

Stat 5444: Model Selection

Consider the ('semi') linear model of the form:

$$y_i = \sum_{j=1}^p f_j(x_{i,j})\beta_j + \epsilon_i, \quad i = 1, \dots, N, \quad (1)$$

where $\epsilon_i \sim \text{Normal}(0, \sigma^2)$. More specifically, we can write $Y = X\beta + \epsilon$, where $Y = [y_1, \dots, y_n]^T$, $\beta = [\beta_1, \dots, \beta_p]^T$, and $X_{i,j} = f_j(x_{i,j})$. Note: In general, figuring out the functions $f_j(\cdot)$ can be difficult, but as a cursory step it is common to look at polynomial evaluations of $x_{i,j}$ (which can potentially lead to the dimensionality of the problem (p) being very large).

Problem 1

Work out the Bayes factor for comparing 2 different models, where each model is of the form given in equation (1), but differ by the number of free parameters (i.e. the

number of coefficients where $\beta_j \neq 0$).

$$BF = \frac{e^{-\frac{1}{2\sigma^2}(Y^TY)} \int_{\beta \in M_1} e^{-\frac{1}{2} \left(-\frac{2\beta_{\setminus j}^T X_{\setminus j}^T Y}{\sigma^2} + \beta_{\setminus j}^T \left(\frac{X_{\setminus j}^T X_{\setminus j}}{\sigma^2} + \frac{I}{\psi^2} \right) \beta_{\setminus j} + \mu^T E^{-1} \mu - \mu^T E^{-1} \mu \right)} d\beta}{\left(\frac{1}{\sqrt{2\pi}\psi} \right) \int_{\beta \in M_2} e^{-\frac{1}{2} \left(-\frac{2\beta^T X^T Y}{\sigma^2} + \beta^T \left(\frac{X^T X}{\sigma^2} + \frac{I}{\psi^2} \right) \beta + \mu^T E^{-1} \mu - \mu^T E^{-1} \mu \right)} d\beta}$$

$$BF = \frac{\sqrt{2\pi}\psi \times e^{\frac{1}{2}(\mu_{\setminus j}^T E_{\setminus j}^{-1} \mu_{\setminus j})} (2\pi)^{k/2} |E_{\setminus j}|^{1/2}}{e^{\frac{1}{2}(\mu^T E^{-1} \mu)} (2\pi)^{\frac{k+1}{2}} |E|^{1/2}}$$

$$E_{\setminus j} = \left(\frac{X_{\setminus j}^T X_{\setminus j}}{\sigma^2} + \frac{I}{\psi^2} \right)^{-1}$$

$$E = \left(\frac{X^T X}{\sigma^2} + \frac{I}{\psi^2} \right)^{-1}$$

$$\log(BF) = \log(numerator) - \log(denominator)$$

$$\log(numerator) = \log(\psi) + 0.5 * \frac{Y^T X_{\setminus j}}{\sigma^2} \left(\left(\frac{X_{\setminus j}^T X_{\setminus j}}{\sigma^2} + \frac{I}{\psi^2} \right)^{-1} \right)^T \frac{X_{\setminus j}^T Y}{\sigma^2} + 0.5 * \log(|E_{\setminus j}|)$$

$$\log(den) = 0.5 * \frac{Y^T X}{\sigma^2} \left(\left(\frac{X^T X}{\sigma^2} + \frac{I}{\psi^2} \right)^{-1} \right)^T \frac{X^T Y}{\sigma^2} + 0.5 * \log(|E|)$$

Problem 2

Using the data provided in **Model1_5444.txt**, search for the ‘true’ model using the following selection procedures:

- **Least Absolute Shrinkage Selection Operator (LASSO)**

$$\hat{\beta}_{LASSO} = ||Y - X\beta||_2^2 + \lambda \sum_{j=1}^p |\beta_j|,$$

- **Bayesian Information Criterion (BIC):** ‘Deviance’ + $\log(N)\Delta_p$, where Δ_p denotes the difference in the number of parameters used in the compared models,

- **Akaike Information Criterion (AIC):** ‘Deviance’ + $2\Delta_p$, where Δ_p denotes the difference in the number of parameters used in the compared models,
- **Stochastic Search Variable Selection (SSVS):**

$$\pi(\beta_j) = \pi_0\delta(\beta_j = 0) + (1 - \pi_0)N(\beta_j|0, \psi^2).$$

In your comparison between methods, describe in detail how you ‘tuned’ your method (if tuning is required), and report your selected model (including coefficient estimates).

Homework 6

Problem 2:

Stochastic Search Variable Selection:

The implementation of the code is similar to what was taught in the class. The main steps of the algorithm is as explained below,

Gibbs sampling steps:

- For t=1:T

For j=1:p

$$BF^{-1} = \frac{N(0|0, \psi^2)}{N(0|E, V)}$$

$$\hat{p} = \left(1 + \frac{1 - \pi}{\pi} BF^{-1}\right)^{-1}$$

$$\text{Where } V = \left(\frac{1}{\sigma^2} X_j^T X_j + \frac{1}{\psi^2}\right)^{-1}, E = \left(\frac{1}{\sigma^2} X_j^T X_j + \frac{1}{\psi^2}\right)^{-1} X_j^T (Y - X_{-j} \beta_{-j})$$

If $\text{BERNOULI}(\hat{p}) = 1$

$$\beta_j^t = 0$$

Else

$$\beta_j^t \sim N(\beta_j | E, V)$$

End

$$\text{Sample } \phi \sim \text{Gamma}\left(\frac{N}{2}, \frac{\text{Residual}}{2}\right)$$

END

The most important factor affecting the algorithm are

- 1.) Tuning the σ , standard deviation of response
- 2.) Tuning the ψ , the standard deviation for the beta coeff model

To address the ψ part, standardization is recommended for the predictor variables and hence $\psi = 1.69$ or 1.96 or 2.25

In the present exercise the comparative study is also done.

The code is attached in appendix 1

```

clear all
clc
close all
%% Stochastic Variable selection algorithm matlab script
% This code is based on conjugate priors

%% Generate pseudo random data

init_data.order=3;

% init_data.predictor=randn([init_data.N,init_data.predictor_np]);

% init_data.nofterms=floor(size(init_data.predictor_column,2)/2);
% rng(1,'v5uniform');
%
init_data.actualmodelterms=randi([1,size(init_data.predictor_column,2)],[init_data.nofterms,1]);
%
init_data.actualmodelterms=sort(randsample(size(init_data.predictor_column,2),[init_data.nofterms]));
%
init_data.actualmodelpredictors=init_data.predictor_column(:,init_data.actualmodelterms);
% beta_predictors=4*randn([length(init_data.actualmodelterms),1]);
%
init_data.response=init_data.actualmodelpredictors*beta_predictors+0.3*randn([init_data.N,1]);

data=xlsread('model1.xlsx');
%% SVSS algorithm initialization

data_available.response=data(:,1);
init_data.predictor=data(:,2:end);
init_data=generate_data(init_data);
% init_data.predictor_column=[ones(length(data_available.response),1)
init_data.predictor_column];
% data_available.response=zscore(init_data.response);
% data_available.predictor_library=init_data.predictor_column;
data_available.predictor_library=zscore(init_data.predictor_column);
data_available.predictor_library=[ones(length(data_available.response),1)
data_available.predictor_library];

data_available.init_pi0=0.5;
data_available.Xvar=1.96^2;
% M1=[10:-1:2]%init_data.actualmodelterms;
% M2=[10:-1:2 1];
%
%
```

```

%
% [bf,pr]=compute_bayes_fac(M1,M2,data_available);
% pr.m1_m2
%% MCMC starting

disp('Gibbs sampling running');
disp('%%%%%%%%%%%%');

T=40000;flg=0;
p=size(data_available.predictor_library,2);
model.heatmap=cell(T,1);
model.global_indicator=cell(T,1);
model.global_indicator{1,1}=ones(1,p);
model.selected=cell((p),1);
model.sumterms=zeros(p,1);
N=length(data_available.response);
data_available.var_Y=0.3;
list_var=1:1:p;
for t=2:T

    model.indicator=model.global_indicator{t-1,1};
    data_available.beta=model.indicator;
    if (rem(t*100/T,20)==0)
        flg=flg+1;
        fprintf('Gibbs sampling: %d percent done\n',20*flg);
    end

    for j=1:p
        M1=j;
        M2=find(list_var~=j);
        [bf_inv,pr]=compute_bayes_fac_V2(M1,M2,data_available);

        if binornd(1,pr.H0)==1
            model.indicator(1,j)=0;
            model.betaind(1,j)=0;
        else
            model.indicator(1,j)=normrnd(pr.E,sqrt(pr.V));
            model.betaind(1,j)=1;
        end
        data_available.beta=model.indicator;
    end
    model.residuals=data_available.response-
data_available.predictor_library*model.indicator';
    model.B=(sum(model.residuals.^2))*0.5;
    model.phi(t,1)=gamrnd(N/2,1/model.B);
    data_available.var_Y=1/model.phi(t,1);
    model.global_indicator{t,1}=model.indicator;
    model.global_betaindicator{t,1}=model.betaind;
    model.heatmap{t,1}=find(model.indicator~=0);
end

