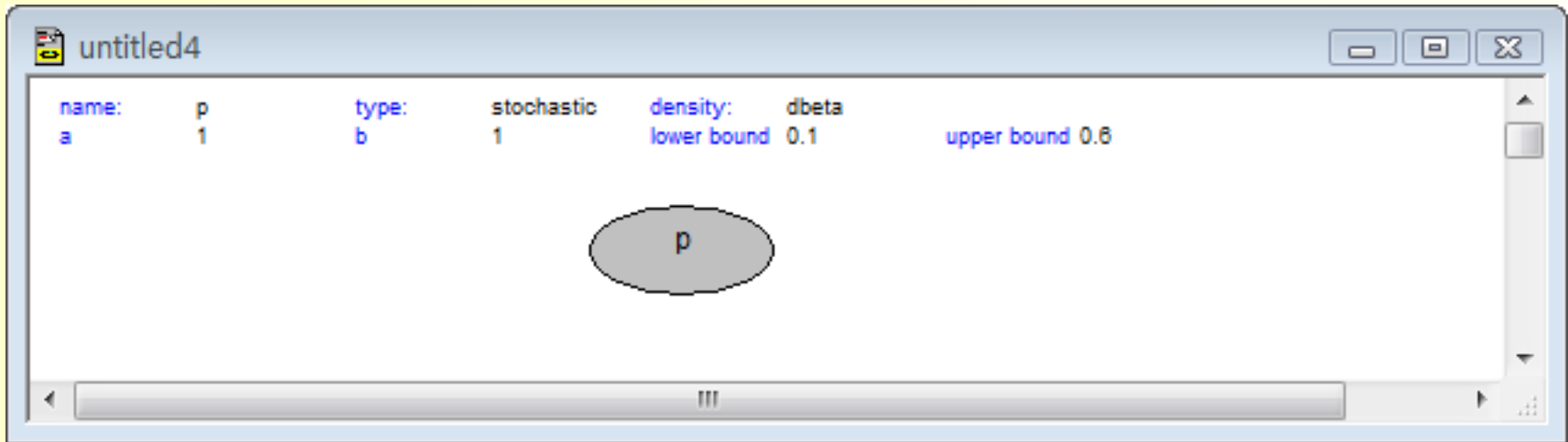
- Two ways to transfer into a WinBUGS document
 1. Menu bar > Doodle > Write Code.
 - highlight the code > Click on Attributes > 16 point.
 - ~ is read “distributed as”





2. Select/ Copy/Paste your doodle into WinBUGS new document (still a doodle!)

- Menu bar > File > New
- Menu bar > Edit > Copy
- Menu bar > Edit > Paste

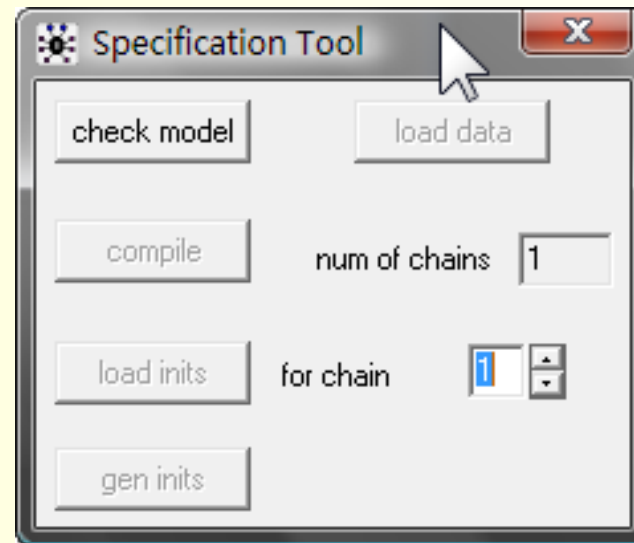
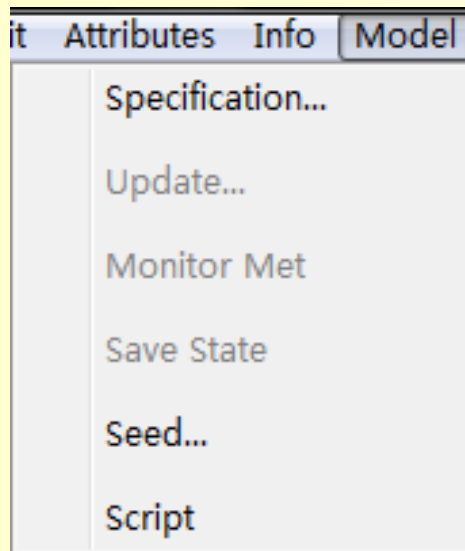


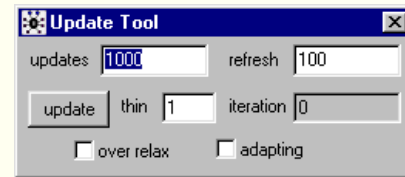
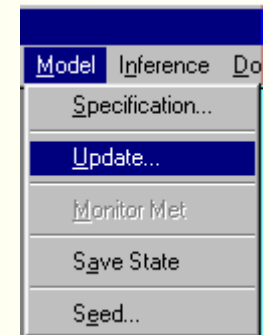
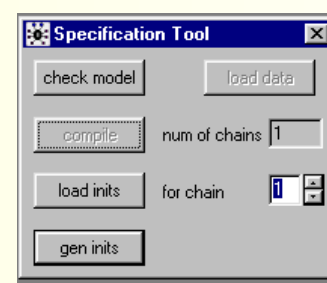
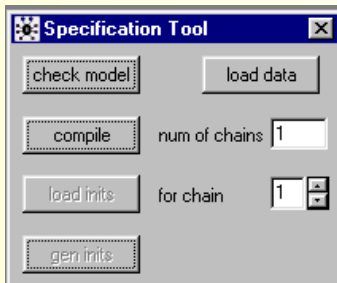
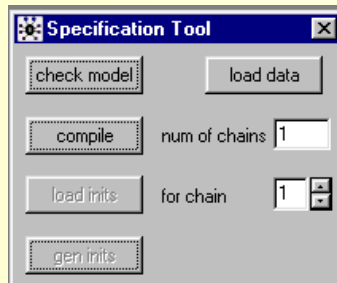
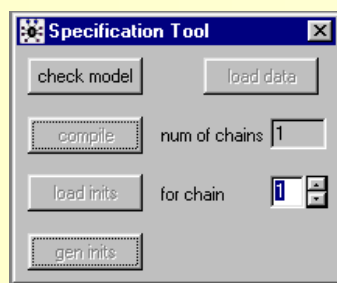
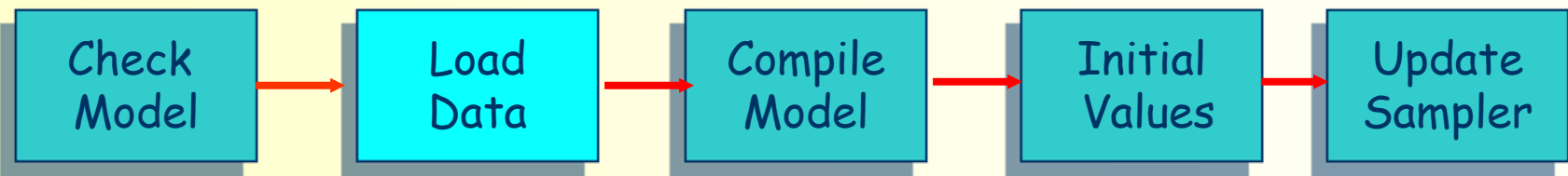
Simplest doodle—a directed graph



Running BUGS

- Menu bar > model > Specification...





