Identified the Best Predictors Of Systolic Blood Pressure Through Investigating The Impact Of Different Genes And Environmental Factors

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## Introduction

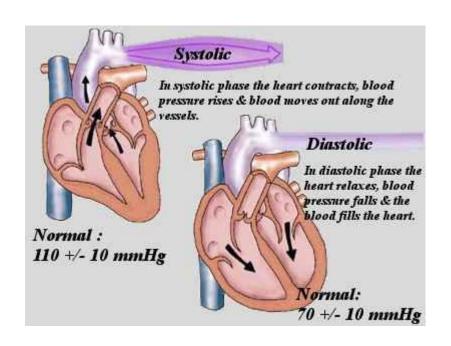
- At GlaxoSmithKline (GSK), a study was conducted for genetic and genomic research to enable the medical community to accurately prescribe the right medicine for the right patient.
- Of the 500 subjects, 250 had blood pressure less than and 250 had high blood pressure. The 501 variables consist of one response variable (systolic blood pressure) and 500 predictors (17 clinical covariates and 483 genetic markers).
- Objective of the present study is to find best predictors for systolic blood pressure.
- We use multiple regression, regularization and logistic regression to analyze the data.

# Background

- One billion people suffer with high blood pressure
- 7.1 million deaths each year
- ▶ 55-year-old with normal blood pressure (BP) has a 90% lifetime prospect of developing hypertension



## **Blood Pressure**



High blood pressure

140/90 mmHg or higher

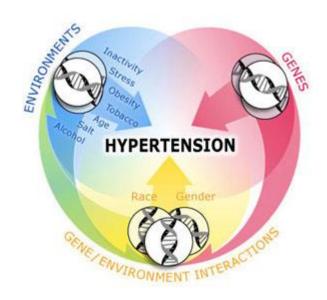
Prehypertension

between 120-139 mmHg and/or 80-89 mmHg

Normal blood pressure

less than 120/80 mmHg

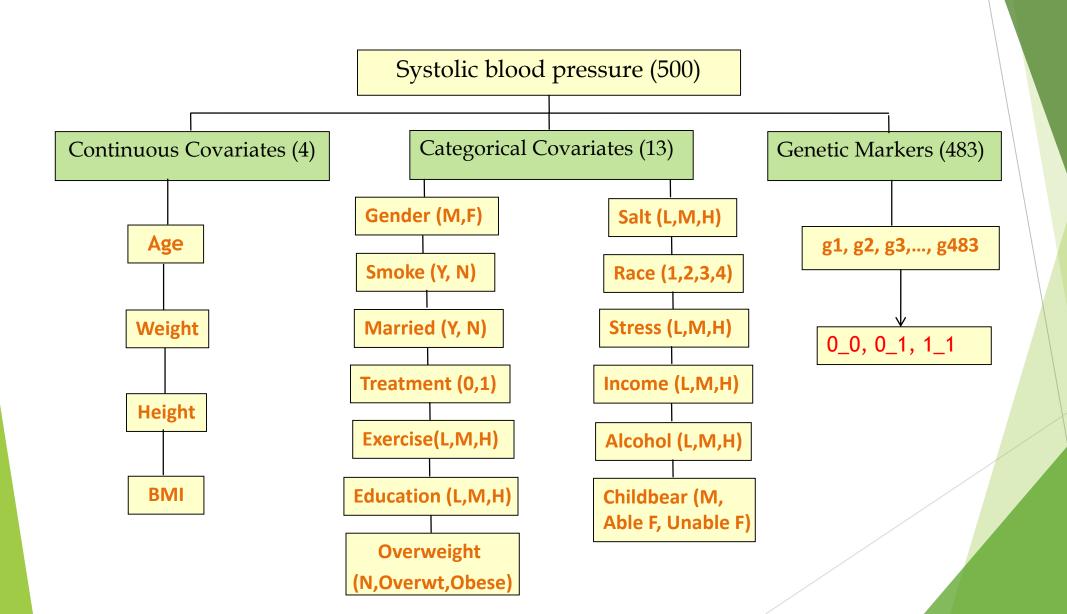
# Causes and effects of high blood pressure



