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Modelling (& many co-workers!)

*The STAT-JR package
and it's potential use with social network
models*



Summary

- Background to STAT-JR package
- Some screen shots of the program features
- Multiple membership models for spatial models/ social networks
- STAT-JR named in memory of Jon Rasbash whose ideas started project.

The E-STAT project and STAT-JR

STAT-JR developed jointly by LEMMA II and E-STAT ESRC nodes

Consists of a set of components many of which we have an alpha version for which contains:

Templates for model fitting, data manipulation, input and output controlled via a web browser interface.

Currently model fitting for 90% of the models that MLwiN can fit in MCMC plus some it can't including greatly sped up REALCOM templates

Some interoperability with MLwiN, WinBUGS, R, Stata and SPSS (written by Camille Szmaragd)

STAT-JR

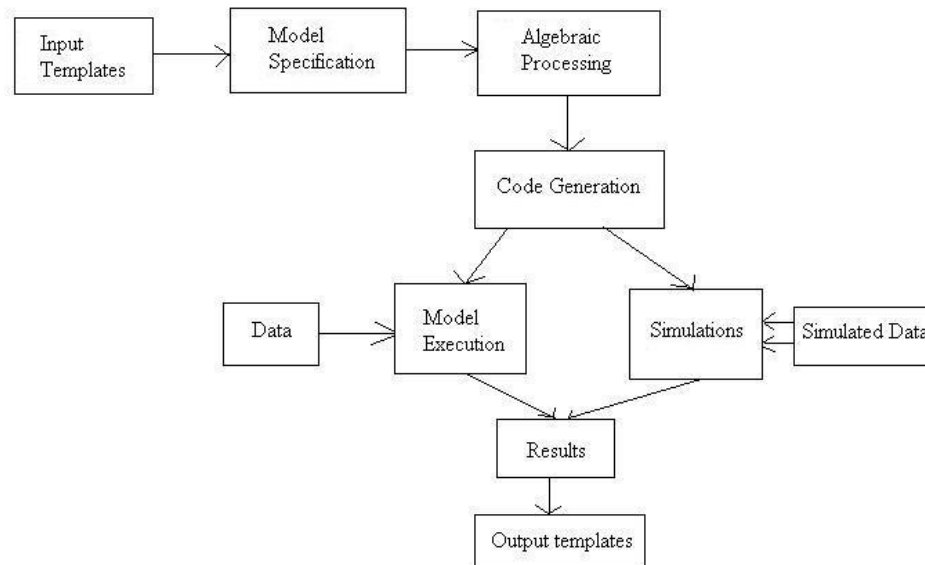
Jon identified 3 groups of users:

- Novice practitioners who want to use statistical software that is user friendly and maybe tailored to their discipline
- Advanced practitioners who are the experts in their fields and also want to develop tools for the novice practitioners
- Algorithm Developers who want their algorithms used by practitioners.
- See <http://www.cmm.bristol.ac.uk/research/NCESS-EStat/news.shtml> for details of Advanced User's guide for

STAT-JR

STAT-JR component based approach

Below is an early diagram of how we envisioned the system. Here you will see boxes representing components some of which are built into the STAT-JR system. The system is written in Python with currently a VB.net algebra processing system. A team of coders (currently me, Chris, Danius, Camille and Bruce) work together on the system.



Templates

- Consist of a set of code sections for advanced users to write.

For a model template it consists of at least:

- an invars method which specifies inputs and types
- An outbug method that creates (BUGS like) model code for the algebra system
- An (optional) outlatex method can be used for outputting LaTeX code for the model.

Other optional functions required for more complex templates

Regression 1 Example

```
from EStat.Templating import *
from mako.template import Template as
    MakoTemplate
import re
```

```
class Regression1(Template):
    'A model template for fitting 1 level Normal multiple
    regression model in E-STAT only. To be used in
    documentation.'
```

```
    tags = [ 'model' , '1-Level' ]
```

```
    invars = ""
    y = DataVector('response: ')
    tau = ParamScalar()
    sigma = ParamScalar()
    x = DataMatrix('explanatory variables: ')
    beta = ParamVector()
    beta.ncols = len(x)
    ""
```

```
    outbug = ""
    model{
        for (i in 1:length($y)) {
            ${y}[i] ~ dnorm(mu[i], tau)
            mu[i] <- ${mmult(x, 'beta', 'i')}
        }

        # Priors
        % for i in range(0, x.ncols()):
        beta${i} ~ dflat()
        % endfor
        tau ~ dgamma(0.001000, 0.001000)
        sigma <- 1 / sqrt(tau)
    }
    ""

    outlatex = r""
    \begin{aligned}
    \mbox{\${y}}_i & \sim \mbox{N}(\mu_i, \sigma^2) \\
    \mu_i & = \\
    & \${mmulttex(x, r'\beta', 'i')} \\
    %for i in range(0, len(x)):
    \beta_{\${i}} & \propto 1 \\
    %endfor
    \tau & \sim \Gamma(0.001, 0.001) \\
    \sigma^2 & = 1 / \tau \\
    \end{aligned}
    ""
```

Invars function

```
'''  
    invars = '''  
y = DataVector('response: ')  
tau = ParamScalar()  
sigma = ParamScalar()  
x = DataMatrix('explanatory variables: ')  
beta = ParamVector()  
beta.ncols = len(x)  
'''
```


An example of STAT-JR – setting up a model

Stat-JR Demonstrator

Template: Regression1 [Change Dataset](#) tutorial [Change View Summary](#)

Configuration [Start again](#)

response: normexam

explanatory variables: cons_standit

Name of output results: out

Random Seed:

length of burnin:

number of iterations:

thinning:

Inputs: {Y: 'normexam', X: 'cons_standit'}

Done

An example of STAT-JR – setting up a model



Configuration

Administrator

response: normexam

explanatory variables: cons,standlrt

Name of output results: out

Random Seed:

length of burnin:

number of iterations:

thinning:

done

Equations for model and model code

thinning: 1

Equation rendering

$$\text{normexam}_i \sim N(\mu_i, \sigma^2)$$

$$\mu_i = \beta_0 \text{cons}_i + \beta_1 \text{standlrt}_i + u_{\text{school}[i]}$$

$$u_{\text{school}[i]} \sim N(0, \sigma_u^2)$$

$$\beta_0 \propto 1$$

$$\beta_1 \propto 1$$

$$\tau \sim \Gamma(0.001, 0.001)$$

$$\sigma^2 = 1/\tau$$

$$\tau_u \sim \Gamma(0.001, 0.001)$$

$$\sigma_u^2 = 1/\tau_u$$

Model

```

model {
  for (i in 1:length(normexam)) {
    normexam[i] ~ dnorm(mu[i], tau)
    mu[i] <- cons[i] * beta0 + standlrt[i] * beta1 + u[school[i]] * cons[i]
  }
  for (j in 1:length(u)) {
    u[j] ~ dnorm(0, tau_u)
  }
  # Priors
  beta0 ~ dflat()
  beta1 ~ dflat()
  tau ~ dgamma(0.001000, 0.001000)
  sigma <- 1 / sqrt(tau)
  tau_u ~ dgamma(0.001000, 0.001000)
  sigma_u <- 1 / sqrt(tau_u)
}

```

Run Simulate Selection Simulate Specify starting Values Code JS Set

Inputs: {y: 'normexam', 'L2ID': 'school', 'D': 'Normal', 'x': 'cons,standlrt'}

Done

Note Equations use MATHJAX and so underlying LaTeX can be copied and paste. The model code is based around the WinBUGS language with some variation. This is a more complex template for 2 level models.