```

```

        if flg>1
        if isempty(model.heatmap{t,1})==0
%           length(model.selected(length(model.heatmap{t,1})))=len;
            model.selected{length(model.heatmap{t,1}),1}(1,end+1)=t;

model.sumterms(length(model.heatmap{t,1}),1)=model.sumterms(length(model.heatmap{t,1}),1)+1;
        end
    end
end
%%
[~,Ind] = max(model.sumterms);newjj=1;jj=1;member=[];clstr=0;
while(jj<length(model.selected{Ind,1}))
    tterm=model.selected{Ind,1}(1,jj);cntr=0;
    totmember=0;clstr=clstr+1;newjj=[];
    for kk=jj:length(model.selected{Ind,1})
        tterm2=model.selected{Ind,1}(1,kk);
        comp_model=model.global_betaindicator{tterm2,1};
        if(ismember(comp_model,model.global_betaindicator{tterm,1}, 'rows'))
            totmember=totmember+1;
            member{clstr,1}(totmember)=tterm2;
            clstrlen(clstr,1)=totmember;
        else
%           clstr=clstr+1
            newjj(1,end+1)=kk;
        end
    end
    if isempty(newjj)
        break
    else
        jj=newjj(1);
    end
end
[~,indc]=max(clstrlen);
model.betarray=cell2mat(model.global_indicator(member{indc,1}));
model.meanbetarray=mean(model.betarray);
disp(model.meanbetarray);

%%

clear beta_var
beta_var{1}='Intercept';
for ii=1:50
    beta_var{1+ii}=(sprintf('beta-%d',ii));
end
flg=51;
for ii=1:50
    for jj=ii:ii
        flg=flg+1;
        beta_var{flg}=(sprintf('beta_%d_%d',ii,jj));
    end
end

```

```

        end
    end

    flg=101;
    for ii=1:50
        for jj=ii:ii
            for kk=jj:jj
                flg=flg+1;
                beta_var{flg}=(sprintf('beta_%d_%d_%d',ii,jj,kk));
            end
        end
    end
    beta_var=beta_var';
    VarNames={'Beta', 'Value'};
    T=table(beta_var,model.meanbetarray', 'VariableNames',VarNames);
    %%

function [bf_inv,pr]=compute_bayes_fac_V2(M1,M2,data_available)
    Y=data_available.response;
    pi0=data_available.init_pi0;
    Y_var=data_available.var_Y;
    N=length(data_available.response);
    X_variance=data_available.Xvar;
    X_M1=data_available.predictor_library(:,M1); %X_j
    X_M2=data_available.predictor_library(:,M2); %X_(-j)
    Beta_M2=data_available.beta(:,M2)';

    Variance=1/((1/Y_var)*(X_M1'*X_M1)+(1/X_variance));
    Meanval=Variance*X_M1'*(Y-X_M2*Beta_M2)/Y_var;
    log_num=log(1/sqrt(X_variance));
    log_denom=log(1/sqrt(Variance))-0.5*(Meanval^2)/Variance;
    log_bayes_fac_inv=log_num-log_denom;
    bf_inv=exp(log_bayes_fac_inv);
    pr.H0=inv(1+((1-pi0)/pi0)*bf_inv);
    pr.E=Meanval;
    pr.V=Variance;
end

```


Results of SVSS for Model 1

```

clear beta_var
beta_var{1}='Intercept';
for ii=1:50
    beta_var{1+ii}=(sprintf('beta-%d',ii));
end
flg=51;
for ii=1:50
    for jj=ii:ii
        flg=flg+1;
        beta_var{flg}=(sprintf('beta_%d_%d',ii,jj));
    end
end

flg=101;
for ii=1:50
    for jj=ii:ii
        for kk=jj:jj
            flg=flg+1;
            beta_var{flg}=(sprintf('beta_%d_%d_%d',ii,jj,kk));
        end
    end
end
beta_var=beta_var';
VarNames={'Beta','Value'};
T=table(beta_var,model.meanbetarray','VariableNames',VarNames);
disp(T)

```

Beta	Value
'Intercept'	-2.8047
'beta-1'	0
'beta-2'	0
'beta-3'	1.8132
'beta-4'	0
'beta-5'	0
'beta-6'	0
'beta-7'	0
'beta-8'	0
'beta-9'	0
'beta-10'	0
'beta-11'	0
'beta-12'	0
'beta-13'	0
'beta-14'	0
'beta-15'	0
'beta-16'	0
'beta-17'	1.1262
'beta-18'	-0.35505
'beta-19'	0
'beta-20'	0
'beta-21'	0
'beta-22'	0
'beta-23'	0
'beta-24'	0

'beta-25'	0
'beta-26'	0
'beta-27'	2.5055
'beta-28'	0
'beta-29'	0
'beta-30'	0
'beta-31'	0
'beta-32'	0
'beta-33'	0
'beta-34'	0
'beta-35'	0
'beta-36'	0
'beta-37'	0
'beta-38'	0.39425
'beta-39'	0
'beta-40'	-0.6552
'beta-41'	0
'beta-42'	0
'beta-43'	0
'beta-44'	0
'beta-45'	4.3655
'beta-46'	0
'beta-47'	0
'beta-48'	0
'beta-49'	0
'beta-50'	-0.72537
'beta_1_1'	0
'beta_2_2'	0
'beta_3_3'	0
'beta_4_4'	-0.076734
'beta_5_5'	0
'beta_6_6'	0
'beta_7_7'	0
'beta_8_8'	0
'beta_9_9'	0
'beta_10_10'	0.41249
'beta_11_11'	0.29997
'beta_12_12'	0
'beta_13_13'	0
'beta_14_14'	0
'beta_15_15'	-0.13085
'beta_16_16'	0
'beta_17_17'	0
'beta_18_18'	0
'beta_19_19'	0.088538
'beta_20_20'	0
'beta_21_21'	0
'beta_22_22'	0
'beta_23_23'	0
'beta_24_24'	0
'beta_25_25'	0
'beta_26_26'	0
'beta_27_27'	0
'beta_28_28'	-0.42006
'beta_29_29'	0
'beta_30_30'	0
'beta_31_31'	0
'beta_32_32'	0
'beta_33_33'	0
'beta_34_34'	0
'beta_35_35'	-1.1168

'beta_36_36'	0
'beta_37_37'	-0.39743
'beta_38_38'	0
'beta_39_39'	0
'beta_40_40'	0
'beta_41_41'	0
'beta_42_42'	0
'beta_43_43'	0
'beta_44_44'	0
'beta_45_45'	0
'beta_46_46'	0
'beta_47_47'	0
'beta_48_48'	-0.10212
'beta_49_49'	0
'beta_50_50'	0
'beta_1_1_1'	0
'beta_2_2_2'	0.16643
'beta_3_3_3'	0
'beta_4_4_4'	-1.5549
'beta_5_5_5'	0
'beta_6_6_6'	0
'beta_7_7_7'	0
'beta_8_8_8'	0
'beta_9_9_9'	0
'beta_10_10_10'	0
'beta_11_11_11'	0
'beta_12_12_12'	0
'beta_13_13_13'	0.34192
'beta_14_14_14'	0
'beta_15_15_15'	0
'beta_16_16_16'	0
'beta_17_17_17'	0
'beta_18_18_18'	0
'beta_19_19_19'	0
'beta_20_20_20'	0
'beta_21_21_21'	0
'beta_22_22_22'	0
'beta_23_23_23'	0
'beta_24_24_24'	0
'beta_25_25_25'	0
'beta_26_26_26'	0
'beta_27_27_27'	0
'beta_28_28_28'	0
'beta_29_29_29'	0
'beta_30_30_30'	0
'beta_31_31_31'	0
'beta_32_32_32'	-0.16802
'beta_33_33_33'	-0.081993
'beta_34_34_34'	-0.27586
'beta_35_35_35'	0
'beta_36_36_36'	0
'beta_37_37_37'	0
'beta_38_38_38'	0
'beta_39_39_39'	-0.46109
'beta_40_40_40'	0
'beta_41_41_41'	-0.68102
'beta_42_42_42'	0
'beta_43_43_43'	0
'beta_44_44_44'	0
'beta_45_45_45'	0.14931
'beta_46_46_46'	0