Checks Syntax

Start Sampler



- use WinBUGS to look at samples from our prior
- Check Model
 - Select the Doodle (note the hairy boarder)
 - Menu bar - Model - Check model
 - Note the message in bottom left hand corner

```
model is syntactically correct
```

- Compiling the Model
 - Menu bar - Model - Compile
 - Bottom left hand corner

```
model compiled
```

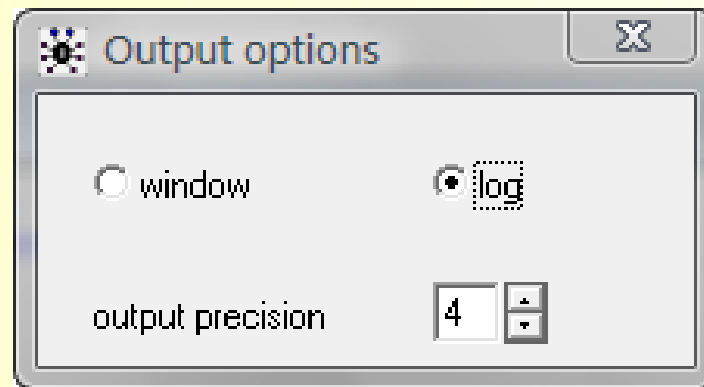


- Load Initial Values

- Menu bar - Model - Gen inits
- Bottom left hand side

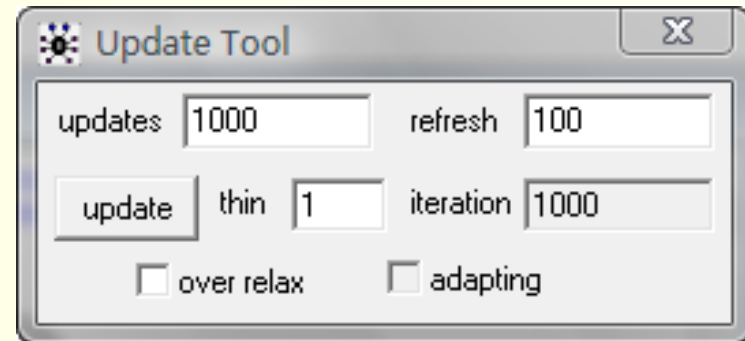
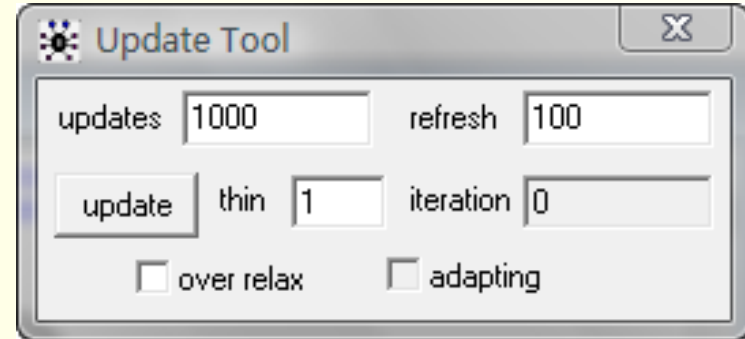
initial values generated

- Before produce result, select Options > Output options > Select the log radio button.





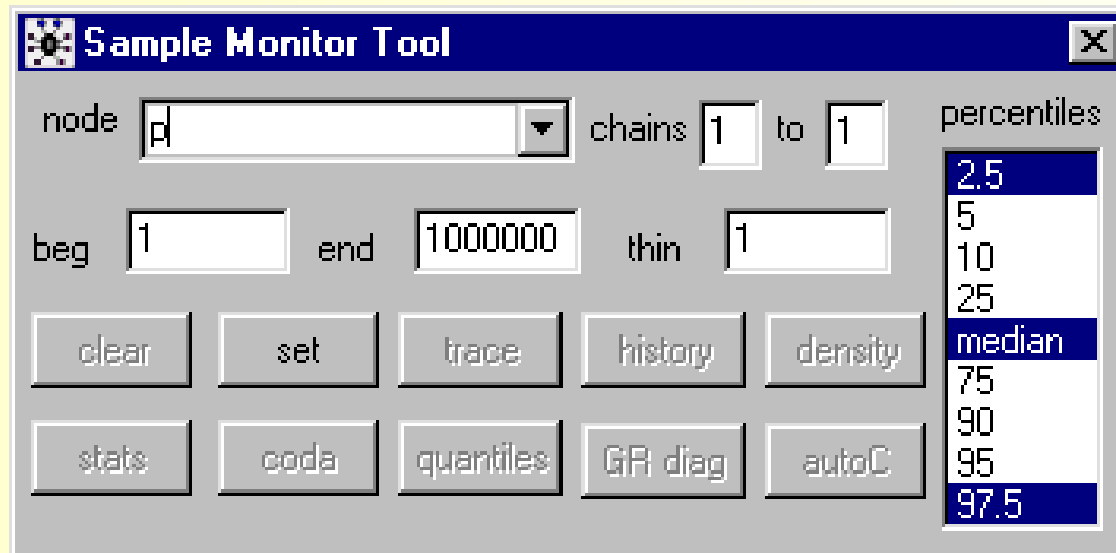
- Update the Model
 - Menu bar - Model – Update
 - 1000 MCMC updates to be carried out.
 - Note:
 - Model has been updated
 - MCMC run did not store any data---useful for “burn-in”
 - Store values by “monitoring” them to
 - Draw inferences
 - Monitor MCMC run





Monitoring Nodes

- Monitoring p , our parameter of interest
- Menu bar - Inference - Samples...
- Type name of node “ p ” to monitor
- Press “set”

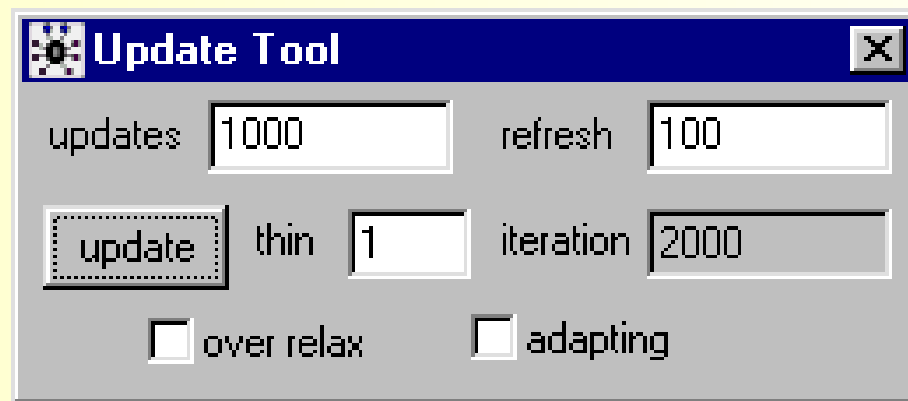


Sample Monitor Tool



● Update & Monitor

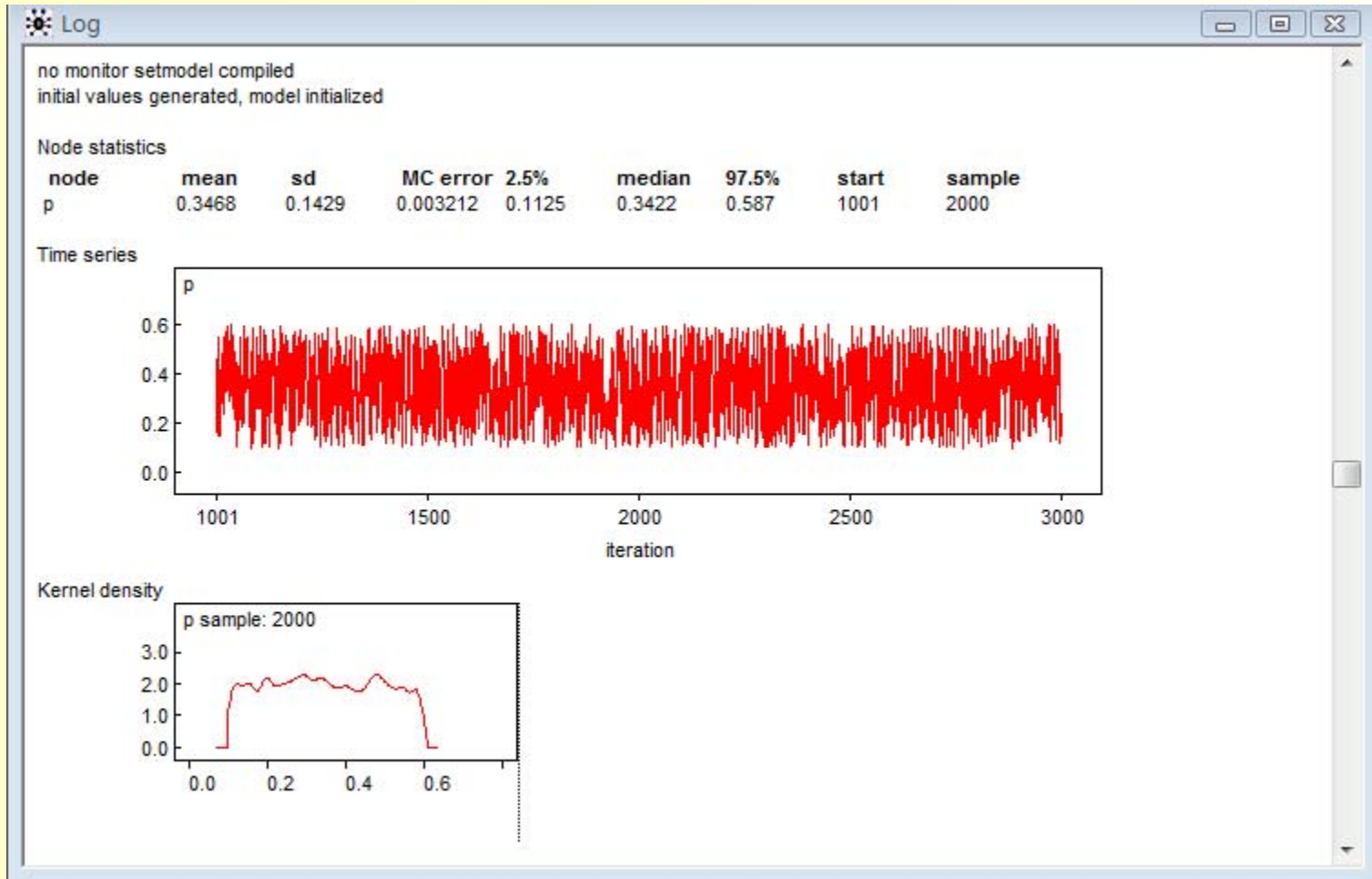
- **Update** model again
- 1000 values “monitored” of the MCMC run for p