Hypertension and risk factors can lead to

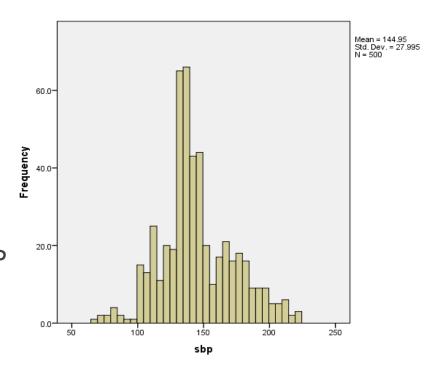
- heart attack
- •stroke
- kidney damage
- vision loss
- memory loss
- •fluid in the lungs

## Variables used in data



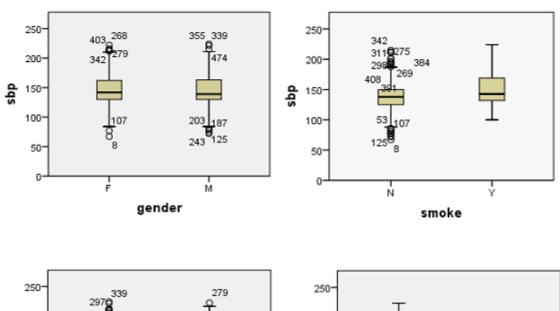
# Response Variable (Blood Pressure)

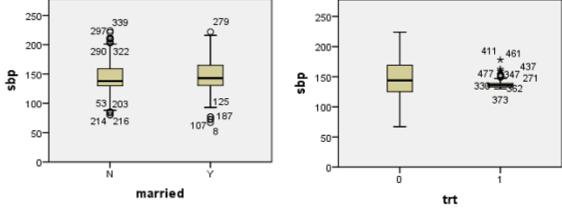
- Of the 500 subjects, 11 are with hypotenstion (SBP less than 90)
- > 72 are with normal blood pressure (SBP less than 120)
- ▶ 167 are with prehypertension (SBP between 120 and 140).
- ▶ 50 percent have high blood pressure (SBP greater than 140) out which 13% can develop life threatening complications (as their SPB is higher than 180).



# **Descriptive Analysis**

Box plots for systolic blood pressure by gender, married, smoke and treatment

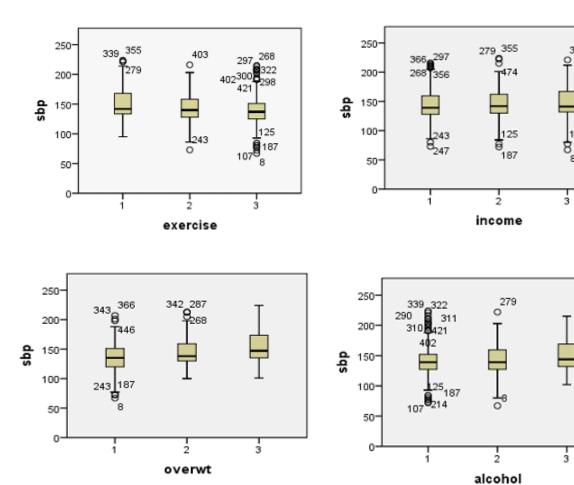




Gender		married		Smoke		Treatment		
	Male	Female	Yes	No	Yes	No	Yes	No
Count	236	264	239	261	266	234	101	399
Mean	145.02	144.89	146.72	143.33	150.03	139.18	137.96	146.72

	t value	Pr(> t )
Smoke	4.4	0.000
Treatment	2.89	0.000

#### Box plots for systolic blood pressure by exercise, income, overweight and alcohal



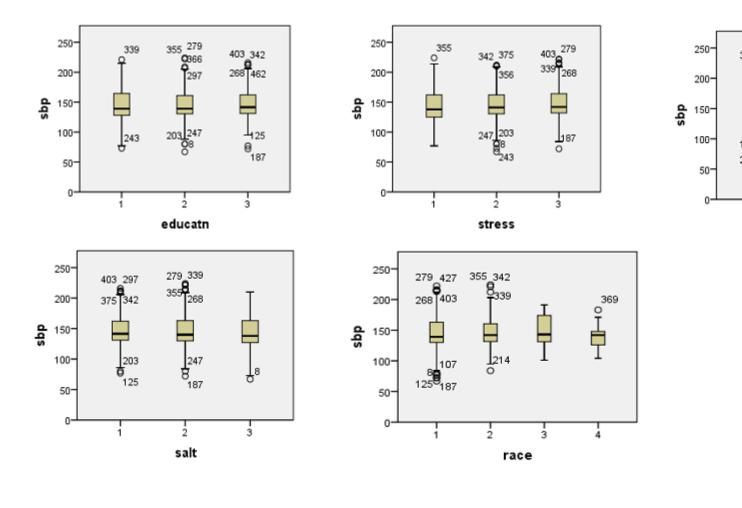
	F-value	Pr(> F )
Exercise	5.784	0.003
Over weight	19.017	0.000
Alcohol	5.281	0.005