'beta_47_47_47'	0.49609
'beta_48_48_48'	0
'beta_49_49_49'	0
'beta_50_50_50'	0.57743

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Derivation for BIC, AIC

data y β_j β_{c-j} $\beta_j = 0$

Derivation for AIC, BIC

Denominator

$$e^{-\frac{1}{2}(y^T y - 2\beta_{c-j}^T x_{-j} y)}$$

$$e^{-\frac{1}{2}(y - x_{c-j} \beta_{c-j} - x_j \beta_j)^T (y - x_{c-j} \beta_{c-j} - x_j \beta_j)}$$

$$= e^{-\frac{1}{2}(\underbrace{y^T y}_{\text{data}} - \underbrace{2\beta_{c-j}^T x_{-j}^T y}_{\text{data}} - 2\beta_j^T x_j^T y + \underbrace{\beta_{c-j}^T x_{-j}^T x_{-j} \beta_{c-j}}_{\text{data}} + \underbrace{\beta_j^T x_j^T x_j \beta_j}_{\text{data}} + \underbrace{\beta_{c-j}^T x_{-j}^T x_j \beta_j}_{\text{data}} + \underbrace{\beta_j^T x_j^T x_{-j} \beta_{c-j}}_{\text{data}})}$$

Numerator

$$e^{-\frac{1}{2}(y - x_{c-j} \beta_{c-j} - x_j \beta_j)^T (y - x_{c-j} \beta_{c-j} - x_j \beta_j)}$$

$$= e^{-\frac{1}{2}(y^T y - 2\beta_{c-j}^T x_{-j}^T y + \beta_j^T x_j^T x_j \beta_j)}$$

$$= e^{-K}$$

Denominator

$$= e^{-K} \cdot e^{-\frac{1}{2}(-2\beta_j^T x_j^T y + 2\beta_j^T x_j^T x_j \beta_j + \beta_{c-j}^T x_{-j}^T x_j \beta_{c-j} + \beta_j^T x_j^T x_{-j} \beta_{c-j})}$$

Likelihood ratio

$$-2 \log \frac{L(\hat{\beta} | Y, N, X)}{L(\beta | Y, X, N_{(0)})}$$

without j
log's

$$= -2 \log \frac{e^{-K}}{e^{-K} \cdot e^{-\frac{1}{2} \left[\beta_j^T \left(\frac{X_j^T X_j}{\sigma^2} \right) \beta_j - 2 \beta_j^T \left(\frac{X_j^T (Y - X_j \beta_j)}{\sigma^2} \right) \right]}}$$

$$= -2 \log \left[\frac{1}{e^{-\frac{1}{2} [\beta_j^T V^{-1} \beta_j - 2 \beta_j^T V^{-1} E] + E^T V^{-1} E}} e^{\frac{1}{2} E^T V^{-1} E} \right]$$

$$= -2 \log \left[\frac{\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} E^T V^{-1} E}}{\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} [\beta_j^T V^{-1} \beta_j - 2 \beta_j^T V^{-1} E + E^T V^{-1} E]}} \right]$$

$$= -2 \log \left[\frac{N(0 | E, V)}{N(\beta_j | E, V)} \right] \quad \begin{matrix} V = \left(\frac{X_j^T X_j}{\sigma^2} \right)^{-1} \\ K = \left(\frac{X_j^T X_j}{\sigma^2} \right)^{-1} X_j^T (Y - X_j \beta_j) \end{matrix}$$

$$\Rightarrow AIC = -2 \log \left[\frac{N(0 | E, V)}{N(\beta_j | E, V)} \right] + 2(\Delta \text{Parameters})$$

$$BIC = -2 \log \left[\frac{N(0 | E, V)}{N(\beta_j | E, V)} \right] + \log N(\Delta \text{Parameters})$$

```

clear all
clc
close all
data=xlsread('model1.xlsx');
%% SVSS algorithm initialization

data_available.response=data(:,1);
init_data.predictor=data(:,2:end);
init_data.order=3;
init_data=generate_data(init_data);
% init_data.predictor_column=[ones(length(data_available.response),1)
init_data.predictor_column];
% data_available.response=zscore(init_data.response);
% data_available.predictor_library=init_data.predictor_column;
data_available.predictor_library=zscore(init_data.predictor_column);
data_available.predictor_library=[ones(length(data_available.response),1)
data_available.predictor_library];

data_available.init_pi0=0.5;
data_available.Xvar=1.96^2;
% M1=[10:-1:2]%init_data.actualmodelterms;
% M2=[10:-1:2 1];

% M2=[10:-1:2 1];
%
%
%
% [bf,pr]=compute_bayes_fac(M1,M2,data_available);
% pr.m1_m2
%% MCMC starting

disp('Gibbs sampling running');
disp('%%%%%%%%%%%%');

T=500;flg=0;
p=size(data_available.predictor_library,2);
model.heatmap=cell(T,1);
model.global_indicator=cell(T,1);
beta_init=pinv(data_available.predictor_library'*data_available.predictor_library)*data_available.predictor_library'*data_available.response;
model.global_indicator{1,1}=beta_init';
model.selected=cell(p,1);
model.sumterms=zeros(p,1);
model.betaind=ones(1,p);
N=length(data_available.response);
list_var=1:1:p;data_available.var_Y=var(data_available.response);
for t=2:T

    model.indicator=model.global_indicator{t-1,1};

```

```

data_available.beta=model.indicator;
if (rem(t*100/T,20)==0)
    flg=flg+1;
    fprintf('Gibbs sampling: %d percent done\n',20*flg);
end

for j=1:p

    data_available.beta=zeros(1,p);
    model.betaind(j)=1;
    X_new=data_available.predictor_library(:,(model.betaind==1));
    model.indicator=pinv(X_new'*X_new)*X_new'*data_available.response;
    model.indicator=model.indicator';
    data_available.beta(1,model.betaind==1)=model.indicator;

    M1=j;
    M2=find(list_var~=j);
    [del_BIC,pr]=compute_BIC(M1,M2,data_available);

    if del_BIC<0
        model.betaind(1,j)=0;
    else
        model.betaind(1,j)=1;
    end

end

data_available.beta=zeros(1,p);
X_new=data_available.predictor_library(:,(model.betaind==1));
model.indicator=pinv(X_new'*X_new)*X_new'*data_available.response;
model.indicator=model.indicator';
data_available.beta(1,model.betaind==1)=model.indicator;

model.global_indicator{t,1}=data_available.beta;
model.global_betaindicator{t,1}=model.betaind;
model.heatmap{t,1}=find(model.indicator~=0);

model.residuals=data_available.response-
data_available.predictor_library*data_available.beta';
model.B=(sum(model.residuals.^2))*0.5;
model.phi(t,1)=gamrnd(N/2,1/model.B);
data_available.var_Y=1/model.phi(t,1);

if flg>1
if isempty(model.heatmap{t,1})==0
%     length(model.selected(length(model.heatmap{t,1})))=len;
    model.selected{length(model.heatmap{t,1}),1}(1,end+1)=t;