WinBUGS output--1

- Summary Statistics
 - Select “p” from the Sample Monitor Tool
 - Press “stats” (Sample Monitor Tool)
- MCMC Time Series
 - Press “History” in Sample Monitor Tool
- Kernel Density
 - Press “Density” in the Sample Monitor Tool



Output results in log file



Inference: combining prior with data

- For unemployment example, $n = 14$ people were surveyed and $r = 4$ of them were unemployed.
- Bayesian Model: Beta – Binomial

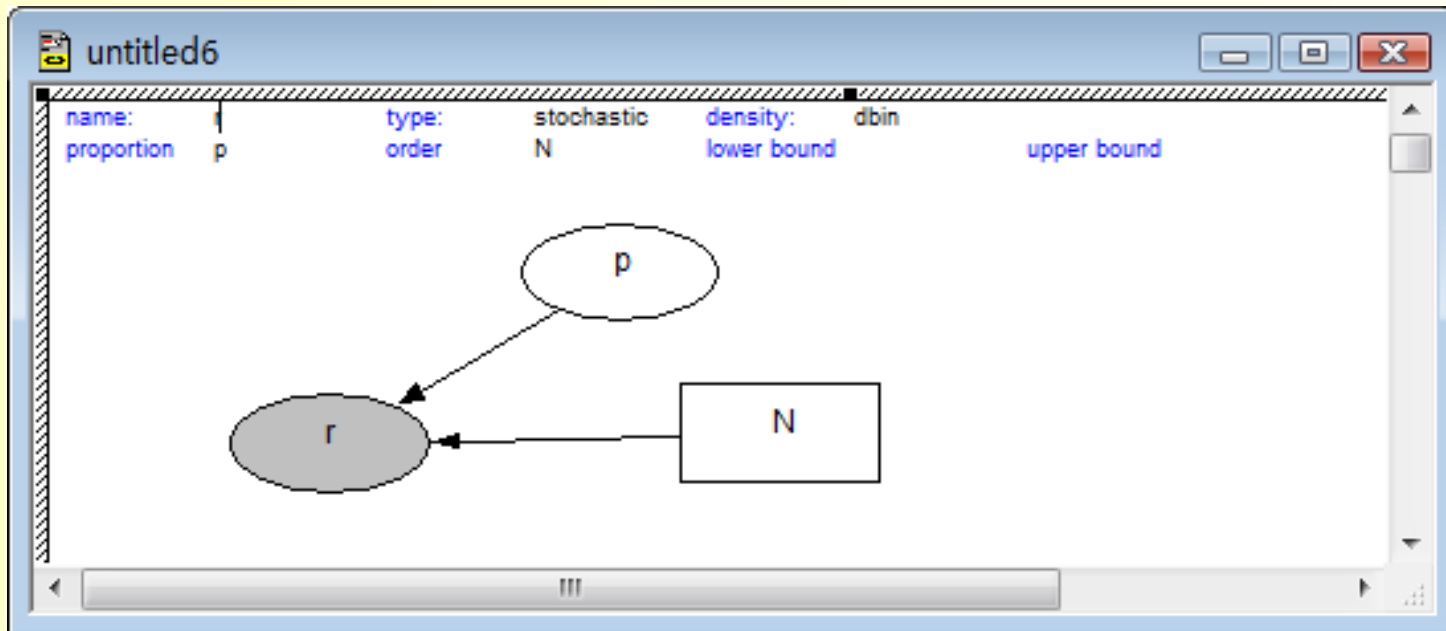
$$\begin{cases} p \sim \text{Beta}(a, b) \\ r | p \sim \text{Bin}(N, p) \end{cases}$$

- $N=14$ (order), $a=1$, $b=1$. Change p to be restricted between 0.2 and 0.45 .



- New doodle/directed graph
 - Create a stochastic node p in $(0.2, 0.45)$ (done!)
 - Add a node N
 - name: N
 - type: `constant`
 - Add a node r
 - Name: r
 - type: `stochastic`
 - density: `dbin`(binomial), with proportion p and order N .
 - add links between the nodes :
 - click on node r (highlighted) .,
 - With the Ctrl key held, click on node p and node N .

- Two relationships in a noodle
 - Single arrows: stochastic dependencies/
statistical relationship (is distributed as)
 - hollow arrows: **logical relationship**



Beta-binomial doodle/directed graph



- To see the code
 - Select Doodle > Write Code
- To enter the data, in the model code window
list(N=14,r=4)

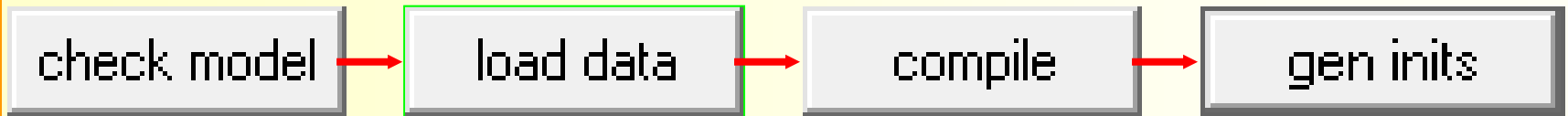
A screenshot of a code editor window titled "untitled8". The window contains the following R code:

```
model;
{
  p ~ dbeta(1,1)|( 0.2,0.45)
  r ~ dbin(p,N)
}
list(N=14, r=4)
```



Running BUGS again!