Equations for model and model code

normexam_{*i*} \sim N(μ_i, σ^2)

$$\mu_i = \beta_0 \text{cons}_i + \beta_1 \text{standlrt}_i + u_{\text{school}[i]}$$

$u_{\text{school}[i]} \sim$ N($0, \sigma_u^2$)

$$\beta_0 \propto 1$$

$$\beta_1 \propto 1$$

$$\tau \sim \Gamma(0.001, 0.001)$$

$$\sigma^2 = 1/\tau$$

$$\tau_u \sim \Gamma(0.001, 0.001)$$

$$\sigma_u^2 = 1/\tau_u$$

inputs: ty, normexam, L2ID, school, D, normal, A, cons,standlrt

Done

Note Equations use MATHJAX and so underlying LaTeX can be copied and paste. The model code is based around the WinBUGS language with some variation. This is a more complex template for 2 level models.

Outbug function

```
outbug = ""
model{
  for (i in 1:length($y)) {
    $y[i] ~ dnorm(mu[i], tau)
    mu[i] <- $mmult(x, 'beta', 'i')}
}

# Priors
% for i in range(0, x.ncols()):
beta${i} ~ dflat()
% endfor
tau ~ dgamma(0.001000, 0.001000)
sigma <- 1 / sqrt(tau)
}
""
```

Model code in detail

```
model {  
  for (i in 1:length(normexam)) {  
    normexam[i] ~ dnorm(mu[i], tau)  
    mu[i] <- cons[i] * beta0 + standlrt[i] * beta1 + u[school[i]] * cons[i]  
  }  
  for (j in 1:length(u)) {  
    u[j] ~ dnorm(0, tau_u)  
  }  
  # Priors  
  beta0 ~ dflat()  
  beta1 ~ dflat()  
  tau ~ dgamma(0.001000, 0.001000)  
  tau_u ~ dgamma(0.001000, 0.001000)  
}
```

For this template the code is, aside from the length function, standard WinBUGS model code.

Bruce's (Demo) algebra system step for parameter u

Testbed

File Window Distributions Functions Preferences About

Graph Nodes

normexam[-] school[-]	mu[-] tau_u	tau	cons[-]	beta0	standlrt[-]	beta1	u[-]
--------------------------	----------------	-----	---------	-------	-------------	-------	------

$$\text{Log posterior} = \tau \left(\sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i \left(-\text{normexam}_i + \beta_0 \text{cons}_i + \beta_1 \text{standlrt}_i \right) \right) u_j - \left(\frac{\text{tau_u}}{2} + \frac{\tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i^2}{2} \right) u_j^2$$

Distribution dnorm

Match $A = -\tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i \left(-\text{normexam}_i + \beta_0 \text{cons}_i + \beta_1 \text{standlrt}_i \right)$

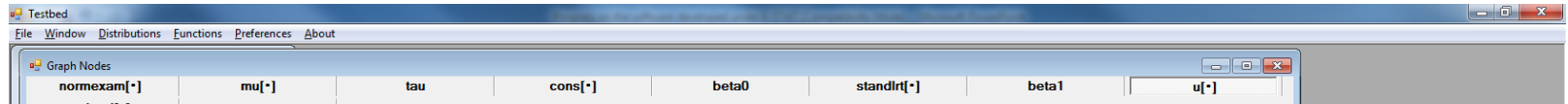
Match $B = -\left(\frac{\text{tau_u}}{2} \right) - \frac{\tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i^2}{2}$

Sampling parameter $\mu = -\frac{\left(\tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i \left(-\text{normexam}_i + \beta_0 \text{cons}_i + \beta_1 \text{standlrt}_i \right) \right)}{2 \left(\frac{\text{tau_u}}{2} + \tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i^2 / 2 \right)}$

Sampling parameter $\tau = 2 \left(\frac{\tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i^2}{\frac{\text{tau_u}}{2} + \frac{\tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i^2}{2}} \right)$

Sampling distribution $u_j \sim \text{dnorm} \left(-\frac{\left(\tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i \left(-\text{normexam}_i + \beta_0 \text{cons}_i + \beta_1 \text{standlrt}_i \right) \right)}{2 \left(\frac{\text{tau_u}}{2} + \tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i^2 / 2 \right)}, 2 \left(\frac{\text{tau_u}}{2} + \frac{\tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i^2}{2} \right) \right)$

Bruce's (Demo) algebra system step for parameter u



$$\text{Sampling parameter } \mu = - \left(\frac{\tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i \left(-\text{normexam}_i + \beta_0 \text{cons}_i + \beta_1 \text{standlrt}_i \right)}{2 \left(\tau \sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i^2 / 2 \right)} \right)$$

$$\text{Sampling parameter } \tau = 2 \left(\frac{\tau}{2} + \frac{\sum_{\substack{i=1 \\ \text{school}_i=j}}^{\text{length}(\text{normexam})} \text{cons}_i^2}{2} \right)$$