```



```

model.sumterms(length(model.heatmap{t,1}),1)=model.sumterms(length(model.heatmap{t,1}),1)+1;
    end
    end
end
%%
[~,Ind] = max(model.sumterms);newjj=1;jj=1;member=[];clstr=0;
while(jj<length(model.selected{Ind,1}))
    tterm=model.selected{Ind,1}(1,jj);cntr=0;
    totmember=0;clstr=clstr+1;newjj=[];
    for kk=jj:length(model.selected{Ind,1})
        tterm2=model.selected{Ind,1}(1,kk);
        comp_model=model.global_betaindicator{tterm2,1};
        if(ismember(comp_model,model.global_betaindicator{tterm,1}, 'rows'))
            totmember=totmember+1;
            member{clstr,1}(totmember)=tterm2;
            clstrlen(clstr,1)=totmember;
        else
            %           clstr=clstr+1
            newjj(1,end+1)=kk;
        end
    end
    if isempty(newjj)
        break
    else
        jj=newjj(1);
    end
end
[~,indc]=max(clstrlen);
model.betarray=cell2mat(model.global_indicator(member{indc,1}));
model.meanbetarray=mean(model.betarray);
disp(model.meanbetarray);

```

```

function [delta_BIC,pr]=compute_BIC(M1,M2,data_available)
    Y=data_available.response;
    pi0=data_available.init_pi0;
    Y_var=data_available.var_Y;
    N=length(data_available.response);
    X_variance=data_available.Xvar;
    X_M1=data_available.predictor_library(:,M1); %X_j
    X_M2=data_available.predictor_library(:,M2); %X_(-j)
    Beta_M2=data_available.beta(:,M2)';
    Beta_M1=data_available.beta(:,M1)';

    Variance=1/((1/Y_var)*(X_M1'*X_M1));
    Meanval=Variance*X_M1'*(Y-X_M2*Beta_M2)/Y_var;

    log_num=log(1/sqrt(Variance))-0.5*(Meanval^2)/Variance;

```

```
log_denom=log(normpdf(Beta_M1,Meanval,sqrt(Variance)));

delta_BIC=-2*(log_num-log_denom)+log(N)*-1;
pr.E=Meanval;
pr.V=Variance;
end
```

Results of BIC for Model 1

```

clear beta_var
beta_var{1}='Intercept';
for ii=1:50
    beta_var{1+ii}=(sprintf('beta-%d',ii));
end
flg=51;
for ii=1:50
    for jj=ii:ii
        flg=flg+1;
        beta_var{flg}=(sprintf('beta_%d_%d',ii,jj));
    end
end

flg=101;
for ii=1:50
    for jj=ii:ii
        for kk=jj:jj
            flg=flg+1;
            beta_var{flg}=(sprintf('beta_%d_%d_%d',ii,jj,kk));
        end
    end
end
beta_var=beta_var';
VarNames={'Beta','Value'};
T=table(beta_var,model.meanbetarray','VariableNames',VarNames);
disp(T)

```

Beta	Value
'Intercept'	-2.8062
'beta-1'	0
'beta-2'	0
'beta-3'	0
'beta-4'	0
'beta-5'	0
'beta-6'	0
'beta-7'	0
'beta-8'	0
'beta-9'	0
'beta-10'	0
'beta-11'	0
'beta-12'	0
'beta-13'	0
'beta-14'	0
'beta-15'	0
'beta-16'	0
'beta-17'	0
'beta-18'	0
'beta-19'	0
'beta-20'	0
'beta-21'	0
'beta-22'	0
'beta-23'	0
'beta-24'	0

'beta-25'	0
'beta-26'	0
'beta-27'	0
'beta-28'	0
'beta-29'	0
'beta-30'	0
'beta-31'	0
'beta-32'	0
'beta-33'	0
'beta-34'	0
'beta-35'	0
'beta-36'	0
'beta-37'	0
'beta-38'	0
'beta-39'	0
'beta-40'	0
'beta-41'	0
'beta-42'	0
'beta-43'	0
'beta-44'	0
'beta-45'	0
'beta-46'	0
'beta-47'	0
'beta-48'	0
'beta-49'	0
'beta-50'	0
'beta_1_1'	0
'beta_2_2'	0
'beta_3_3'	0
'beta_4_4'	0
'beta_5_5'	0
'beta_6_6'	0
'beta_7_7'	0
'beta_8_8'	0
'beta_9_9'	0
'beta_10_10'	0
'beta_11_11'	0
'beta_12_12'	0
'beta_13_13'	0
'beta_14_14'	0
'beta_15_15'	0
'beta_16_16'	0
'beta_17_17'	0
'beta_18_18'	0
'beta_19_19'	0
'beta_20_20'	0
'beta_21_21'	0
'beta_22_22'	0
'beta_23_23'	0
'beta_24_24'	0
'beta_25_25'	0
'beta_26_26'	0
'beta_27_27'	0
'beta_28_28'	0
'beta_29_29'	0
'beta_30_30'	0
'beta_31_31'	0
'beta_32_32'	0
'beta_33_33'	0
'beta_34_34'	0
'beta_35_35'	0

'beta_36_36'	0
'beta_37_37'	0
'beta_38_38'	0
'beta_39_39'	0
'beta_40_40'	1.0361
'beta_41_41'	0
'beta_42_42'	0
'beta_43_43'	0
'beta_44_44'	0
'beta_45_45'	0
'beta_46_46'	-2.5469
'beta_47_47'	0
'beta_48_48'	0
'beta_49_49'	0
'beta_50_50'	0
'beta_1_1_1'	0
'beta_2_2_2'	0
'beta_3_3_3'	0
'beta_4_4_4'	0
'beta_5_5_5'	0
'beta_6_6_6'	0
'beta_7_7_7'	0
'beta_8_8_8'	0
'beta_9_9_9'	0
'beta_10_10_10'	0
'beta_11_11_11'	0
'beta_12_12_12'	0
'beta_13_13_13'	1.0432
'beta_14_14_14'	0
'beta_15_15_15'	0
'beta_16_16_16'	1.2129
'beta_17_17_17'	0
'beta_18_18_18'	0
'beta_19_19_19'	0
'beta_20_20_20'	0
'beta_21_21_21'	0
'beta_22_22_22'	0.46931
'beta_23_23_23'	0.85011
'beta_24_24_24'	0
'beta_25_25_25'	1.3003
'beta_26_26_26'	-0.48432
'beta_27_27_27'	0
'beta_28_28_28'	-0.77138
'beta_29_29_29'	0
'beta_30_30_30'	0
'beta_31_31_31'	0
'beta_32_32_32'	0
'beta_33_33_33'	0
'beta_34_34_34'	0
'beta_35_35_35'	0
'beta_36_36_36'	0
'beta_37_37_37'	0
'beta_38_38_38'	0
'beta_39_39_39'	0
'beta_40_40_40'	2.8751
'beta_41_41_41'	0
'beta_42_42_42'	0
'beta_43_43_43'	0
'beta_44_44_44'	0
'beta_45_45_45'	0
'beta_46_46_46'	0

'beta_47_47_47'	0
'beta_48_48_48'	0
'beta_49_49_49'	0
'beta_50_50_50'	2.2418

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Results of AIC for Model 1

```

clear beta_var
beta_var{1}='Intercept';
for ii=1:50
    beta_var{1+ii}=(sprintf('beta-%d',ii));
end
flg=51;
for ii=1:50
    for jj=ii:ii
        flg=flg+1;
        beta_var{flg}=(sprintf('beta_%d_%d',ii,jj));
    end
end

flg=101;
for ii=1:50
    for jj=ii:ii
        for kk=jj:jj
            flg=flg+1;
            beta_var{flg}=(sprintf('beta_%d_%d_%d',ii,jj,kk));
        end
    end
end
beta_var=beta_var';
VarNames={'Beta','Value'};
T=table(beta_var,model.meanbetarray','VariableNames',VarNames);
disp(T)

```

Beta	Value
'Intercept'	-2.8062
'beta-1'	0
'beta-2'	0
'beta-3'	0
'beta-4'	0
'beta-5'	0
'beta-6'	0
'beta-7'	0
'beta-8'	0
'beta-9'	0
'beta-10'	0
'beta-11'	0
'beta-12'	0
'beta-13'	0
'beta-14'	0
'beta-15'	0
'beta-16'	0
'beta-17'	0
'beta-18'	0
'beta-19'	0
'beta-20'	0
'beta-21'	0
'beta-22'	0
'beta-23'	0
'beta-24'	0