- Select Model > Specification..., do

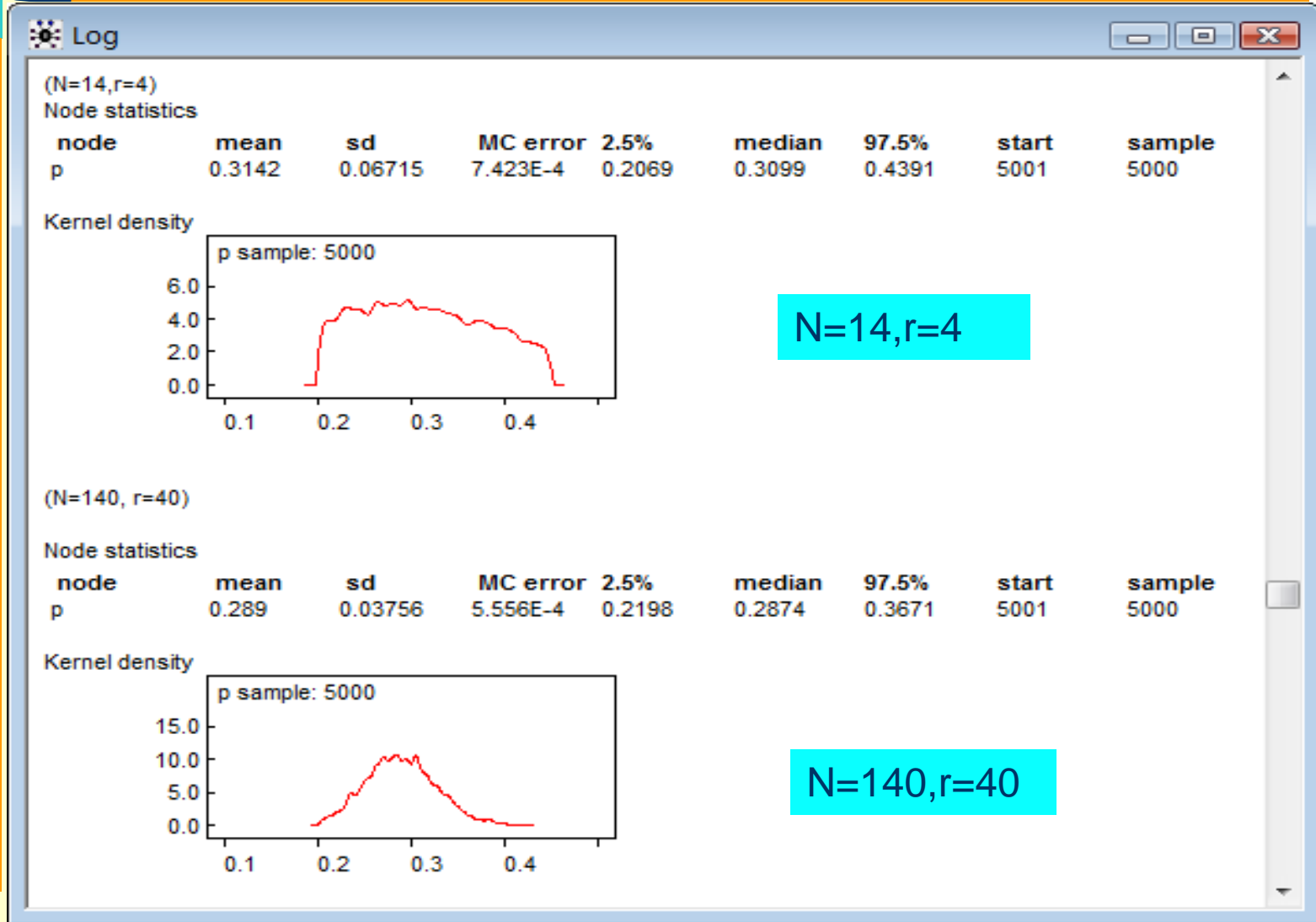


- WinBUGS calculate the appropriate likelihood function and prior for p , and combine them into a posterior distribution. (straightforward for conjugate)
- Select Model > Update..., change updates to 5000
 - WinBUGS generates updated samples of p (from the initial) by combining the prior information on p and the new information on p given by the data r and N .



WinBUGS output--2

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N=14, r=4

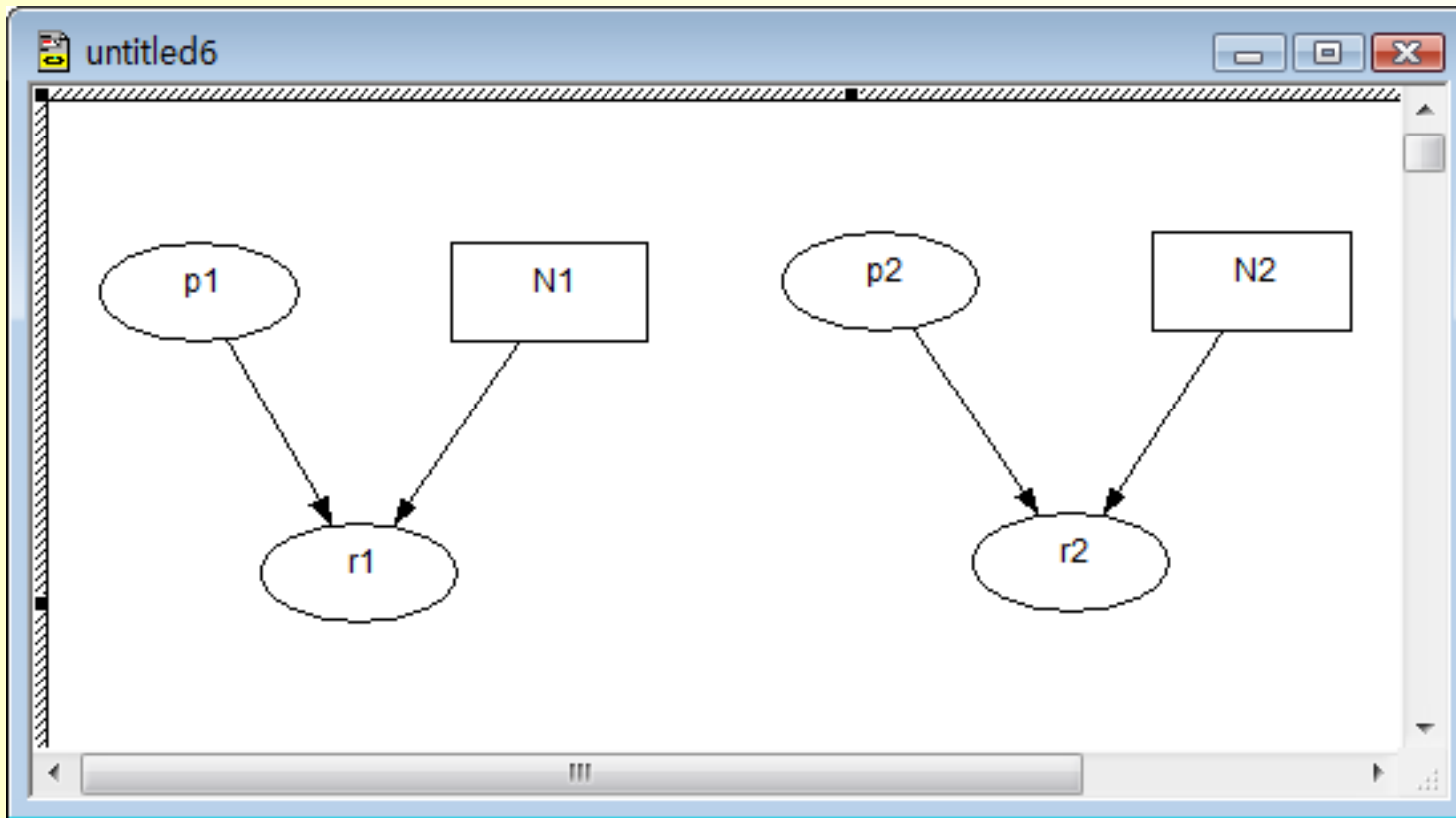
N=140, r=40



2nd example: two props comparison

- Two surveys are taken:
 - $N_1=14, r_1=4$
 - $N_2=12, r_2=5$
- Question: What is the evidence that the underlying rates for the two surveys is really different?
- Naïve thinking: $4/14=29\%, 5/12=42\%$ -> the difference appears to be large.
- Bayesian analysis using WinBUGS: calculate a posterior for the difference or ratio