#### Box plots for systolic blood pressure by education stress salt race and child bearing

279 342 403 268

J107

chidbear



# Statistical Analysis

Our aim of this study is to identify the best clinical and genetic markers predictors for systolic blood pressure

Clinical covariates

- Clinical and Genetic Markers covariates
  - All patients
  - High blood pressure Group /
     Low Blood pressure Group
  - Defining New Categorical response Variables with respect to high and low blood pressure
- R version 3.2.1/GLMNET/CAR/MASS

- 1. Multiple Regression
- 2. Variable selection based on:

Forward Method
Backward Method
Stopwise Method

Stepwise Method

Subset Selection-

Cp and R-Square Adjusted

#### Regularization Methods:

- 1. LASSO (with and without penalty factor)
- 2. Elastic Net (with and without penalty factor)

Regularization Methods with respect to binomial response (logistic regression):

- 1. LASSO (with and without penalty factor)
- 2. Elastic Net (with and without penalty factor)

Approaches

## All Clinical Covariates Into The Model

- Fitting Multiple Regression Model.
- Four effective covariates on systolic blood pressure are :
  - Smoke
  - Exercise
  - Alcohol
  - Treatment
- R<sup>2</sup> adjusted equal to 0.2 indicates that only about 20% of the variation in systolic blood pressure can be explained by the relationship to Clinical predictors.
- Complete table are attached to Appendix A

		Pr(> t )	Significance
	(Intercept)	0.7606	Jigimicanec
	gender	0.2838	
	married	0.174	
$\Delta$	smoke	3.47E-06	***
<b>%</b>	exercise	7.04E-05	***
	age	0.1598	
	weight	0.6183	
	height	0.2406	
	overweight	0.0744	•
_ [	race	0.4114	
	alcohol	1.63E-05	***
<b>A</b>	Treatment	1.90E-06	***
	BMI	0.1965	
	stress	0.0933	
	salt	0.8134	
	chldbear	0.2289	
	income	0.0824	
	educate	0.9935	

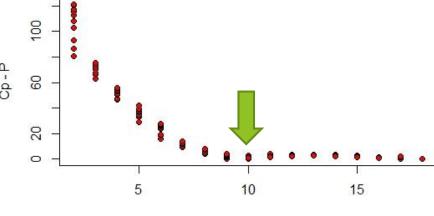
### Variable Selection

- 1. Variable selection based on:
  - ► Forward Method → All Covariates except BMI
    - ► Gender/Married /Smoke/ Age/ Weight/Height/Overweight/Race/ Alcohol/ Treatment / Stress/Salt/Childbearing/ Income/ Education.
  - ► Backward Method:
    - ▶ Married /Smoke/ Exercise/ Age/ Height/ Alcohol/ Treatment /BMI/ Stress/ Income
  - Stepwise Method

Similar

- ▶ Married /Smoke/ Exercise/ Age/ Height/ Alcohol/ Treatment /BMI/ Stress/ Income
- Subset Selection based on Cp:
- Subset Selection based on Adjusted R-Square :

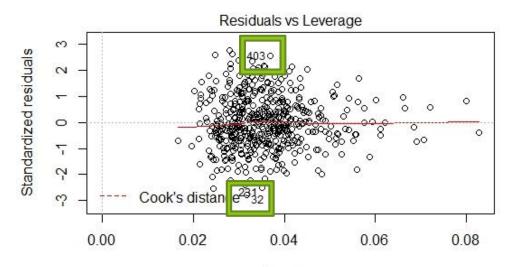
Gender	Married	Smoke	Exercise	Age	Weight	Height	Overweight	/
FALSE	TRUE	TRUE	TRUE	TRUE	FALSE	TRUE	FALSE	Stress
Salt	Childbearing	Income	Education	Race	Alcohol	Treatment	BMI	TRUE
FALSE	FALSE	TRUE	FALSE	FALSE	FALSE	TRUE	TRUE	



Selected Size

## Influence Data

- ► 13 Influence data based on
  - DFBETAS
  - DFFITS
  - Covariance ratios
  - Cook's distances
  - Diagonal elements of the hat matrix
- ► All those methods are available in "CAR" package in R.