'beta-25'	0
'beta-26'	0
'beta-27'	0
'beta-28'	0
'beta-29'	0
'beta-30'	0
'beta-31'	0
'beta-32'	0
'beta-33'	0
'beta-34'	0
'beta-35'	0
'beta-36'	0
'beta-37'	0
'beta-38'	0
'beta-39'	0
'beta-40'	0
'beta-41'	0
'beta-42'	0
'beta-43'	0
'beta-44'	0
'beta-45'	4.2605
'beta-46'	0
'beta-47'	0
'beta-48'	0
'beta-49'	0
'beta-50'	0
'beta_1_1'	0
'beta_2_2'	0
'beta_3_3'	0
'beta_4_4'	0
'beta_5_5'	0
'beta_6_6'	0
'beta_7_7'	0
'beta_8_8'	0
'beta_9_9'	0
'beta_10_10'	0
'beta_11_11'	0
'beta_12_12'	0
'beta_13_13'	0
'beta_14_14'	0
'beta_15_15'	0
'beta_16_16'	0
'beta_17_17'	0
'beta_18_18'	0
'beta_19_19'	0
'beta_20_20'	0
'beta_21_21'	0
'beta_22_22'	0
'beta_23_23'	0
'beta_24_24'	0
'beta_25_25'	0
'beta_26_26'	0
'beta_27_27'	0
'beta_28_28'	0
'beta_29_29'	0
'beta_30_30'	0
'beta_31_31'	0
'beta_32_32'	0
'beta_33_33'	0
'beta_34_34'	0
'beta_35_35'	0

'beta_36_36'	0
'beta_37_37'	0
'beta_38_38'	0
'beta_39_39'	0
'beta_40_40'	0
'beta_41_41'	0
'beta_42_42'	0
'beta_43_43'	0
'beta_44_44'	0
'beta_45_45'	0
'beta_46_46'	-1.0295
'beta_47_47'	0.087889
'beta_48_48'	0
'beta_49_49'	0
'beta_50_50'	0
'beta_1_1_1'	0
'beta_2_2_2'	0
'beta_3_3_3'	0.8069
'beta_4_4_4'	2.0415
'beta_5_5_5'	0.22913
'beta_6_6_6'	0
'beta_7_7_7'	0
'beta_8_8_8'	-0.70745
'beta_9_9_9'	0.24963
'beta_10_10_10'	0.25289
'beta_11_11_11'	-1.1552
'beta_12_12_12'	1.4891
'beta_13_13_13'	0.64603
'beta_14_14_14'	0
'beta_15_15_15'	1.2087
'beta_16_16_16'	0.61023
'beta_17_17_17'	0.23245
'beta_18_18_18'	-0.3791
'beta_19_19_19'	0
'beta_20_20_20'	-1.0719
'beta_21_21_21'	0
'beta_22_22_22'	0
'beta_23_23_23'	-0.52537
'beta_24_24_24'	-0.19916
'beta_25_25_25'	0.11549
'beta_26_26_26'	0.73257
'beta_27_27_27'	0
'beta_28_28_28'	0
'beta_29_29_29'	0
'beta_30_30_30'	0
'beta_31_31_31'	-0.050845
'beta_32_32_32'	0.3242
'beta_33_33_33'	0
'beta_34_34_34'	0
'beta_35_35_35'	0
'beta_36_36_36'	0
'beta_37_37_37'	0
'beta_38_38_38'	0
'beta_39_39_39'	0
'beta_40_40_40'	0
'beta_41_41_41'	0
'beta_42_42_42'	0
'beta_43_43_43'	0
'beta_44_44_44'	0
'beta_45_45_45'	0
'beta_46_46_46'	0

'beta_47_47_47'	0
'beta_48_48_48'	0
'beta_49_49_49'	0
'beta_50_50_50'	0

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PROBLEM 3

Results of SVSS for MODEL 2

```

clear beta_var
beta_var{1}='Intercept';
for ii=1:50
    beta_var{1+ii}=(sprintf('beta-%d',ii));
end
flg=51;
for ii=1:50
    for jj=ii:ii
        flg=flg+1;
        beta_var{flg}=(sprintf('beta_%d_%d',ii,jj));
    end
end

flg=101;
for ii=1:50
    for jj=ii:ii
        for kk=jj:jj
            flg=flg+1;
            beta_var{flg}=(sprintf('beta_%d_%d_%d',ii,jj,kk));
        end
    end
end
beta_var=beta_var';
VarNames={'Beta','Value'};
T=table(beta_var,model.meanbetarray','VariableNames',VarNames);
disp(T)

```

Beta	Value
'Intercept'	-3.5022
'beta-1'	0
'beta-2'	0
'beta-3'	0
'beta-4'	0
'beta-5'	1.7549
'beta-6'	0
'beta-7'	0
'beta-8'	0
'beta-9'	0
'beta-10'	0
'beta-11'	0
'beta-12'	0
'beta-13'	0
'beta-14'	0
'beta-15'	-0.18829
'beta-16'	0
'beta-17'	0
'beta-18'	0
'beta-19'	0
'beta-20'	0.78991
'beta-21'	0
'beta-22'	0.40876
'beta-23'	0
'beta-24'	0

'beta-25'	0
'beta-26'	0
'beta-27'	1.8254
'beta-28'	-0.6037
'beta-29'	0
'beta-30'	0
'beta-31'	0
'beta-32'	0
'beta-33'	-0.045552
'beta-34'	0
'beta-35'	0
'beta-36'	0.64016
'beta-37'	0
'beta-38'	0
'beta-39'	-0.082206
'beta-40'	0
'beta-41'	-0.064825
'beta-42'	0
'beta-43'	0
'beta-44'	0
'beta-45'	4.7117
'beta-46'	0
'beta-47'	0
'beta-48'	0
'beta-49'	-0.081427
'beta-50'	0.031346
'beta_1_1'	0.49304
'beta_2_2'	0
'beta_3_3'	0
'beta_4_4'	0
'beta_5_5'	0
'beta_6_6'	0
'beta_7_7'	0
'beta_8_8'	0
'beta_9_9'	-0.2001
'beta_10_10'	0
'beta_11_11'	0
'beta_12_12'	0
'beta_13_13'	0
'beta_14_14'	0
'beta_15_15'	0
'beta_16_16'	0
'beta_17_17'	0
'beta_18_18'	0
'beta_19_19'	0
'beta_20_20'	0
'beta_21_21'	-0.25226
'beta_22_22'	0
'beta_23_23'	0
'beta_24_24'	0
'beta_25_25'	0
'beta_26_26'	0
'beta_27_27'	0
'beta_28_28'	0.10562
'beta_29_29'	0
'beta_30_30'	0
'beta_31_31'	0
'beta_32_32'	0.45424
'beta_33_33'	0
'beta_34_34'	0
'beta_35_35'	0

'beta_36_36'	0
'beta_37_37'	0
'beta_38_38'	0
'beta_39_39'	0
'beta_40_40'	-0.87016
'beta_41_41'	0
'beta_42_42'	0
'beta_43_43'	0
'beta_44_44'	-0.28638
'beta_45_45'	0
'beta_46_46'	0
'beta_47_47'	0
'beta_48_48'	0
'beta_49_49'	0
'beta_50_50'	-0.14452
'beta_1_1_1'	0
'beta_2_2_2'	-0.054628
'beta_3_3_3'	0
'beta_4_4_4'	0
'beta_5_5_5'	0
'beta_6_6_6'	0.47466
'beta_7_7_7'	0
'beta_8_8_8'	0
'beta_9_9_9'	0
'beta_10_10_10'	0
'beta_11_11_11'	0
'beta_12_12_12'	0
'beta_13_13_13'	0
'beta_14_14_14'	0
'beta_15_15_15'	0
'beta_16_16_16'	0
'beta_17_17_17'	-0.4108
'beta_18_18_18'	0
'beta_19_19_19'	0.77229
'beta_20_20_20'	0
'beta_21_21_21'	0
'beta_22_22_22'	-0.56375
'beta_23_23_23'	-0.25808
'beta_24_24_24'	-0.17031
'beta_25_25_25'	0
'beta_26_26_26'	0
'beta_27_27_27'	0
'beta_28_28_28'	0
'beta_29_29_29'	0
'beta_30_30_30'	0
'beta_31_31_31'	0
'beta_32_32_32'	0
'beta_33_33_33'	0
'beta_34_34_34'	0
'beta_35_35_35'	0
'beta_36_36_36'	0
'beta_37_37_37'	0
'beta_38_38_38'	0
'beta_39_39_39'	0
'beta_40_40_40'	0
'beta_41_41_41'	0
'beta_42_42_42'	0
'beta_43_43_43'	0
'beta_44_44_44'	0
'beta_45_45_45'	0.12778
'beta_46_46_46'	0