- Create two sets of nodes as in 1st example





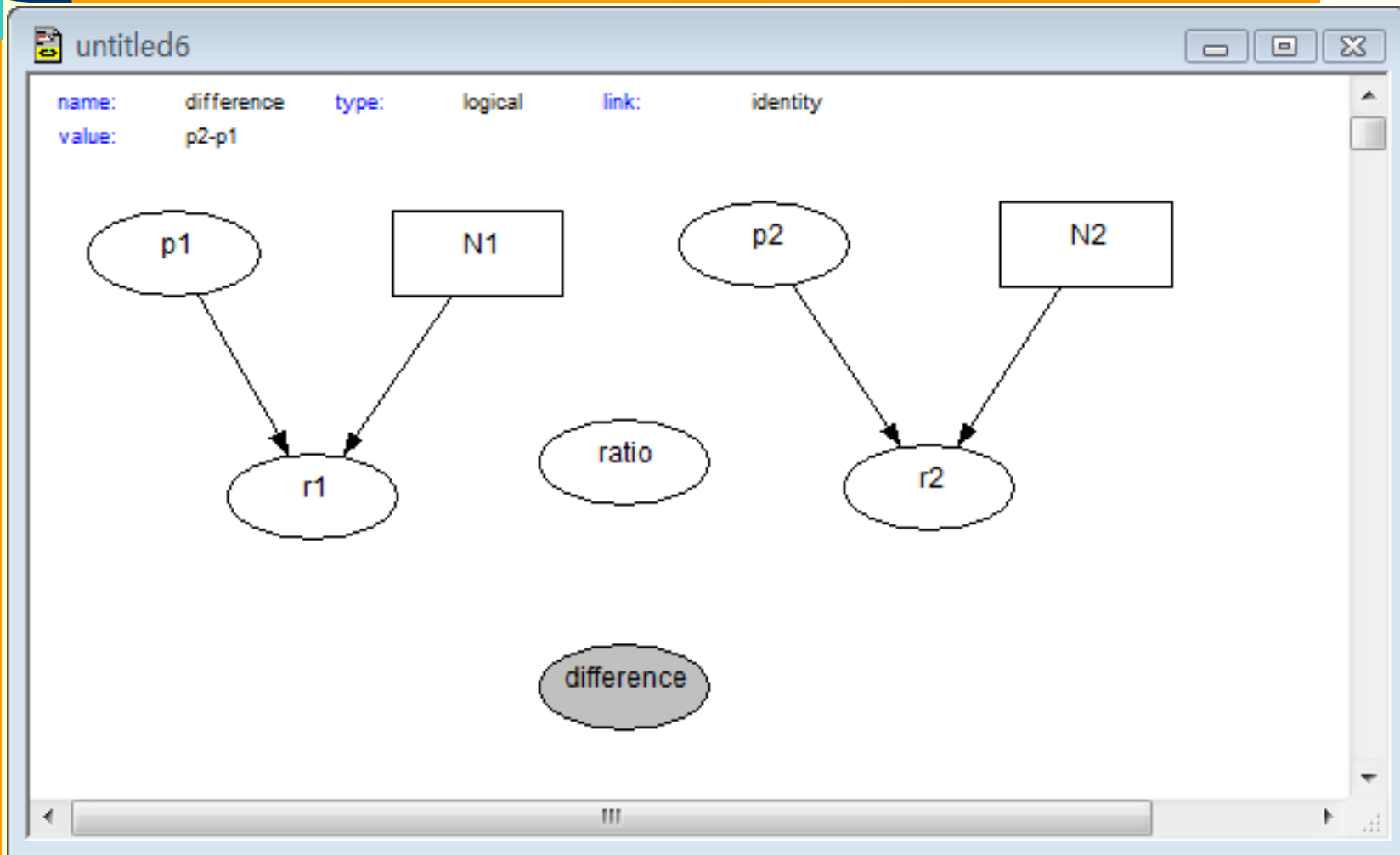
- create two logical nodes

1.

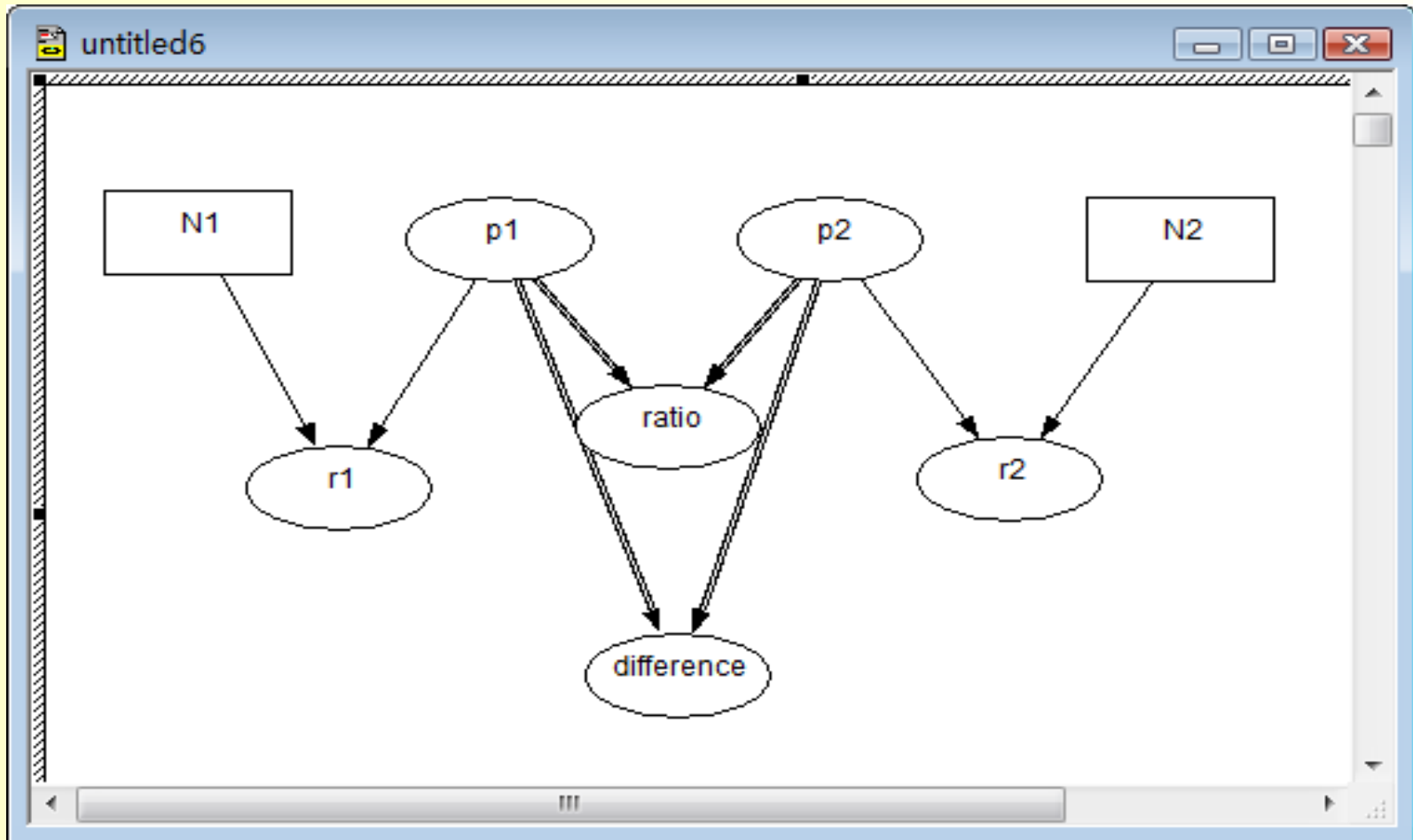
type: logical
name: **ratio**
link: identity
value: $p2/p1$

2.

type: logical
name: **difference**
link: identity
value: $p2-p1$.



- Add the links to finish the model.



Directed Graph



- corresponding code for the two proportion model: Select Doodle > Write Code

```
untitled9
model;
{
  p1 ~ dbeta(1,1)I( 0.2,0.45)
  r1 ~ dbin(p1,N1)
  r2 ~ dbin(p2,N2)
  p2 ~ dbeta(1,1)I( 0.2,0.45)
  ratio <- p2 / p1
  difference <- p2 - p1
}
```



- Summary: node connectors:

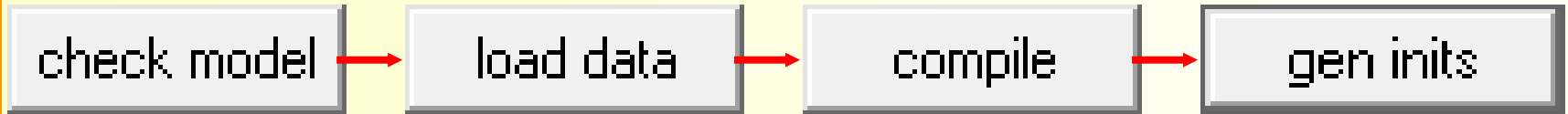
| connector | Solid arrow | Hollow arrow |
|-----------------------|--------------------------------------|------------------------------------|
| meaning | Stochastic /statistical relationship | Logical/deterministic relationship |
| Sign in BUGS language | \leftarrow | \sim |
| example | <code>diff<-p2-p1</code> | <code>r~dbin(p,N)</code> |

- Add the data before analyzing the model
 - In the model window type:
`list(N1=14, r1=4, N2=12, r2=5)`



Running BUGS again!

