

Simulation

```
# Example 2 of the thesis
generate = function(n, # number of observations
                    p, # number of predictors
                    ps = 3, # number of strong signals
                    pwbc = 30, # number of WBC signals
                    pwai = 30, # number of WAI signals
                    corr = 0.7, # correlation coefficient (see example 2)
                    c = 20 # the c in the definition of strong and weak
                    ) {
  if (ps + pwbc + pwai >= n) {
    stop("number of true predictors should be less than number of observations")
  }
  if (abs(corr) > 1) {
    stop("correlation coefficient must be between -1 and 1")
  }
  # We hope that beta_strong = 20 and beta_weak = 0.5
  if ((c*sqrt(log(p)/n)) <= 0.5 & (c*sqrt(log(p)/n)) >= 20) {
    stop("please select other n and p")
  }
  corr_matrix = matrix(rep(0, len = p^2), nrow = p)
  corr_num = pwbc %/% ps
  # Correlations between strong & wbc
  for (i in 1:(ps - 1)) {
    for (j in (ps + 1 + (i - 1)*corr_num):(ps + i*corr_num)) {
      corr_matrix[i, j] = corr
      corr_matrix[j, i] = corr
      for (k in j:(ps + i*corr_num)) {
        corr_matrix[j, k] = corr
        corr_matrix[k, j] = corr
      }
    }
  }
  for (j in (ps + 1 + (ps - 1)*corr_num):(ps + pwbc)) {
    corr_matrix[i, j] = corr
    corr_matrix[j, i] = corr
    for (k in j:(ps + pwbc)) {
      corr_matrix[j, k] = corr
      corr_matrix[k, j] = corr
    }
  }
  # Correlations within wai
  for (j in (ps + pwbc + 1):(ps + pwbc + pwai)) {
    for (k in j:(ps + pwbc + pwai)) {
      corr_matrix[j, k] = corr
      corr_matrix[k, j] = corr
    }
  }
}
```

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}
diag(corr_matrix) = 1
X = mvrnorm(n, mu = rep(0, p), Sigma = corr_matrix, tol = 1)
beta = c(rep(20, ps), rep(0.5, pwbc + pwai), rep(0, p - ps - pwbc - pwai))
Y = X %*% beta + rnorm(n)
df = as.data.frame(cbind(Y, X))
colnames(df)[1] <- "y"
for (i in 2:(1 + ps)) {
  colnames(df)[i] <- paste("strong", i - 1, sep = "_")
}
for (i in (2 + ps):(1 + ps + pwbc)) {
  colnames(df)[i] <- paste("wbc", i - 1 - ps, sep = "_")
}
for (i in (2 + ps + pwbc):(1 + ps + pwbc + pwai)) {
  colnames(df)[i] <- paste("wai", i - 1 - ps - pwbc, sep = "_")
}
for (i in (2 + ps + pwbc + pwai):(1 + p)) {
  colnames(df)[i] <- paste("null", i - 1 - ps - pwbc - pwai, sep = "_")
}
return(df)
}

```

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# Number of observations
n = 1000
# Number of predictors
p = 200

# Data manipulation
df = generate(n, p)
X = as.matrix(df[1:p + 1])
Y = as.matrix(df[1])

# Forward Selection
fit_forward = step(object = lm(y ~ 1, data = df),
  scope = formula(lm(y ~ ., data = df)),
  direction = "forward",
  k = 2,
  trace = 0)
summary(fit_forward)

```

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##
## Call:
## lm(formula = y ~ strong_2 + strong_1 + strong_3 + wai_1 + wai_12 +
##   wai_20 + wai_17 + wai_18 + wbc_8 + wai_27 + wai_22 + wai_23 +
##   wbc_5 + wai_9 + wai_26 + wai_13 + wbc_3 + wai_4 + wai_25 +
##   wbc_2 + wai_11 + wai_24 + wai_10 + wbc_9 + wai_15 + wbc_1 +
##   wai_3 + wai_30 + wbc_7 + wai_2 + wai_8 + wai_29 + wbc_10 +
##   wai_19 + wai_21 + wai_6 + wai_7 + wbc_6 + wai_14 + wai_5 +
##   wai_28 + wai_16 + wbc_4 + null_4 + null_133 + null_32 + null_19 +
##   null_20 + null_17 + null_41 + null_78 + null_77 + wbc_13 +
##   null_87 + null_115 + null_68 + null_66 + null_111 + null_29 +
##   null_30 + null_99 + null_120 + null_91 + null_52 + null_98 +
##   null_90 + null_12 + wbc_22 + null_46 + null_9, data = df)

```