Leverage lm(y ~ x)

Influence sample	dffit	cov.r	cook.d	hat
8	-0.41	0.83	0.01	0.2
32	-0.51	0.8	0.01	0.3
204	-0.12	1.13	0	0.3
231	-0.48	0.85	0.01	0.4
243	-0.43	0.86	0.01	0.3
339	0.38	0.89	0.01	0.3
355	0.47	0.8	0.01	0.3
356	0.44	0.87	0.01	0.3
366	0.41	0.83	0.01	0.2
375	0.45	0.83	0.01	0.3
403	0.5	0.84	0.01	0.4
474	0.42	0.86	0.01	0.3
485	-0.05	1.11	0	0.7

## Final model

- Selecting covariates based on previous variables selection methods.
  - Married /Smoke/ Exercise/ Age/ Height/ Alcohol/ Treatment /BMI/ Stress/ Income
  - Final Model After eliminating influence data
- ▶ R-squared adj equals to .22 which is increased by 0.02.
  - There might be some other effective factors
- The MSE is 543.93

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riables

- AIC of the model is 3089.52
- There are some Interactions effect, They were negligible because of small effect on the Model.

#### Interesting results:

Regardless of the R-squared, the significant coefficients still represent the mean change in the response for one unit of change in the predictor while holding other predictors in the model constant.

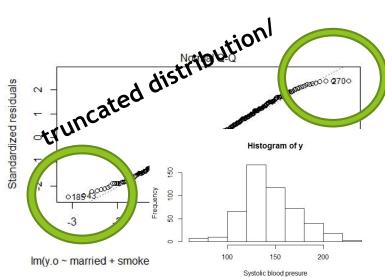
If age or height differed by one unit, and other covariates did not differ, systolic blood pressure will differ by 0.2 or 0.5 units, on average, respectively.

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	58.87889	17.7727	3.313	0.000994	***
Married/Y	3.84265	2.16786	1.773	0.076948	
Smoke/Y	10.40449	2.18799	4.755	2.64E-06	***
Exercise/2	-9.52305	2.70537	-3.52	0.000473	***
Exercise/3	-10.12955	2.53552	-3.995	7.50E-05	***
age	0.21189	0.08147	2.601	0.009592	**
height	0.55435	0.21693	2.556	0.010916	*
Alcohol/2	3.34206	2.66318	1.255	0.21013	
Alcohol/3	13.9706	2.6886	5.196	3.03E-07	***
Treatment/1	-13.99667	2.70463	-5.175	3.38E-07	***
BMI	1.148	0.16038	7.158	3.15E-12	***
Stress/2	2.9581	2.68051	1.104	0.270347	
Stress/3	4.30706	2.67563	1.61	0.108124	
Income/2	1.35101	2.59781	0.52	0.603267	
Income/3	5.77526	2.663	2.169	0.030604	*

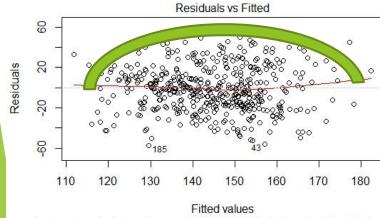
Categorical Variable

In Smoke group, we would expect that smoker have 10 units higher systolic blood pressure than Non smoker, on average, keeping all other covariates same.

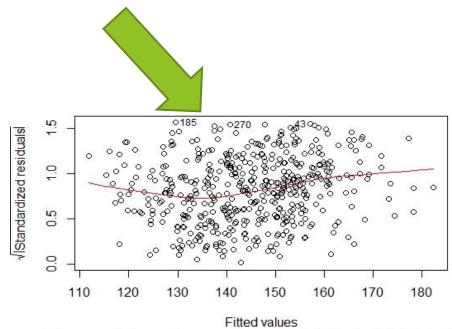
## Goodness of fit:



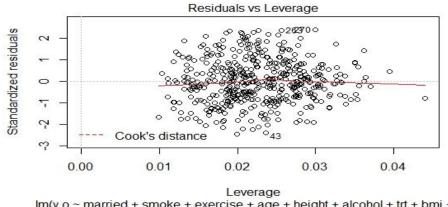
Using Transformation method have not been helpful



Im(y.o ~ married + smoke + exercise + age + height + alcohol + trt + bmi +



Im(y.o ~ married + smoke + exercise + age + height + alcohol + trt + bmi + ...



lm(y.o ~ married + smoke + exercise + age + height + alcohol + trt + bmi +

# Variable Selection based on Regularization Method:

The LASSO (Least Absolute Shrinkage and Selection Operator):

LASSO is a regression method that involves penalizing the absolute size of the regression coefficients. This is convenient when we want some automatic variable selection specially with large number of variable, or when dealing with highly correlated predictors.