- Select Model > Specification..., do

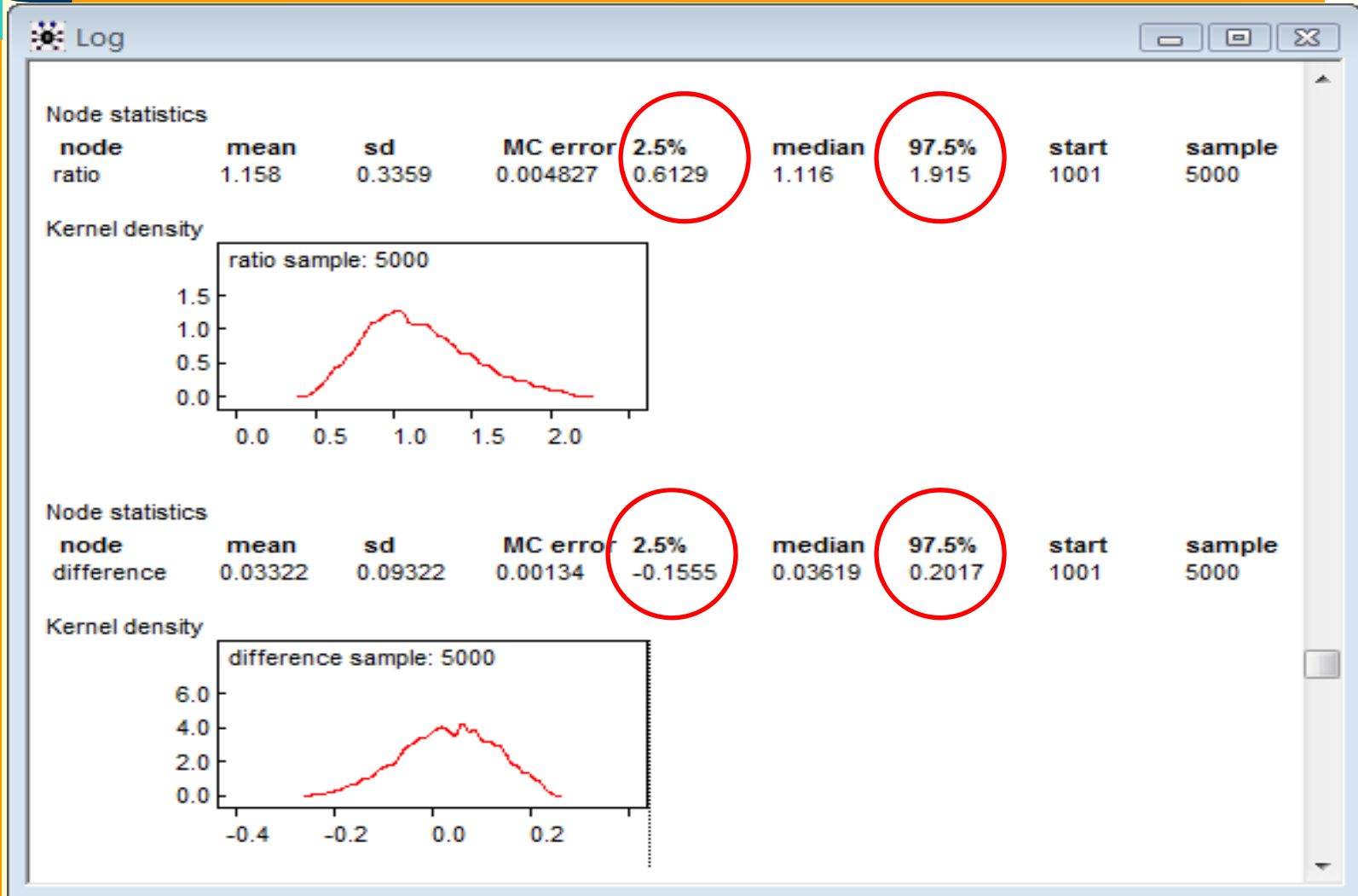


- Sample Monitor Tool to set **ratio** and **difference**
- Select Model > Update..., change updates to 5000



WinBUGS output--3

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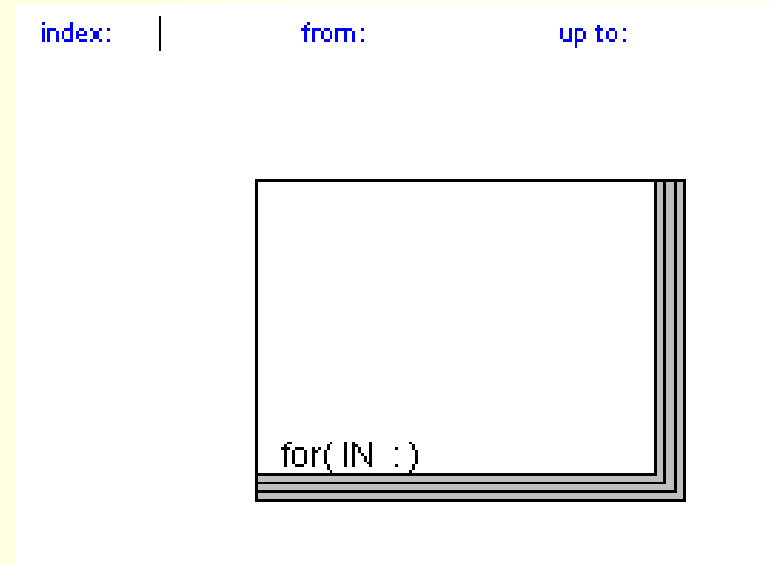
Conclusion

- The sample rates of unemployment are **NOT significantly different**. Reasons:
 - The area under the difference curve to the right of zero is not much larger than the area to the left of zero. Also the area to the right of one under the ratio curve is not much larger than the area to the left of one.
 - the mean difference is close to zero and the mean ratio is close to one
 - 95% credible interval for difference (-0.16,0.20)
95% credible interval for ratio (0.61, 1.92)



More complex models with plates

- Plates
 - Allow more complex structure, e.g.,
 - Repeated measures
 - Hierarchical models
- Creating a plate
 - CTRL + mouse click in Doodle Window
 - Deleting a plate: CTRL + Del





Case Study: pumps data

| PUMP | t_i | x_i |
|------|-------|-------|
| 1 | 94.5 | 5 |
| 2 | 15.7 | 1 |
| 3 | 62.9 | 5 |
| 4 | 126 | 14 |
| 5 | 5.24 | 3 |
| 6 | 31.4 | 19 |
| 7 | 1.05 | 1 |
| 8 | 1.05 | 1 |
| 9 | 2.1 | 4 |
| 10 | 10.5 | 22 |

- t_i : the length of operation time of the pump (in 1000s of hours).
- x_i : the number of failures



- Gamma-Poisson **hierarchical** model

for $i = 1, 2, \dots, 10$

$$x_i \sim \text{Poisson}(t_i \theta_i)$$

$$\theta_i \sim \text{Gamma}(\alpha, \beta)$$

$$\alpha \sim \text{Exponential}(1.0)$$

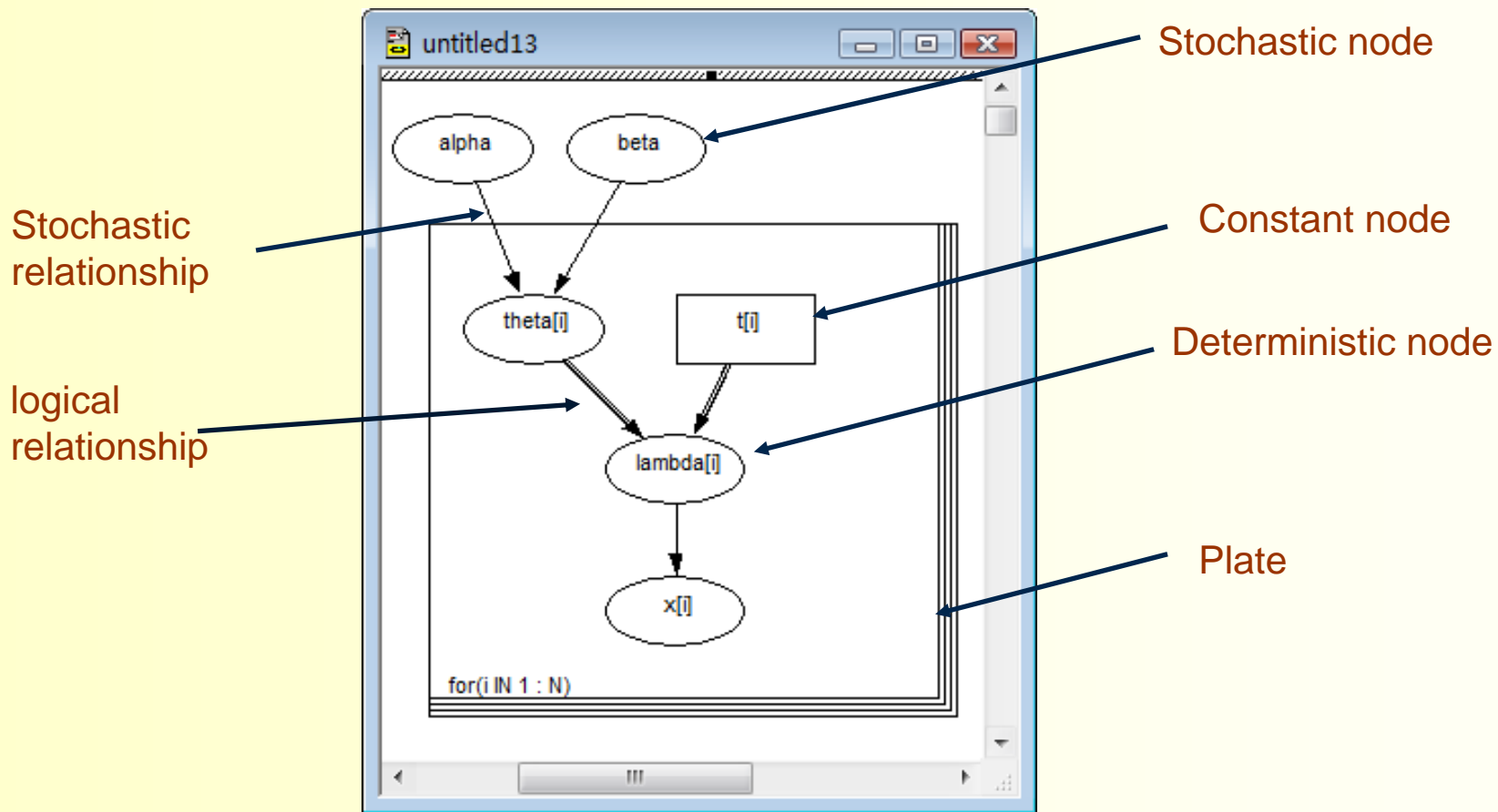
$$\beta \sim \text{Gamma}(0.1, 1.0)$$

- Parameters of interest: $\theta_i, i = 1, 2, \dots, 10$



Model specification through DoodleBUGS

- **Nodes**
 - Constants, denoted by rectangles
 - Stochastic nodes, denoted by ellipses
 - Deterministic nodes, logical function of other nodes
- **Edges**
 - Directed links
 - Solid arrow: stochastic dependence
 - Hollow arrow: logical function
 - Undirected links, dashed line
 - representing an upper or lower bound
- **Plates**, repeated parts of the graph