Output of generated C++ code



```

void iterate() {
// Update tau
{
    double sum0=0;
    for(int i=0; i<length(normexam); i++) {
        double se10 = double(normexam[int(i)]);
        double se11 = double(cons[int(i)]);
        double se12 = double(standlrt[int(i)]);
        double se15 = double(school[int(i)]);
        double se14 = double(u[int(se15)]);
        sum0+=pow(((se10-(beta0*se11))-(beta1*se12))-(se11*se14)),2);
    }

    std::tr1::gamma_distribution<double> gamma(0.001+(0.5*length(normexam)));
    tau = (1.0 / (0.001000+(sum0/2))) * gamma(eng);
}

// Update sigma_u
{
    sigma_u = (1/sqrt(tau_u));
}

// Update u
for(int j=0; j<length(u); j++){
{
    double sum0=0;
    double sum1=0;
    for(int id_i=0; id_i<length(idmap_school[j]); id_i++) { int i = idmap_school[j][id_i];
        double se10 = double(cons[int(i)]);
        double se11 = double(normexam[int(i)]);
        double se13 = double(standlrt[int(i)]);
        sum0+=(se10*(1-se11)+(beta0*se10)+(beta1*se13));
    }
    for(int id_i=0; id_i<length(idmap_school[j]); id_i++) { int i = idmap_school[j][id_i];
        double se14 = double(cons[int(i)]);
        sum1+=pow(se14,2);
    }

    std::tr1::normal_distribution<double> normal(-((tau*sum0)/(2*((tau_u/2)+((tau*sum1)/2)))), 1/sqrt((2*((tau_u/2)+((tau*sum1)/2))));

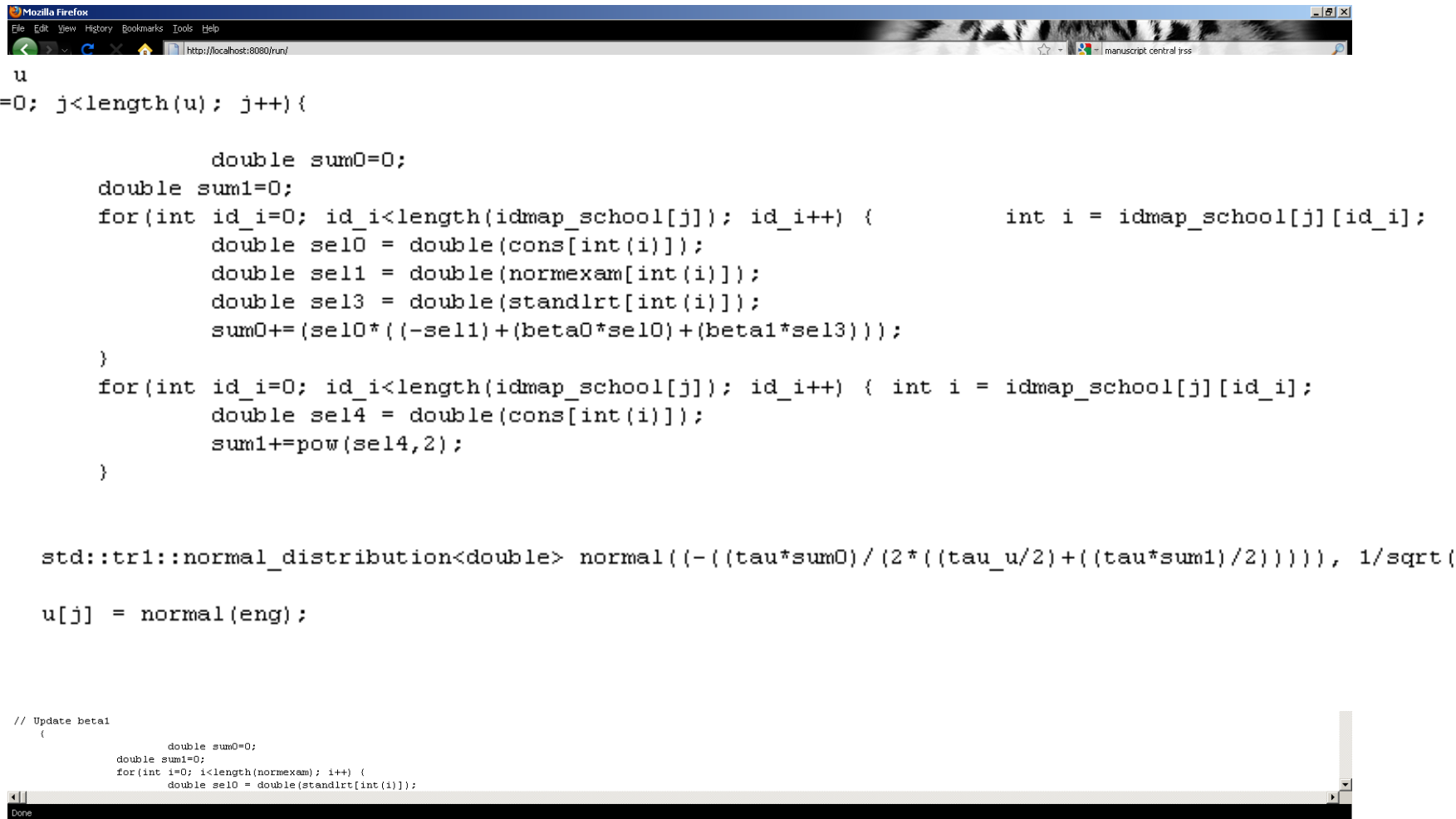
    u[j] = normal(eng);
}
}

// Update beta1
{
    double sum0=0;
    double sum1=0;
    for(int i=0; i<length(normexam); i++) {
        double se10 = double(standlrt[int(i)]);

```

The package can output C++ code that can then be taken away by software developers and modified.

Output of generated C++ code



```

// Update u
for(int j=0; j<length(u); j++){
    {
        double sum0=0;
        double sum1=0;
        for(int id_i=0; id_i<length(idmap_school[j]); id_i++) {           int i = idmap_school[j][id_i];
            double sel0 = double(cons[int(i)]);
            double sel1 = double(normexam[int(i)]);
            double sel3 = double(standlrt[int(i)]);
            sum0+=(sel0*((-sel1)+(beta0*sel0)+(beta1*sel3)));
        }
        for(int id_i=0; id_i<length(idmap_school[j]); id_i++) { int i = idmap_school[j][id_i];
            double sel4 = double(cons[int(i)]);
            sum1+=pow(sel4,2);
        }

        std::tr1::normal_distribution<double> normal((-((tau*sum0)/(2*((tau_u/2)+((tau*sum1)/2))))), 1/sqrt(

        u[j] = normal(eng);

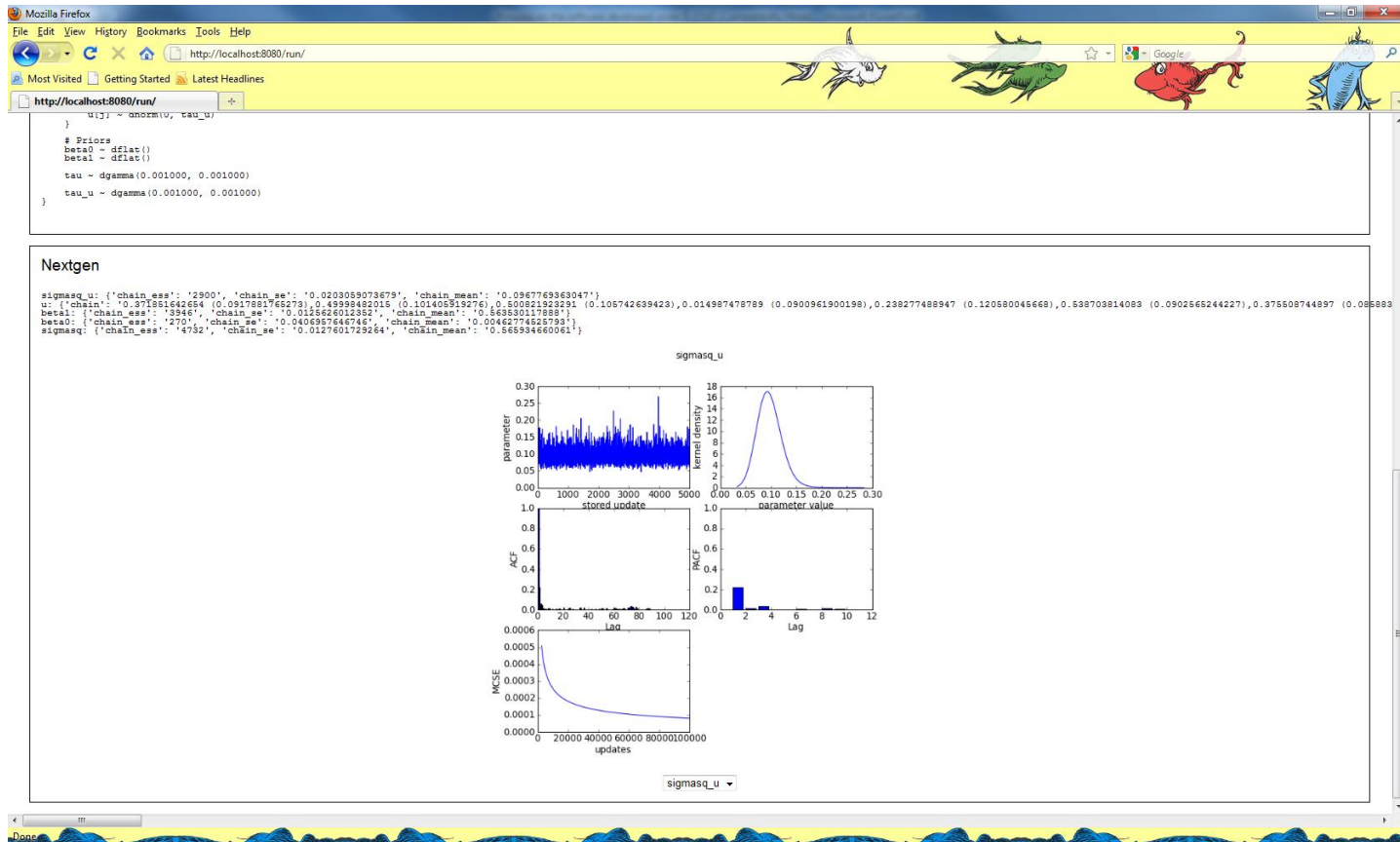
    }}

// Update beta1
{
    double sum0=0;
    double sum1=0;
    for(int i=0; i<length(normexam); i++) {
        double sel0 = double(standlrt[int(i)]);