Limitation In LASSO Method:

the LASSO selects at most n variables. Also if there is a group of highly correlated variables, then the LASSO tends to select one variable from a group and ignore the others.

The Elastic Net:

This method overcomes the limitations of the LASSO.

## Clinical and Genetic Markers covariates

LASSO(without Penalty Factor): Elastic Net (without Penalty Factor):

MSE=381.43 MSE=385.96

R-Square adj = 0.56 R-Square adj = 0.56

Selected Covariates are the same for both Methods:

8 out of 17 - Married/ Smoke /Exercise /Weight/ Overweight/ Alcohol/ Treatment/BMI Stress

58 out of 483

- g7 g9 g10 g46 g48 **g50** g59 g63 g86 g92
- ▶ g108 g120 g122 g135 g137 g150 g160 g168 g169 g175 g179 g182 g187 g191
- **g200** g204 g222 g231 g232 g271 g279 g288 g289 g292 g295 **g298**
- ▶ g309 g330 g337 g348 g356 g362 g364 g366 g371 g377 g391
- ▶ g411 g412 g422 g425 g438 g443 g447 **g453** g465 g469 g480

Genetic Markers	Related Coefficient
G50	13.4
G200	10.65
G298	2.5
G453	2.27

# High/low Blood pressure groups

- > Based on provided information on website we have divided all patient into two groups.
- Elastic Net methods were applied (n<p)</p>

#### **Significant Covariates**



High blood pressure group

Smoke / overweight /alcohol g50 g200 g377 g432 g458 g459

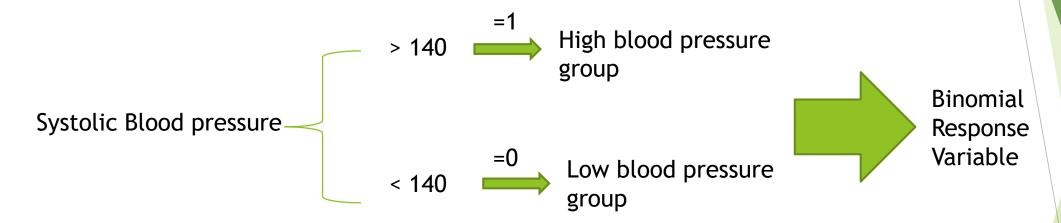
MSE = 217.36

Low blood pressure group

Age /weight / treatment /BMI g5 g7 g9 g47 g95 g111 g112 g132 g169 g177 g191 g200 g205 g207 g214 g216 g242 g253 g285 g327 g357 g359 g377 g385 g393 g427 g434 g464 g474 g478 g480

MSE= 189.27

# Define New Categorical response Variables



# Regularization Methods with respect to binomial distribution of y

- Regularization Methods were employed with respect to binomial distribution of y, Selected variables are:
  - Married/ exercise/ overweight/ Treatment/ BMI/ stress.
  - ▶ g10 g36 g49 g50 g65 g75 g86 g98
  - g120 g122 g137 g150 g168 g187 g191
  - **g200** g204 g231 g279 g298
  - ▶ g309 g330g385 g391
  - g412 g425 g447 g450 g453 g460 g469 g475
- ▶ In this model, The MSE equals to 0.63
- Based on this model, we can calculate the probability of having High or Low blood pressure given selected covariates.

## Conclusion and Recommendations

- According to the results of this study the main predictive factors for systolic Blood pressure were introduced based on regularization method.
- We have found that two genetics markers: "g50" and "g200" were very important predictors for predicting systolic blood pressure( because they have appeared in most models)
- We recommend further study to investigate the association between systolic blood pressure and clinical and genetic markers variables based on principal component analysis and clustering methods to bring out strong patterns in predictors.