'beta_47_47_47'	0.282
'beta_48_48_48'	0
'beta_49_49_49'	0
'beta_50_50_50'	0

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Results of AIC for Model 2

```

clear beta_var
beta_var{1}='Intercept';
for ii=1:50
    beta_var{1+ii}=(sprintf('beta-%d',ii));
end
flg=51;
for ii=1:50
    for jj=ii:ii
        flg=flg+1;
        beta_var{flg}=(sprintf('beta_%d_%d',ii,jj));
    end
end

flg=101;
for ii=1:50
    for jj=ii:ii
        for kk=jj:jj
            flg=flg+1;
            beta_var{flg}=(sprintf('beta_%d_%d_%d',ii,jj,kk));
        end
    end
end
beta_var=beta_var';
VarNames={'Beta','Value'};
T=table(beta_var,model.meanbetarray','VariableNames',VarNames);
disp(T)

```

Beta	Value
'Intercept'	-3.5017
'beta-1'	0
'beta-2'	0
'beta-3'	0
'beta-4'	0
'beta-5'	0
'beta-6'	0
'beta-7'	0
'beta-8'	0
'beta-9'	0
'beta-10'	0
'beta-11'	0
'beta-12'	0
'beta-13'	0
'beta-14'	0
'beta-15'	0
'beta-16'	0
'beta-17'	0
'beta-18'	0
'beta-19'	0
'beta-20'	0
'beta-21'	0
'beta-22'	0
'beta-23'	0
'beta-24'	0

'beta-25'	0
'beta-26'	0
'beta-27'	0
'beta-28'	0
'beta-29'	0
'beta-30'	0
'beta-31'	0
'beta-32'	0
'beta-33'	0
'beta-34'	0
'beta-35'	0
'beta-36'	0
'beta-37'	0
'beta-38'	0
'beta-39'	0
'beta-40'	0
'beta-41'	0
'beta-42'	0
'beta-43'	0
'beta-44'	0
'beta-45'	2.7288
'beta-46'	0
'beta-47'	0
'beta-48'	0
'beta-49'	0
'beta-50'	0
'beta_1_1'	0
'beta_2_2'	0
'beta_3_3'	0
'beta_4_4'	0
'beta_5_5'	0
'beta_6_6'	0
'beta_7_7'	0
'beta_8_8'	0
'beta_9_9'	0
'beta_10_10'	0
'beta_11_11'	0
'beta_12_12'	0
'beta_13_13'	0
'beta_14_14'	0
'beta_15_15'	0
'beta_16_16'	0
'beta_17_17'	0
'beta_18_18'	0
'beta_19_19'	0
'beta_20_20'	0
'beta_21_21'	0
'beta_22_22'	0
'beta_23_23'	0
'beta_24_24'	0
'beta_25_25'	0
'beta_26_26'	-1.2892
'beta_27_27'	0
'beta_28_28'	0
'beta_29_29'	0
'beta_30_30'	0
'beta_31_31'	0
'beta_32_32'	0
'beta_33_33'	0
'beta_34_34'	0
'beta_35_35'	0

'beta_36_36'	0
'beta_37_37'	0
'beta_38_38'	0
'beta_39_39'	0
'beta_40_40'	0
'beta_41_41'	0
'beta_42_42'	0
'beta_43_43'	0
'beta_44_44'	0
'beta_45_45'	0
'beta_46_46'	0
'beta_47_47'	0.83638
'beta_48_48'	0
'beta_49_49'	0
'beta_50_50'	0
'beta_1_1_1'	0
'beta_2_2_2'	-0.0091925
'beta_3_3_3'	-1.4423
'beta_4_4_4'	0
'beta_5_5_5'	0
'beta_6_6_6'	0.86654
'beta_7_7_7'	0
'beta_8_8_8'	-0.96641
'beta_9_9_9'	-0.058233
'beta_10_10_10'	-0.87767
'beta_11_11_11'	0.60232
'beta_12_12_12'	2.093
'beta_13_13_13'	0
'beta_14_14_14'	-0.0030683
'beta_15_15_15'	0.86737
'beta_16_16_16'	-0.80581
'beta_17_17_17'	-0.68928
'beta_18_18_18'	0.44008
'beta_19_19_19'	0
'beta_20_20_20'	1.6519
'beta_21_21_21'	-0.089795
'beta_22_22_22'	-0.74024
'beta_23_23_23'	1.1953
'beta_24_24_24'	-1.3925
'beta_25_25_25'	-0.53152
'beta_26_26_26'	-0.68827
'beta_27_27_27'	0.088807
'beta_28_28_28'	0
'beta_29_29_29'	-1.6681
'beta_30_30_30'	0
'beta_31_31_31'	1.0072
'beta_32_32_32'	0
'beta_33_33_33'	0
'beta_34_34_34'	0
'beta_35_35_35'	0
'beta_36_36_36'	-1.3651
'beta_37_37_37'	-0.75567
'beta_38_38_38'	3.0849
'beta_39_39_39'	0
'beta_40_40_40'	1.451
'beta_41_41_41'	0
'beta_42_42_42'	0
'beta_43_43_43'	-0.47591
'beta_44_44_44'	1.6789
'beta_45_45_45'	0
'beta_46_46_46'	0

'beta_47_47_47'	0
'beta_48_48_48'	0
'beta_49_49_49'	-2.0835
'beta_50_50_50'	0

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Results of BIC for Model 2

```

clear beta_var
beta_var{1}='Intercept';
for ii=1:50
    beta_var{1+ii}=(sprintf('beta-%d',ii));
end
flg=51;
for ii=1:50
    for jj=ii:ii
        flg=flg+1;
        beta_var{flg}=(sprintf('beta_%d_%d',ii,jj));
    end
end

flg=101;
for ii=1:50
    for jj=ii:ii
        for kk=jj:jj
            flg=flg+1;
            beta_var{flg}=(sprintf('beta_%d_%d_%d',ii,jj,kk));
        end
    end
end
beta_var=beta_var';
VarNames={'Beta','Value'};
T=table(beta_var,model.meanbetarray','VariableNames',VarNames);
disp(T)

```

Beta	Value
'Intercept'	-3.5017
'beta-1'	0
'beta-2'	0
'beta-3'	0
'beta-4'	0
'beta-5'	0
'beta-6'	0
'beta-7'	0
'beta-8'	0
'beta-9'	0
'beta-10'	0
'beta-11'	0
'beta-12'	0
'beta-13'	0
'beta-14'	0
'beta-15'	0
'beta-16'	0
'beta-17'	0
'beta-18'	0
'beta-19'	0
'beta-20'	0
'beta-21'	0
'beta-22'	0
'beta-23'	0
'beta-24'	0

'beta-25'	0
'beta-26'	0
'beta-27'	0
'beta-28'	0
'beta-29'	0
'beta-30'	0
'beta-31'	0
'beta-32'	0
'beta-33'	0
'beta-34'	0
'beta-35'	0
'beta-36'	0
'beta-37'	0
'beta-38'	0
'beta-39'	0
'beta-40'	0
'beta-41'	0
'beta-42'	0
'beta-43'	0
'beta-44'	0
'beta-45'	0
'beta-46'	0
'beta-47'	0
'beta-48'	0
'beta-49'	0
'beta-50'	0
'beta_1_1'	0
'beta_2_2'	0
'beta_3_3'	0
'beta_4_4'	0
'beta_5_5'	0
'beta_6_6'	0
'beta_7_7'	0
'beta_8_8'	0
'beta_9_9'	0
'beta_10_10'	0
'beta_11_11'	0
'beta_12_12'	0
'beta_13_13'	0
'beta_14_14'	0
'beta_15_15'	0
'beta_16_16'	0
'beta_17_17'	0
'beta_18_18'	0
'beta_19_19'	0
'beta_20_20'	0
'beta_21_21'	0
'beta_22_22'	0
'beta_23_23'	0
'beta_24_24'	0
'beta_25_25'	0
'beta_26_26'	-1.2153
'beta_27_27'	0
'beta_28_28'	0
'beta_29_29'	0
'beta_30_30'	0
'beta_31_31'	0
'beta_32_32'	0
'beta_33_33'	0
'beta_34_34'	0
'beta_35_35'	0