```

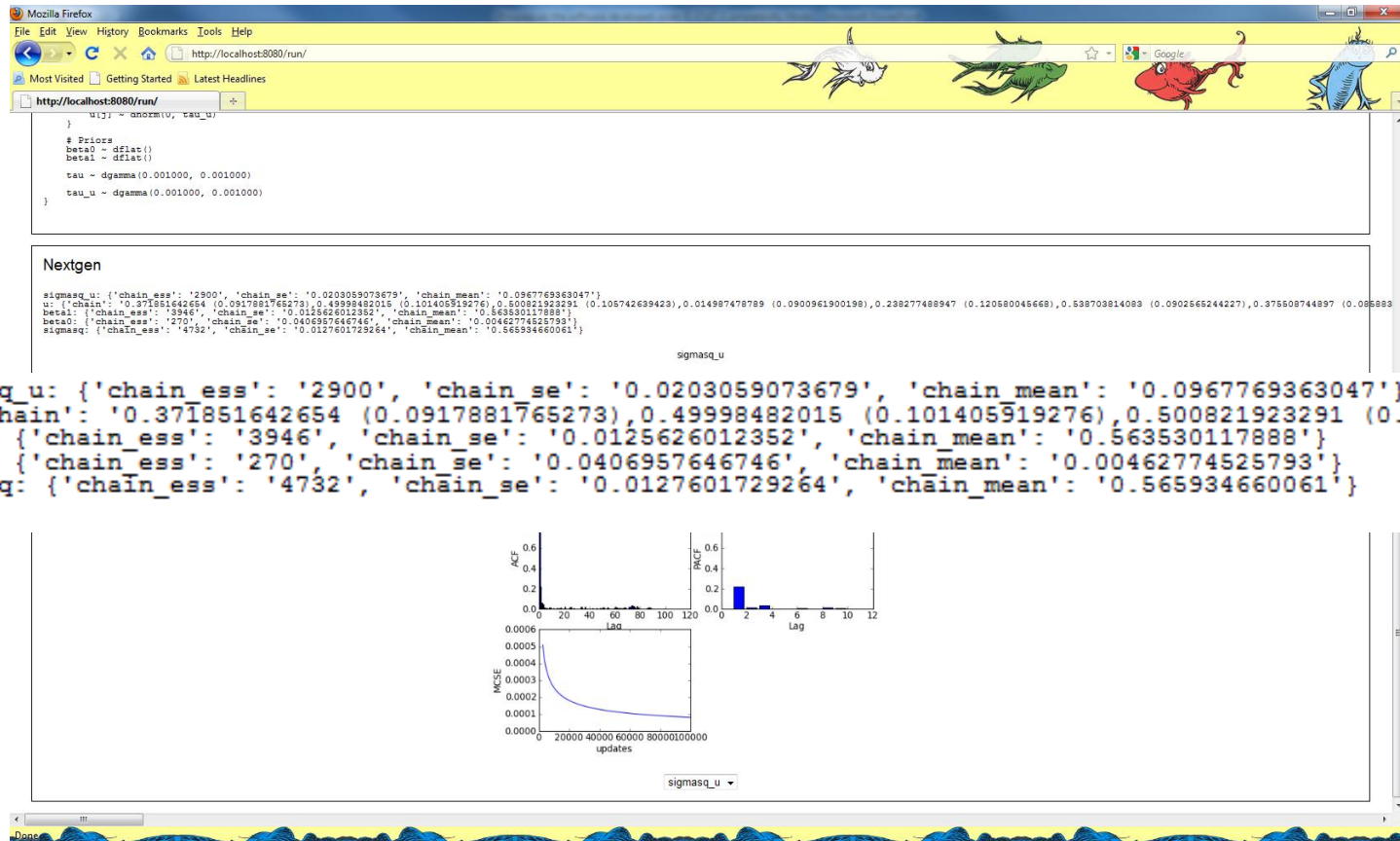
The package can output C++ code that can then be taken away by software developers and modified.