### References

- ▶ [1] <u>David W. Hosmer, Jr., Stanley Lemeshow; Rodney X. Sturdivant</u>, "Applied Logistic Regression, 3rd Edition", John Wiley & Sons, New Jersey, 2013.
- ▶ [2] Zou, Hui, Hastie, Trevor, <u>"Regularization and Variable Selection via the Elastic Net"</u>. <u>Journal of the Royal Statistical Society</u>, Series B: 301-320, 2005.
- ▶ [3] Friedman, Jerome; Trevor Hastie; Rob Tibshirani. "Regularization Paths for Generalized Linear Models via Coordinate Descent". Journal of Statistical Software: 1-22, 2010.
- ▶ [4] <a href="http://cran.r-project.org/web/packages/glmnet/index.html">http://cran.r-project.org/web/packages/glmnet/index.html</a>
- ▶ **[5]** http://web.stanford.edu/~hastie/glmnet/glmnet\_alpha.html
- ► [6]Rodríguez, G. Lecture Notes on Generalized Linear Models. URL: <a href="http://data.princeton.edu/wws509/notes/">http://data.princeton.edu/wws509/notes/</a>, 2007.
- ▶ **[7] Hastie**, Trevor, **Tibshirani**, Robert, **Friedman**, Jerome, "The Elements of Statistical Learning, Data Mining, Inference, and Prediction", Second Edition, 2009.

# Appendix A

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	36.47519			0.4527	
genderM	-2.85038	3.04392	-0.936	0.3495	
marriedY	2.77287	2.32273	1.194	0.2332	
smokeY	10.74273	2.34442	4.582	5.89E-06	***
as.factor(x\$exercise)2	-11.47064	2.9038	-3.95	9.00E-05	***
as.factor(x\$exercise)3	-10.72057	2.7085	-3.958	8.71E-05	***
age	0.1095	0.08686	1.261	0.2081	
weight	-0.13328	0.14571	-0.915	0.3608	
height	1.17103	0.7474	1.567	0.1178	
as.factor(x\$overwt)2	8.98298	4.336	2.072	0.0388	*
as.factor(x\$overwt)3	11.92226	5.92162	2.013	0.0446	*
as.factor(x\$race)2	0.3626	2.92517	0.124	0.9014	
as.factor(x\$race)3	0.55622	5.34617	0.104	0.9172	
as.factor(x\$race)4	-6.72193	5.83917	-1.151	0.2502	
as.factor(x\$alcohol)2	1.43863	2.85796	0.503	0.6149	
as.factor(x\$alcohol)3	13.1264	2.89709	4.531	7.45E-06	***
as.factor(x\$trt)1	-15.26782	2.92514	-5.22	2.69E-07	***
bmi	1.39409	0.83965	1.66	0.0975	•
as.factor(x\$stress)2	2.23425	2.88105	0.775	0.4384	
as.factor(x\$stress)3	5.04992	2.86234	1.764	0.0783	
as.factor(x\$salt)2	1.52332	2.88632	0.528	0.5979	
as.factor(x\$salt)3	0.98388	2.81447	0.35	0.7268	
as.factor(x\$chldbear)2	-4.5792	3.20196	-1.43	0.1533	
as.factor(x\$chldbear)3	NA	NA	NA	NA	
as.factor(x\$income)2	0.97177	2.77538	0.35	0.7264	
as.factor(x\$income)3	4.3264	2.84459	1.521	0.1289	
as.factor(x\$educatn)2	0.45128	2.82983	0.159	0.8734	
as.factor(x\$educatn)3	-0.57115	2.81851	-0.203	0.8395	

## Appendix B:

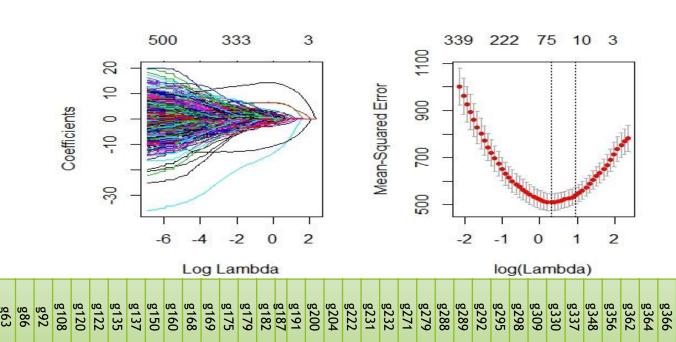
Min MSE obtained by Lambda = 2.14

married

married

exercise

LASSO



g191 g187 g182

g222 g204 g200

g179

g175

1.0169

g160

g108

g48 g46 g10 0.9390

g295 g292

g330 g309 g298 g348

g356

g366 g364 g362

g337

g371

g391 g377 g422

1.8368

0.1210

g279

g289 g288

g271

g232

g231

# Appendix C:

Min MSE obtained by Lambda 0.03516715