Model specification through BUGS language

```
model
{
  for (i in 1 : N) {
    theta[i] ~ dgamma(alpha, beta)
    lambda[i] <- theta[i] * t[i]
    x[i] ~ dpois(lambda[i])
  }
  alpha ~ dexp(1)
  beta ~ dgamma(0.1, 1.0)
}
```



Format data and specify initial values

- Data format

- R/S-Plus format using list()

```
list(t = c(94.3, 15.7, 62.9, 126, 5.24, 31.4, 1.05, 1.05, 2.1, 10.5),  
     x = c(5, 1, 5, 14, 3, 19, 1, 1, 4, 22),  
     N = 10)
```

- **Rectangular format**

- Initial values

```
list(alpha = 1, beta = 1)
```

```
list(alpha = 10, beta = 10)
```



- Data in rectangular format
 - Data are usually stored in the same **odc** file with END sign
 - To conserve space, hide/fold data in rectangular format by
 - **T o o l s > C r e a t e F o l d**
 - **Edit menu> select >copy>paste between two arrows**
 - **A name can be inserted in between**
 - Load the data by highlight the first row
 - Repeat the procedure for multiple sets of data

```
data
list(N = 10)
⇒
t[]   x[]
94.3  5
15.7  1
62.9  5
126   14
5.24  3
31.4  19
1.05  1
1.05  1
2.1   4
10.5  22
END
⇐
```

```
data
list(N = 10)
⇒Data(t,x)←
```



```
winbugs-demo4-complete

model;
{
  for( i in 1 : N ) {
    x[i] ~ dpois(lambda[i])
    lambda[i] <- theta[i] * t[i]
    theta[i] ~ dgamma(alpha,beta)
  }
  alpha ~ dexp(1)
  beta ~ dgamma(1,1)
}

data
list(N = 10)
→Data(t,x)←

Initial values
list(alpha = 1, beta = 1)
list(alpha = 10, beta = 10)
```




Anything missing?

- What kind of output can WinBUGS give us?
- What kind of sampling method does WinBUGS implement?



Output of WinBUGS

● Samples at Inference Menu

- ☞ **trace**: plots the variable value against iteration number.
- ☞ **history**: plots out a complete trace for the variable.
- ☞ **density**: smoothed kernel density estimate for continuous variable or a histogram for discrete variable.
- ☞ **auto cor**: auto correlation, up to lag-50
- ☞ **stats**: Summary statistics, pooling over the chains selected.
- ☞ **coda**: output the monitored values to CODA or BOA

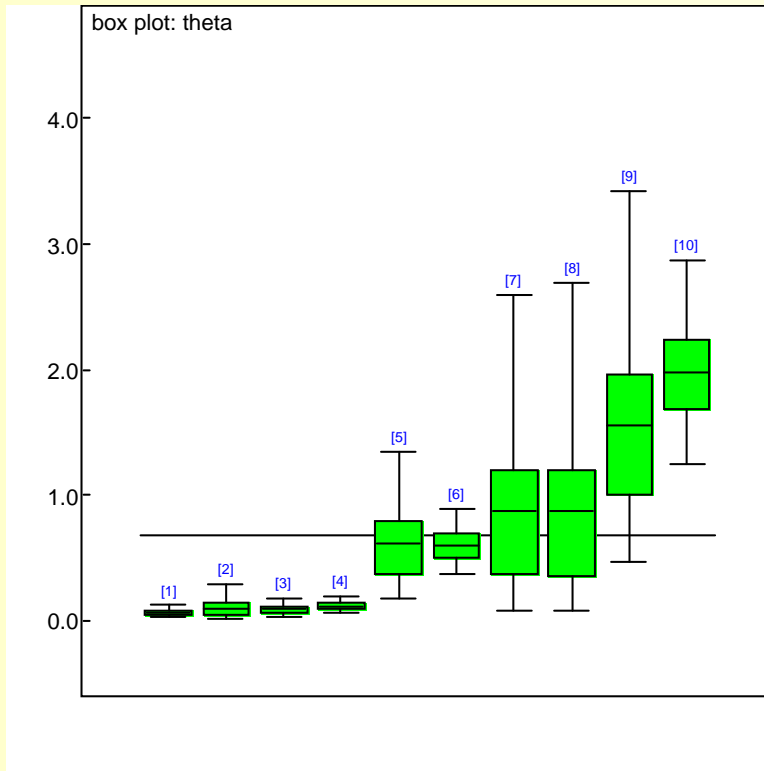


- ∞ **quantiles**: plots out the running mean with running 95% confidence intervals against iteration number.
- ∞ **bgr diag**: Brooks-Gelman-Rubin convergence statistic Brooks and Gelman (1998)
 - **Green**: the width of the central 80% interval of the pooled runs
 - **Blue**: the average width of the 80% intervals within the individual runs
 - **Red**: their ratio R ($=$ pooled / within).
 - **Convergence**: R to 1, and with convergence of both the pooled and within interval widths to stability.

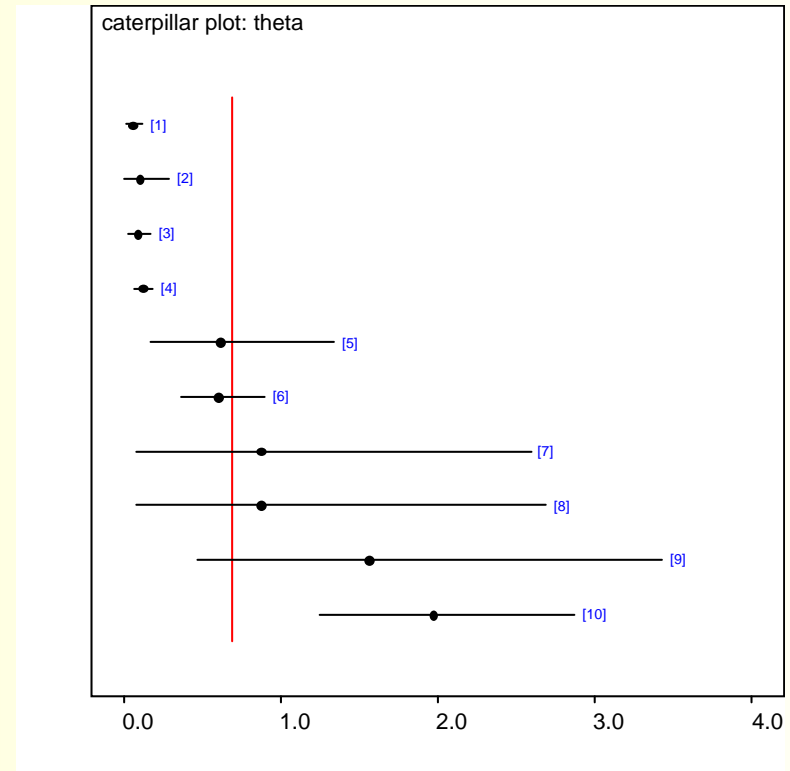


- Compare at Inference Menu

Box Plot



Caterpillar Plot





- **Rank at Inference Menu**
 - **stats:** the distribution of the ranks of each component of the variable.
 - **histogram:** the empirical distribution of the simulated rank for each component
- **DIC at Inference Menu**
 - **Deviance Information Criterion**



sampling methods in WinBUGS

Sampling methods are used in the following hierarchies

- **Continuous target distribution**

- **Conjugate** Direct sampling using standard algorithms
 - **Log-concave** Derivative-free adaptive rejection sampling
 - **Non-log-concave (restricted range)** Slice sampling
 - **Non-log-concave (unrestricted range)** Metropolis
-

- **Discrete target distribution**

- Finite upper bound Inversion
- Shifted Poisson Direct sampling using standard algorithm



R2WinBUGS: WinBUGS run in R

- It is a R package that provides the tools/functions to call WinBUGS directly.
- It automatically writes the data and scripts in a format readable by WinBUGS for processing in batch mode.
- Then it is possible to read resulting data into R to give
 - summary statistics
 - graphical summary
 - convergence diagnostics
 -



How to install R2WinBUGS

Suppose that the internet connection is available.