Output from the E-STAT engine



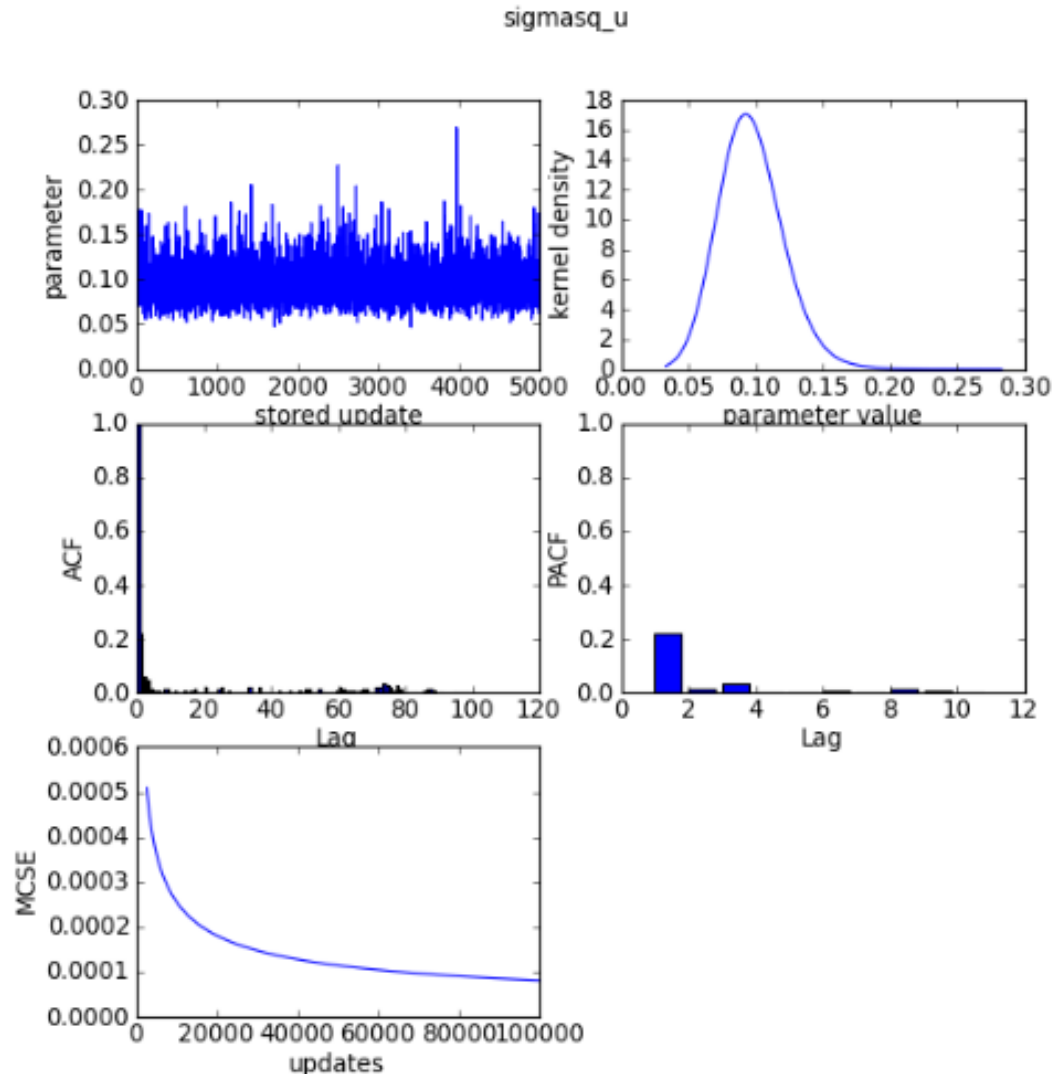
Here the six-way plot functionality is in part taken over to STAT-JR after the model has run. In fact graphs for all parameters are calculated and stored as picture files so can be easily viewed quickly.

Output from the E-STAT engine

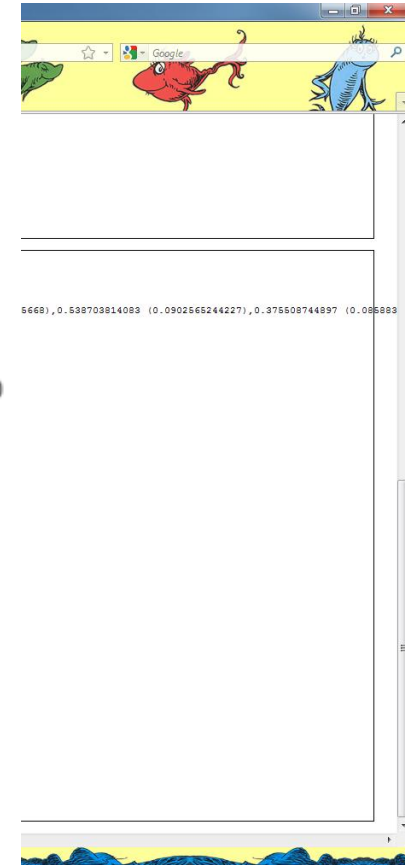


Here the six-way plot functionality is in part taken over to STAT-JR after the model has run. In fact graphs for all parameters are calculated and stored as picture files so can be easily viewed quickly.

Output from the E-STAT engine



can be easily viewed quickly.



to

o

Interoperability with WinBUGS

Stat-JR Demonstrator

Template: 2LevelMod [Change Dataset](#) [tutorial](#) [Change View](#) [Summary](#)

Configuration [Start again](#)

response: normexam
 Level 2 ID: school
 specify distribution: Normal
 explanatory variables: cons,standlrt
 Name of output results: out
 Is estimation method by MCMC: Yes
 Choose estimation engine - eSTAT, WinBUGS, MLwiN: WinBUGS
 number of chains: 3
 Random Seed: 1
 length of burnin: 1000
 number of iterations: 5000
 thinning: 1

Equation rendering

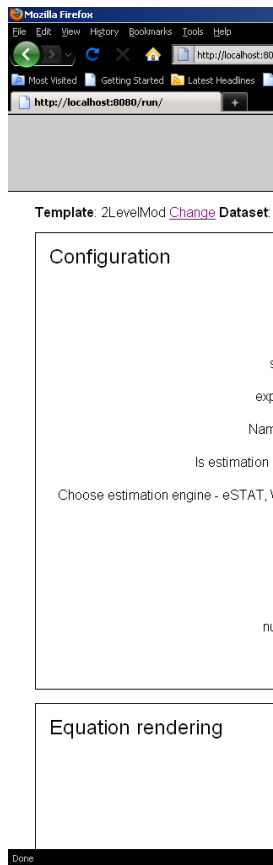
$$\text{normexam}_i \sim N(\mu_i, \sigma^2)$$

$$\mu_i = \beta_0 \text{cons}_i + \beta_1 \text{standlrt}_i + u_{\text{school}[i]}$$

$$u_{\text{school}[i]} \sim N(0, \sigma_u^2)$$

Interoperability in the user interface is obtained via a few extra inputs. In fact in the template code user written functions are required for all packages apart from WinBUGS. The transfer of data between packages is however generic.