'beta_36_36'	0
'beta_37_37'	0
'beta_38_38'	0
'beta_39_39'	0.44337
'beta_40_40'	-0.46521
'beta_41_41'	0
'beta_42_42'	0
'beta_43_43'	0
'beta_44_44'	0
'beta_45_45'	0
'beta_46_46'	0
'beta_47_47'	0
'beta_48_48'	0
'beta_49_49'	0
'beta_50_50'	0
'beta_1_1_1'	0
'beta_2_2_2'	0
'beta_3_3_3'	0
'beta_4_4_4'	0
'beta_5_5_5'	0
'beta_6_6_6'	0.3798
'beta_7_7_7'	0.74895
'beta_8_8_8'	-0.43907
'beta_9_9_9'	0
'beta_10_10_10'	-0.56148
'beta_11_11_11'	0.74686
'beta_12_12_12'	0
'beta_13_13_13'	0
'beta_14_14_14'	0
'beta_15_15_15'	1.134
'beta_16_16_16'	0
'beta_17_17_17'	0
'beta_18_18_18'	0
'beta_19_19_19'	0
'beta_20_20_20'	1.3983
'beta_21_21_21'	0
'beta_22_22_22'	0
'beta_23_23_23'	0.60994
'beta_24_24_24'	0
'beta_25_25_25'	0
'beta_26_26_26'	0
'beta_27_27_27'	0
'beta_28_28_28'	0
'beta_29_29_29'	-0.4576
'beta_30_30_30'	0
'beta_31_31_31'	0
'beta_32_32_32'	0
'beta_33_33_33'	0
'beta_34_34_34'	0
'beta_35_35_35'	0
'beta_36_36_36'	0
'beta_37_37_37'	0
'beta_38_38_38'	0
'beta_39_39_39'	0
'beta_40_40_40'	4.3362
'beta_41_41_41'	0
'beta_42_42_42'	0
'beta_43_43_43'	0
'beta_44_44_44'	0
'beta_45_45_45'	0
'beta_46_46_46'	0

'beta_47_47_47'	0
'beta_48_48_48'	0
'beta_49_49_49'	0
'beta_50_50_50'	0

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PROBLEM 4

Results of SVSS for MODEL 3

```

clear beta_var
beta_var{1}='Intercept';
for ii=1:50
    beta_var{1+ii}=(sprintf('beta-%d',ii));
end
flg=51;
for ii=1:50
    for jj=ii:ii
        flg=flg+1;
        beta_var{flg}=(sprintf('beta_%d_%d',ii,jj));
    end
end

flg=101;
for ii=1:50
    for jj=ii:ii
        for kk=jj:jj
            flg=flg+1;
            beta_var{flg}=(sprintf('beta_%d_%d_%d',ii,jj,kk));
        end
    end
end
beta_var=beta_var';
VarNames={'Beta','Value'};
T=table(beta_var,model.meanbetarray','VariableNames',VarNames);
disp(T)

```

Beta	Value
'Intercept'	-2.4608
'beta-1'	0
'beta-2'	0
'beta-3'	1.444
'beta-4'	0
'beta-5'	0
'beta-6'	0
'beta-7'	0
'beta-8'	0
'beta-9'	0
'beta-10'	-0.50624
'beta-11'	0
'beta-12'	-0.12246
'beta-13'	0
'beta-14'	-0.21602
'beta-15'	0
'beta-16'	0
'beta-17'	1.0265
'beta-18'	-0.70354
'beta-19'	0
'beta-20'	0
'beta-21'	0
'beta-22'	0
'beta-23'	0
'beta-24'	0

'beta-25'	0.65805
'beta-26'	0
'beta-27'	1.5223
'beta-28'	0
'beta-29'	0
'beta-30'	0.10231
'beta-31'	0
'beta-32'	0
'beta-33'	0
'beta-34'	0
'beta-35'	0
'beta-36'	-0.22166
'beta-37'	0
'beta-38'	0
'beta-39'	0
'beta-40'	0
'beta-41'	0
'beta-42'	0
'beta-43'	0
'beta-44'	0
'beta-45'	4.8476
'beta-46'	0
'beta-47'	0
'beta-48'	0
'beta-49'	-0.12472
'beta-50'	0
'beta_1_1'	-0.21892
'beta_2_2'	0
'beta_3_3'	0
'beta_4_4'	0
'beta_5_5'	0
'beta_6_6'	0
'beta_7_7'	0
'beta_8_8'	0
'beta_9_9'	0
'beta_10_10'	0
'beta_11_11'	0
'beta_12_12'	0
'beta_13_13'	-0.12009
'beta_14_14'	0
'beta_15_15'	0.11178
'beta_16_16'	0
'beta_17_17'	0
'beta_18_18'	0
'beta_19_19'	0
'beta_20_20'	0.3688
'beta_21_21'	-0.50193
'beta_22_22'	0
'beta_23_23'	0
'beta_24_24'	0
'beta_25_25'	0.72544
'beta_26_26'	0
'beta_27_27'	0
'beta_28_28'	0
'beta_29_29'	0
'beta_30_30'	0
'beta_31_31'	0
'beta_32_32'	0
'beta_33_33'	0
'beta_34_34'	0
'beta_35_35'	0

'beta_36_36'	0.32062
'beta_37_37'	0
'beta_38_38'	0
'beta_39_39'	0
'beta_40_40'	0
'beta_41_41'	0
'beta_42_42'	0
'beta_43_43'	0.056383
'beta_44_44'	0
'beta_45_45'	0
'beta_46_46'	0
'beta_47_47'	0
'beta_48_48'	0
'beta_49_49'	0
'beta_50_50'	-0.63091
'beta_1_1_1'	0
'beta_2_2_2'	0
'beta_3_3_3'	0
'beta_4_4_4'	0
'beta_5_5_5'	0
'beta_6_6_6'	0
'beta_7_7_7'	0
'beta_8_8_8'	-0.10892
'beta_9_9_9'	0
'beta_10_10_10'	0
'beta_11_11_11'	0
'beta_12_12_12'	0
'beta_13_13_13'	0
'beta_14_14_14'	0
'beta_15_15_15'	-0.42356
'beta_16_16_16'	0
'beta_17_17_17'	0
'beta_18_18_18'	0
'beta_19_19_19'	0
'beta_20_20_20'	0
'beta_21_21_21'	0
'beta_22_22_22'	0
'beta_23_23_23'	0
'beta_24_24_24'	0
'beta_25_25_25'	0
'beta_26_26_26'	0
'beta_27_27_27'	0
'beta_28_28_28'	0
'beta_29_29_29'	0
'beta_30_30_30'	0
'beta_31_31_31'	0
'beta_32_32_32'	0
'beta_33_33_33'	0
'beta_34_34_34'	0
'beta_35_35_35'	0
'beta_36_36_36'	0
'beta_37_37_37'	-0.11843
'beta_38_38_38'	0
'beta_39_39_39'	0
'beta_40_40_40'	0
'beta_41_41_41'	0
'beta_42_42_42'	0
'beta_43_43_43'	0
'beta_44_44_44'	0.3025
'beta_45_45_45'	0.052739
'beta_46_46_46'	0

'beta_47_47_47'	0
'beta_48_48_48'	0
'beta_49_49_49'	0
'beta_50_50_50'	0.32223

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Results of BIC for Model 3

```

clear beta_var
beta_var{1}='Intercept';
for ii=1:50
    beta_var{1+ii}=(sprintf('beta-%d',ii));
end
flg=51;
for ii=1:50
    for jj=ii:ii
        flg=flg+1;
        beta_var{flg}=(sprintf('beta_%d_%d',ii,jj));
    end
end

flg=101;
for ii=1:50
    for jj=ii:ii
        for kk=jj:jj
            flg=flg+1;
            beta_var{flg}=(sprintf('beta_%d_%d_%d',ii,jj,kk));
        end
    end
end
beta_var=beta_var';
VarNames={'Beta','Value'};
T=table(beta_var,model.meanbetarray','VariableNames',VarNames);
disp(T)