- Start R
- First way: at R command prompt type

`install.packages("R2WinBUGS")`

- Second way: install from R menu “packages”



How to use R2WinBUGS

We recommend to use R2WinEdt as a script editor and use school data as an example---to fit the hierarchical normal model.

- Step 1: Start R
- Step 2: Start WinBUGS
- Step 3: load R2WinBUGS from R by typing
`library("R2WinBUGS")` # or Before using bugs()
load RWinEdt from R by typing
`library("RWinEdt")`

Remark 1: We use the **school data** in Section 5.5 of "Bayesian Data Analysis(Gelman, 2004)" to illustrate



- Step 4: **Model specification** --- establish the **Bugs model file**: schools.bug

```
model {  
  for (j in 1:J){  
    y[j] ~ dnorm (theta[j], tau.y[j])  
    theta[j] ~ dnorm (mu.theta, tau.theta)  
    tau.y[j] <- pow(sigma.y[j], -2)  
  }  
  mu.theta ~ dnorm (0.0, 1.0E-6)  
  tau.theta <- pow(sigma.theta, -2)  
  sigma.theta ~ dunif (0, 1000)  
}
```



Remark 2: In Bugs, the normal distribution is parameterized by its precision (=1/variance)

Remark 3: For convenience, use τ for precisions and σ for standard deviations.



– Step 5: **data specification** (school.dat)

| | school | estimate | sd |
|---|--------|----------|------|
| A | | 28.39 | 14.9 |
| B | | 7.94 | 10.2 |
| C | | -2.75 | 16.3 |
| D | | 6.82 | 11.0 |
| E | | -0.64 | 9.4 |
| F | | 0.63 | 11.4 |
| G | | 18.01 | 10.4 |
| H | | 12.16 | 17.6 |



- Step 6: Data read, starting values and parameters specification (R_school_full.R)

```
schools<-read.table("c:/bugs.R/schools.dat",head=T)
J <- nrow(schools)
y <- schools$estimate
sigma.y <- schools$sd
data <- list("J","y","sigma.y")
inits <- function()
  list(theta=rnorm(J,0,100),
mu.theta=rnorm(1,0,100),
sigma.theta=runif(1,0,100))
parameters <- c("theta","mu.theta","sigma.theta")
```



Remark 5: Bugs does not require all parameters to be initialized, but it is a good idea to do so.

Starting values can be constructed

- randomly (as in the example above)
- with simple initial values
- from crude estimates



- Step 7: Run Bugs (MCMC simulation) with given number of chains for given number of iterations.

```
schools.sim <- bugs(data, inits, parameters,  
model.file="c:/bugs.R/schools.bug", n.chains=3, n.iter=1000,  
bugs.directory = "E:/WinBUGS14/")
```

Remark 6:

- n.chains= specifies the number of chains
- n.iter= specifies the number of iterations

Remark 7: While Bugs is running, it opens a new window and freezes R.



– Step 8: Show the results

- 8.1 Numerical output/summaries of the simulations for the parameters

```
print(schools.sim)  
schools.sim$summary
```

- 8.2 Graphical summaries of the inferences and convergence.

```
plot(schools.sim)
```

Remark 8:

- n.eff : rough measure of effective sample size
- Rhat : potential scale reduction factor
- pD: effective number of parameters
- DIC: estimate of expected predictive error



```
> print(schools.sim)
```

```
Inference for Bugs model at "c:/bugsR/schools.txt", fit using WinBUGS,  
3 chains, each with 1000 iterations (first 500 discarded)
```

```
n.sims = 1500 iterations saved
```

| | mean | sd | 2.5% | 25% | 50% | 75% | 97.5% | Rhat | n.eff |
|-------------|------|-----|------|------|------|------|-------|------|-------|
| theta[1] | 10.6 | 7.5 | -0.7 | 5.5 | 9.4 | 14.1 | 28.9 | 1 | 170 |
| theta[2] | 7.7 | 5.9 | -3.6 | 3.8 | 7.7 | 11.3 | 19.6 | 1 | 400 |
| theta[3] | 6.4 | 6.6 | -9.2 | 2.8 | 6.6 | 10.0 | 18.8 | 1 | 1500 |
| theta[4] | 7.3 | 5.9 | -4.4 | 3.5 | 7.4 | 11.0 | 19.6 | 1 | 450 |
| theta[5] | 5.6 | 5.7 | -7.5 | 2.5 | 5.9 | 9.2 | 16.2 | 1 | 1100 |
| theta[6] | 6.1 | 6.2 | -7.1 | 2.7 | 6.1 | 9.9 | 17.4 | 1 | 650 |
| theta[7] | 9.6 | 6.4 | -0.1 | 5.0 | 8.9 | 13.3 | 24.9 | 1 | 340 |
| theta[8] | 8.4 | 7.1 | -4.4 | 3.8 | 8.2 | 12.2 | 23.5 | 1 | 1000 |
| mu.theta | 7.8 | 4.5 | 0.0 | 4.6 | 7.8 | 10.7 | 16.6 | 1 | 280 |
| sigma.theta | 5.5 | 5.2 | 0.2 | 1.7 | 4.2 | 7.8 | 18.0 | 1 | 110 |
| deviance | 60.2 | 1.8 | 57.1 | 59.2 | 60.0 | 61.2 | 64.4 | 1 | 280 |

```
For each parameter, n.eff is a crude measure of effective sample size,  
and Rhat is the potential scale reduction factor (at convergence, Rhat=1).
```

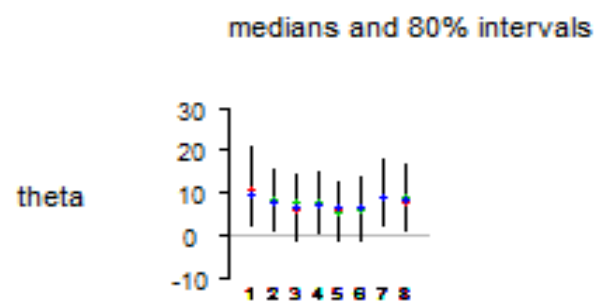
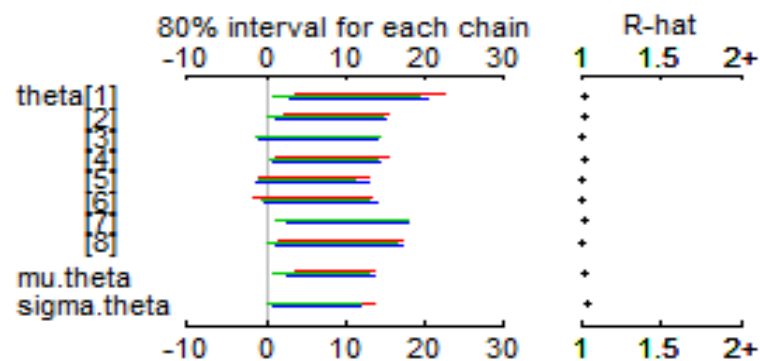
```
DIC info (using the rule,  $pD = \text{var}(\text{deviance})/2$ )
```

```
pD = 1.6 and DIC = 61.9
```

```
DIC is an estimate of expected predictive error (lower deviance is better).
```

文件 历史 重设大小

Bugs model at "c:/bugsR/schools.txt", fit using WinBUGS, 3 chains, each with 1000 iterations (first 500 discarded)





Summary

- BUGS is a power tool
 - Bayesian Analysis
 - Simulation Tool
- Model specification in WinBUGS
 - Graphical Models
 - DoodleBUGS: Direct graphics
 - Simple representation of model
 - BUGS language
- WinBUGS run in R (R2WinBUGS)
also: in S-plus, Stata, SAS,...



A Reminder

- WinBUGS is Easy to use!

But

- **MCMC** sampling can be **dangerous** in WinBUGS !



References

- [1] Petra Kuhnert, The Ecology Centre, UQ., An Introduction to WinBUGS 1.4
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- [3] D. J. Lunn, A. Thomas, N. BEST, D. Spiegelhalter, WinBUGS – A Bayesian modeling framework: Concepts, structure, and extensibility, *Statistics and Computing* (2000) **10, 325–337**
- [4] David Spiegelhalter, Andrew Thomas, Nicky Best, Dave Lunn, WinBUGS1.4 User Manual, 2003
- [5] Bayesing Modelling using WinBUGS by Ioannis Ntzoufras, 2009

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