Interoperability with WinBUGS



response: normexam

Level 2 ID: school

specify distribution: Normal

explanatory variables: cons,standlrt

Name of output results: out

Is estimation method by MCMC: Yes

Choose estimation engine - eSTAT, WinBUGS, MLwiN: WinBUGS

number of chains: 3

Random Seed: 1

length of burnin: 1000

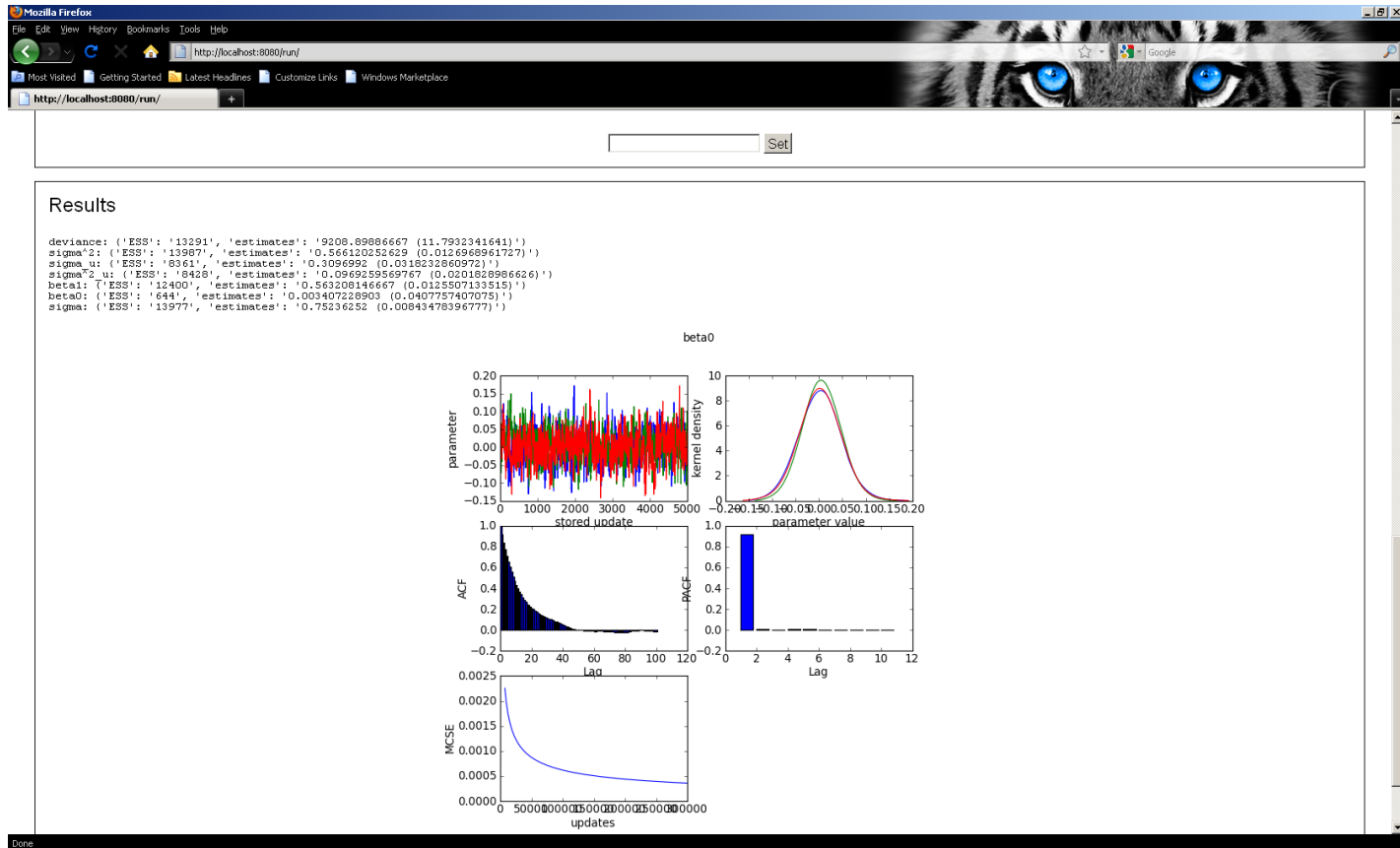
number of iterations: 5000

thinning: 1



Interoperability in the user interface is obtained via a few extra inputs. In fact in the template code user written functions are required for all packages apart from WinBUGS. The transfer of data between packages is however generic.

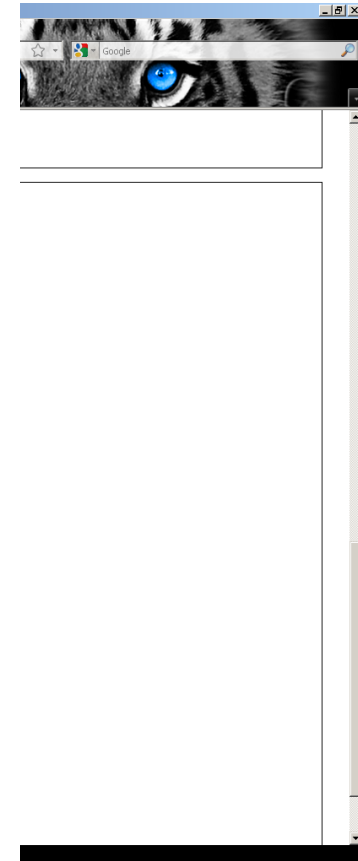
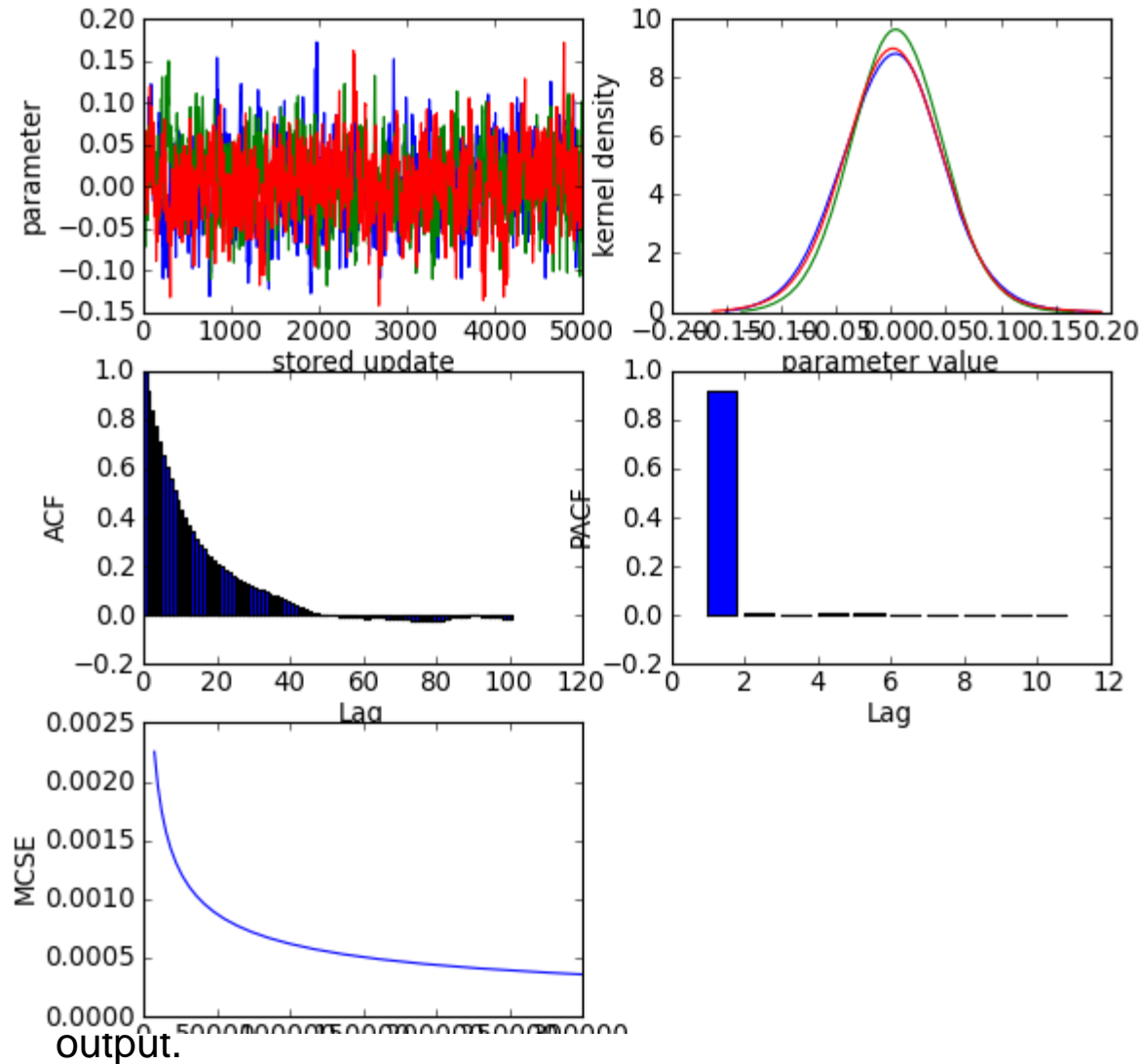
Output from WinBUGS with multiple chains