```

Beta	Value
'Intercept'	-2.4541
'beta-1'	0
'beta-2'	0
'beta-3'	0
'beta-4'	0
'beta-5'	0
'beta-6'	0
'beta-7'	0
'beta-8'	0
'beta-9'	0
'beta-10'	0
'beta-11'	0
'beta-12'	0
'beta-13'	0
'beta-14'	0
'beta-15'	0
'beta-16'	0
'beta-17'	0
'beta-18'	0
'beta-19'	0
'beta-20'	0
'beta-21'	0
'beta-22'	0
'beta-23'	0
'beta-24'	0

'beta-25'	0
'beta-26'	0
'beta-27'	0
'beta-28'	0
'beta-29'	0
'beta-30'	0
'beta-31'	0
'beta-32'	0
'beta-33'	0
'beta-34'	0
'beta-35'	0
'beta-36'	0
'beta-37'	0
'beta-38'	0
'beta-39'	0
'beta-40'	0
'beta-41'	0
'beta-42'	0
'beta-43'	0
'beta-44'	0
'beta-45'	4.9432
'beta-46'	0
'beta-47'	0
'beta-48'	0
'beta-49'	0
'beta-50'	0
'beta_1_1'	0
'beta_2_2'	0
'beta_3_3'	0
'beta_4_4'	0
'beta_5_5'	0
'beta_6_6'	0
'beta_7_7'	0
'beta_8_8'	0
'beta_9_9'	0
'beta_10_10'	0
'beta_11_11'	0
'beta_12_12'	0
'beta_13_13'	0
'beta_14_14'	0
'beta_15_15'	0
'beta_16_16'	0
'beta_17_17'	0
'beta_18_18'	0
'beta_19_19'	0
'beta_20_20'	0
'beta_21_21'	0
'beta_22_22'	0
'beta_23_23'	0
'beta_24_24'	0
'beta_25_25'	0
'beta_26_26'	0
'beta_27_27'	0
'beta_28_28'	0
'beta_29_29'	0
'beta_30_30'	0
'beta_31_31'	0
'beta_32_32'	0
'beta_33_33'	0
'beta_34_34'	0
'beta_35_35'	0

'beta_36_36'	0
'beta_37_37'	0
'beta_38_38'	0
'beta_39_39'	0
'beta_40_40'	0
'beta_41_41'	0
'beta_42_42'	0
'beta_43_43'	0
'beta_44_44'	0
'beta_45_45'	-0.70933
'beta_46_46'	0
'beta_47_47'	0
'beta_48_48'	0
'beta_49_49'	0
'beta_50_50'	0
'beta_1_1_1'	0.012363
'beta_2_2_2'	0
'beta_3_3_3'	0
'beta_4_4_4'	0
'beta_5_5_5'	0
'beta_6_6_6'	0
'beta_7_7_7'	0.39368
'beta_8_8_8'	-0.66544
'beta_9_9_9'	0
'beta_10_10_10'	0
'beta_11_11_11'	0
'beta_12_12_12'	0
'beta_13_13_13'	0
'beta_14_14_14'	0
'beta_15_15_15'	0
'beta_16_16_16'	0
'beta_17_17_17'	0
'beta_18_18_18'	0
'beta_19_19_19'	-0.10556
'beta_20_20_20'	0
'beta_21_21_21'	0
'beta_22_22_22'	0
'beta_23_23_23'	0
'beta_24_24_24'	0
'beta_25_25_25'	0
'beta_26_26_26'	0
'beta_27_27_27'	0
'beta_28_28_28'	0
'beta_29_29_29'	0
'beta_30_30_30'	0
'beta_31_31_31'	0
'beta_32_32_32'	0
'beta_33_33_33'	0
'beta_34_34_34'	0
'beta_35_35_35'	0
'beta_36_36_36'	0
'beta_37_37_37'	0
'beta_38_38_38'	0
'beta_39_39_39'	0
'beta_40_40_40'	0
'beta_41_41_41'	0
'beta_42_42_42'	0
'beta_43_43_43'	0
'beta_44_44_44'	0
'beta_45_45_45'	0
'beta_46_46_46'	0

'beta_47_47_47'	0
'beta_48_48_48'	0
'beta_49_49_49'	0
'beta_50_50_50'	0

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Results of AIC for Model 3

```

clear beta_var
beta_var{1}='Intercept';
for ii=1:50
    beta_var{1+ii}=(sprintf('beta-%d',ii));
end
flg=51;
for ii=1:50
    for jj=ii:ii
        flg=flg+1;
        beta_var{flg}=(sprintf('beta_%d_%d',ii,jj));
    end
end

flg=101;
for ii=1:50
    for jj=ii:ii
        for kk=jj:jj
            flg=flg+1;
            beta_var{flg}=(sprintf('beta_%d_%d_%d',ii,jj,kk));
        end
    end
end
beta_var=beta_var';
VarNames={'Beta','Value'};
T=table(beta_var,model.meanbetarray','VariableNames',VarNames);
disp(T)

```

Beta	Value
'Intercept'	-2.4541
'beta-1'	0
'beta-2'	0
'beta-3'	0
'beta-4'	0
'beta-5'	0
'beta-6'	0
'beta-7'	0
'beta-8'	0
'beta-9'	0
'beta-10'	0
'beta-11'	0
'beta-12'	0
'beta-13'	0
'beta-14'	0
'beta-15'	0
'beta-16'	0
'beta-17'	0
'beta-18'	0
'beta-19'	0
'beta-20'	0
'beta-21'	0
'beta-22'	0
'beta-23'	0
'beta-24'	0

'beta-25'	0
'beta-26'	0
'beta-27'	0
'beta-28'	0
'beta-29'	0
'beta-30'	0
'beta-31'	0
'beta-32'	0
'beta-33'	0
'beta-34'	0
'beta-35'	0
'beta-36'	0
'beta-37'	0
'beta-38'	0
'beta-39'	0
'beta-40'	0
'beta-41'	0
'beta-42'	0
'beta-43'	0
'beta-44'	0
'beta-45'	4.9261
'beta-46'	0
'beta-47'	0
'beta-48'	0
'beta-49'	0
'beta-50'	0
'beta_1_1'	0
'beta_2_2'	0
'beta_3_3'	0
'beta_4_4'	0
'beta_5_5'	0
'beta_6_6'	0
'beta_7_7'	0
'beta_8_8'	0
'beta_9_9'	0
'beta_10_10'	0
'beta_11_11'	0
'beta_12_12'	0
'beta_13_13'	0
'beta_14_14'	0
'beta_15_15'	0
'beta_16_16'	0
'beta_17_17'	0
'beta_18_18'	0
'beta_19_19'	0
'beta_20_20'	0
'beta_21_21'	0
'beta_22_22'	0
'beta_23_23'	0
'beta_24_24'	0
'beta_25_25'	0
'beta_26_26'	0
'beta_27_27'	0
'beta_28_28'	0
'beta_29_29'	0
'beta_30_30'	0
'beta_31_31'	0
'beta_32_32'	0
'beta_33_33'	0
'beta_34_34'	0
'beta_35_35'	0

'beta_36_36'	0
'beta_37_37'	0
'beta_38_38'	0
'beta_39_39'	0
'beta_40_40'	0
'beta_41_41'	0
'beta_42_42'	0
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'beta_46_46'	0
'beta_47_47'	0
'beta_48_48'	0
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'beta_2_2_2'	-1.579
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'beta_4_4_4'	0
'beta_5_5_5'	0
'beta_6_6_6'	0.36939
'beta_7_7_7'	0.69033
'beta_8_8_8'	-0.82567
'beta_9_9_9'	0
'beta_10_10_10'	0
'beta_11_11_11'	-0.13872
'beta_12_12_12'	0
'beta_13_13_13'	0
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'beta_15_15_15'	-0.71175
'beta_16_16_16'	0
'beta_17_17_17'	0
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'beta_19_19_19'	0.42133
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'beta_28_28_28'	-0.298
'beta_29_29_29'	0
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'beta_31_31_31'	0
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'beta_49_49_49'	0
'beta_50_50_50'	0

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