STAT-JR generates appropriate files and then fires up WinBUGS. Multiple Chains are superimposed in the sixway plot output.

Output from WinBUGS with multiple chains

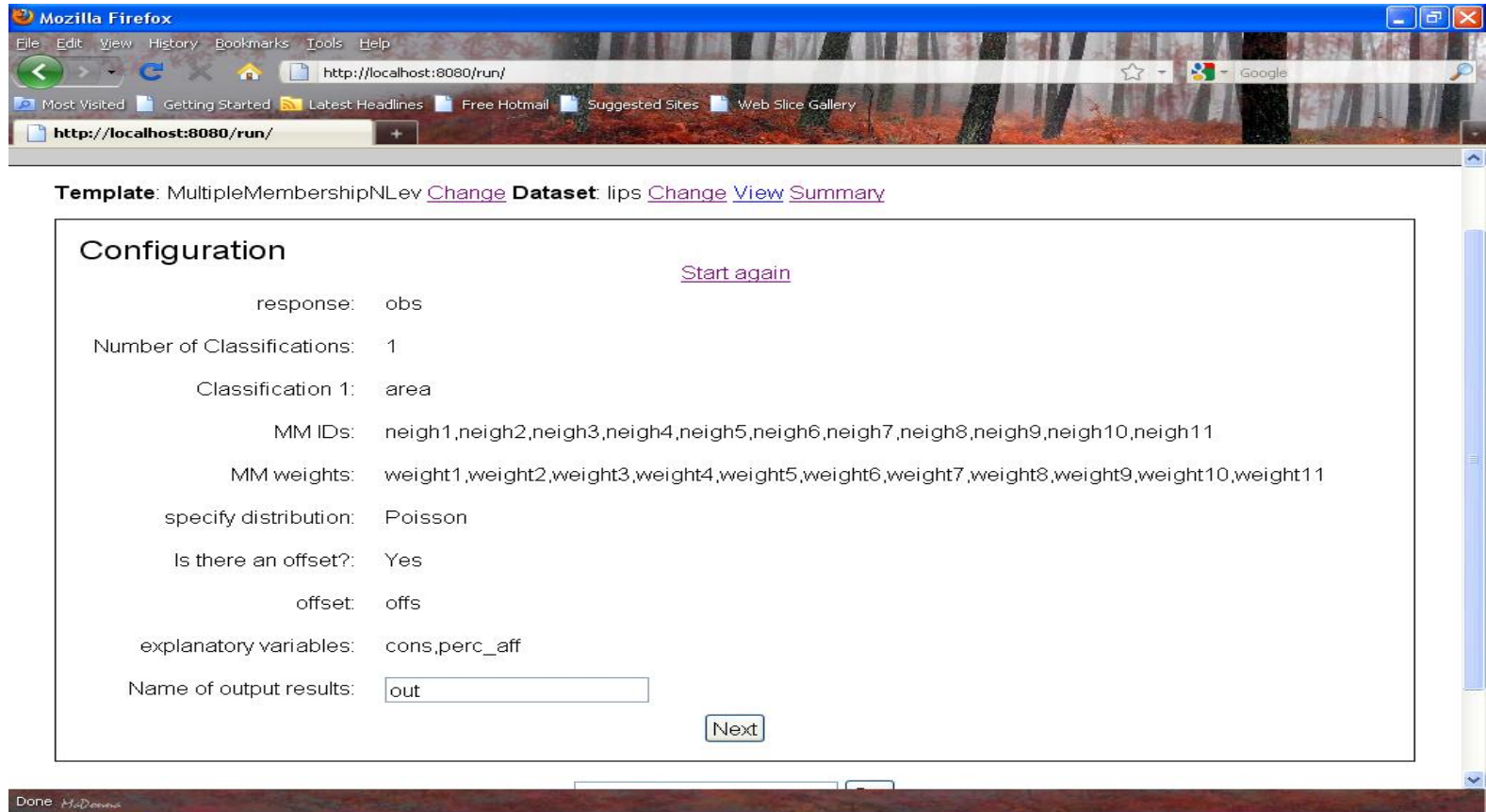
beta0



Multiple Membership Models

- Example is Scottish lip cancer data
- Response is Poisson (number of cases)
- Use as offset expected cases based on population size, makeup
- One predictor – percaff – percentage in agriculture, farming, fishing.
- Use the template MultipleMembershipNLev to allow both own random effect and neighbour random effects
- Template will allow fitting in STAT-JR engine, WinBUGS or MLwiN.

Inputs for model



Mozilla Firefox

File Edit View History Bookmarks Tools Help

http://localhost:8080/run/

Most Visited Getting Started Latest Headlines Free Hotmail Suggested Sites Web Slice Gallery

http://localhost:8080/run/

Template: MultipleMembershipNLev [Change Dataset:](#) lips [Change](#) [View Summary](#)

Configuration [Start again](#)

response: obs

Number of Classifications: 1

Classification 1: area

MM IDs: neigh1,neigh2,neigh3,neigh4,neigh5,neigh6,neigh7,neigh8,neigh9,neigh10,neigh11

MM weights: weight1,weight2,weight3,weight4,weight5,weight6,weight7,weight8,weight9,weight10,weight11

specify distribution: Poisson

Is there an offset?: Yes

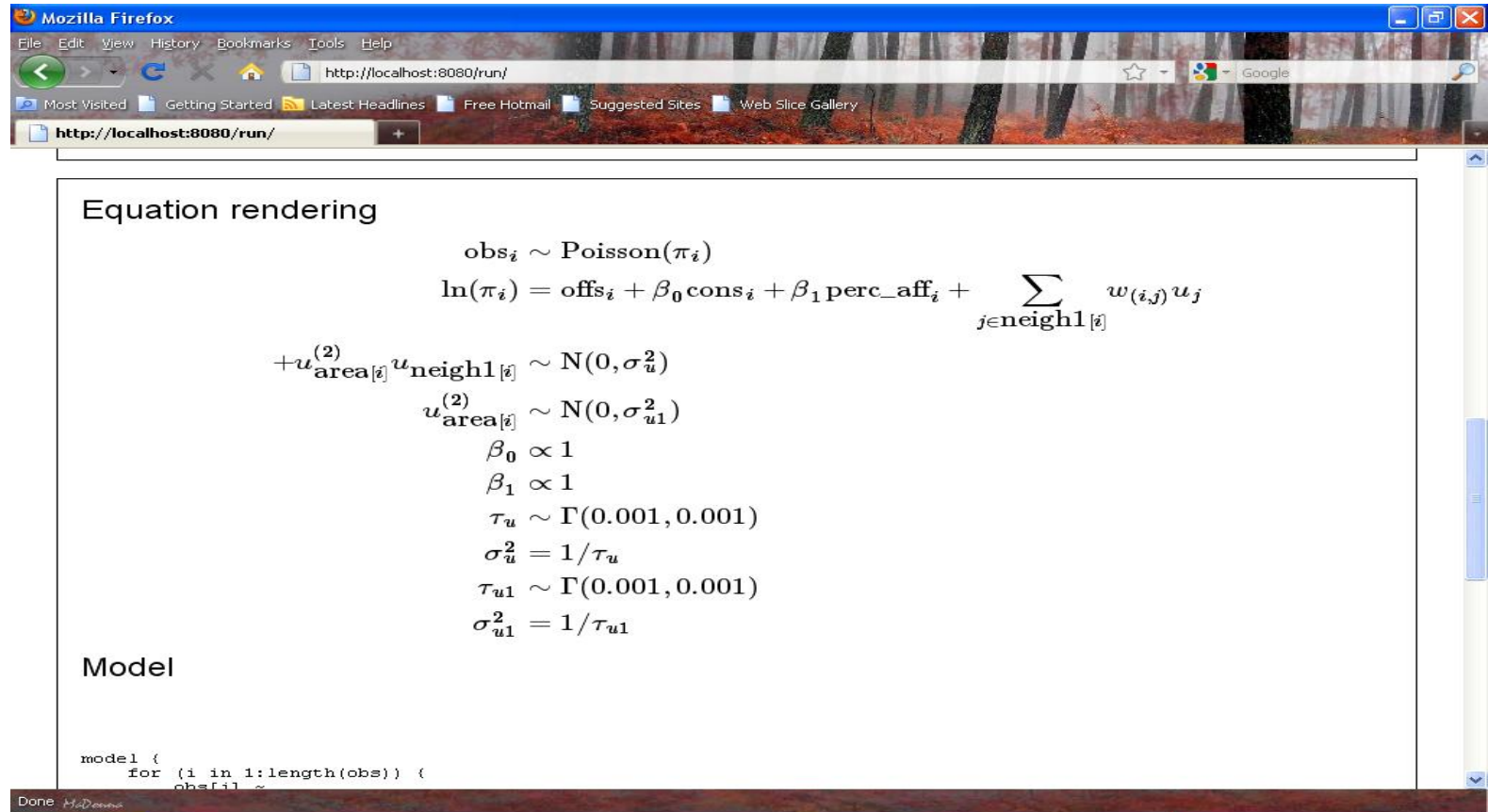
offset: offs

explanatory variables: cons,perc_aff

Name of output results:

Done: Mozilla

LaTeX for Model



Equation rendering

$$\text{obs}_i \sim \text{Poisson}(\pi_i)$$

$$\ln(\pi_i) = \text{offs}_i + \beta_0 \text{cons}_i + \beta_1 \text{perc_aff}_i + \sum_{j \in \text{neigh1}[i]} w_{(i,j)} u_j$$

$$+ u_{\text{area}[i]}^{(2)} u_{\text{neigh1}[i]} \sim \text{N}(0, \sigma_u^2)$$

$$u_{\text{area}[i]}^{(2)} \sim \text{N}(0, \sigma_{u1}^2)$$

$$\beta_0 \propto 1$$

$$\beta_1 \propto 1$$

$$\tau_u \sim \Gamma(0.001, 0.001)$$

$$\sigma_u^2 = 1/\tau_u$$

$$\tau_{u1} \sim \Gamma(0.001, 0.001)$$

$$\sigma_{u1}^2 = 1/\tau_{u1}$$

Model

```

model {
  for (i in 1:length(obs)) {
    obs[i] ~
  
```

Done Madam

Model Code

```

model {
  for (i in 1:length(obs)) {
    obs[i] ~
    dpois(p[i])
    ln(p[i]) <-
    offs[i] +
    cons[i] * beta0 + perc_aff[i] * beta1
  }
  + ul[area[i]] * cons[i]
  + weight1[i] * umm[neigh1[i]]
  + weight2[i] * umm[neigh2[i]]
  + weight3[i] * umm[neigh3[i]]
  + weight4[i] * umm[neigh4[i]]
  + weight5[i] * umm[neigh5[i]]
  + weight6[i] * umm[neigh6[i]]
  + weight7[i] * umm[neigh7[i]]
  + weight8[i] * umm[neigh8[i]]
  + weight9[i] * umm[neigh9[i]]
  + weight10[i] * umm[neigh10[i]]
  + weight11[i] * umm[neigh11[i]]
}

  for (i1 in 1:length(ul)) {
    ul[i1] ~ dnorm(0, tau_ul)
  }

  for (j in 1:length(umm)) {
    umm[j] ~ dnorm(0, tau_u)
  }

  # Priors
  beta0 ~ dflat()
  beta1 ~ dflat()

  tau_ul ~ dgamma(0.001000, 0.001000)
  sigma2_ul <- 1 / tau_ul

  tau_u ~ dgamma(0.001000, 0.001000)
  sigma2_u <- 1/tau_u
}

```

Run Simulate Selection Simulate Specify starting Values Code JS Save imputed every:

Done *MsDennis*

The E-STAT project – still to come

We have lots of work to do:

- Parallel processing.
- E-books.
- Optimising code generation.
- Improving algebra system.
- Suites of templates for missing data and social network models.
- Interoperability with SAS and hooking up more templates for other packages.