```

##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.07408 -0.64076  0.01612  0.66739  3.10227
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.007722   0.032167  -0.240  0.810342
## strong_2     25.295489   0.053955 468.829 < 2e-16 ***
## strong_1     20.009386   0.057698 346.794 < 2e-16 ***
## strong_3     20.001008   0.032731 611.077 < 2e-16 ***
## wai_1         0.401739   0.059137   6.793 1.95e-11 ***
## wai_12        0.635099   0.057610  11.024 < 2e-16 ***
## wai_20        0.542396   0.054782   9.901 < 2e-16 ***
## wai_17        0.634601   0.056204  11.291 < 2e-16 ***
## wai_18        0.479990   0.057422   8.359 2.29e-16 ***
## wbc_8         0.676563   0.054999  12.301 < 2e-16 ***
## wai_27        0.613699   0.059088  10.386 < 2e-16 ***
## wai_22        0.492599   0.056626   8.699 < 2e-16 ***
## wai_23        0.480641   0.056631   8.487 < 2e-16 ***
## wbc_5         0.411673   0.061168   6.730 2.96e-11 ***
## wai_9         0.490152   0.056567   8.665 < 2e-16 ***
## wai_26        0.629849   0.055654  11.317 < 2e-16 ***
## wai_13        0.516524   0.057551   8.975 < 2e-16 ***
## wbc_3         0.525768   0.056827   9.252 < 2e-16 ***
## wai_4         0.437596   0.057018   7.675 4.19e-14 ***
## wai_25        0.546339   0.055745   9.801 < 2e-16 ***
## wbc_2         0.590524   0.057704  10.234 < 2e-16 ***
## wai_11        0.577421   0.055997  10.312 < 2e-16 ***
## wai_24        0.465604   0.057651   8.076 2.06e-15 ***
## wai_10        0.521127   0.057877   9.004 < 2e-16 ***
## wbc_9         0.528853   0.056047   9.436 < 2e-16 ***
## wai_15        0.523171   0.056172   9.314 < 2e-16 ***
## wbc_1         0.431697   0.054348   7.943 5.67e-15 ***
## wai_3         0.541725   0.056938   9.514 < 2e-16 ***
## wai_30        0.517091   0.057789   8.948 < 2e-16 ***
## wbc_7         0.475720   0.055825   8.522 < 2e-16 ***
## wai_2         0.506553   0.058338   8.683 < 2e-16 ***
## wai_8         0.407771   0.058822   6.932 7.74e-12 ***
## wai_29        0.470951   0.057996   8.120 1.47e-15 ***
## wbc_10        0.474960   0.056241   8.445 < 2e-16 ***
## wai_19        0.475889   0.055307   8.604 < 2e-16 ***
## wai_21        0.463898   0.058200   7.971 4.61e-15 ***
## wai_6         0.466351   0.055845   8.351 2.45e-16 ***
## wai_7         0.470673   0.056429   8.341 2.64e-16 ***
## wbc_6         0.508014   0.056659   8.966 < 2e-16 ***
## wai_14        0.496141   0.059367   8.357 2.33e-16 ***
## wai_5         0.410162   0.057236   7.166 1.57e-12 ***
## wai_28        0.433474   0.059279   7.312 5.66e-13 ***
## wai_16        0.396260   0.058253   6.802 1.84e-11 ***
## wbc_4         0.361506   0.055073   6.564 8.69e-11 ***
## null_4        0.113890   0.031928   3.567 0.000379 ***
## null_133      0.106999   0.032228   3.320 0.000935 ***
## null_32       0.108137   0.032420   3.335 0.000885 ***

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## null_19      -0.104419    0.032099   -3.253 0.001183 **
## null_20       0.081789    0.031861    2.567 0.010411 *
## null_17      -0.088082    0.031955   -2.756 0.005958 **
## null_41       0.079033    0.032349    2.443 0.014746 *
## null_78      -0.086122    0.031661   -2.720 0.006648 **
## null_77       0.076606    0.031872    2.404 0.016434 *
## wbc_13        0.138575    0.047775    2.901 0.003813 **
## null_87       0.067510    0.032833    2.056 0.040045 *
## null_115      0.073677    0.031980    2.304 0.021452 *
## null_68      -0.064951    0.032653   -1.989 0.046982 *
## null_66       0.062138    0.029636    2.097 0.036292 *
## null_111     -0.050826    0.031548   -1.611 0.107509
## null_29      -0.058981    0.031320   -1.883 0.059990 .
## null_30      -0.052218    0.031538   -1.656 0.098117 .
## null_99       0.059342    0.031502    1.884 0.059909 .
## null_120     -0.057090    0.031631   -1.805 0.071421 .
## null_91      -0.055598    0.033354   -1.667 0.095865 .
## null_52      -0.050285    0.033033   -1.522 0.128281
## null_98      -0.048739    0.031200   -1.562 0.118594
## null_90       0.047010    0.030621    1.535 0.125070
## null_12      -0.048739    0.031757   -1.535 0.125186
## wbc_22        0.078191    0.048296    1.619 0.105788
## null_46      -0.048111    0.031580   -1.523 0.127982
## null_9       -0.047278    0.031808   -1.486 0.137515
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9758 on 929 degrees of freedom
## Multiple R-squared:  0.9996, Adjusted R-squared:  0.9995
## F-statistic: 3.145e+04 on 70 and 929 DF,  p-value: < 2.2e-16
```

```
params_forward = fit_forward$coefficients[-1]
params_forward
```

```
##      strong_2    strong_1    strong_3      wai_1      wai_12      wai_20
## 25.29548861 20.00938587 20.00100828 0.40173914 0.63509941 0.54239595
##      wai_17      wai_18      wbc_8      wai_27      wai_22      wai_23
## 0.63460104 0.47999028 0.67656317 0.61369859 0.49259906 0.48064118
##      wbc_5      wai_9      wai_26      wai_13      wbc_3      wai_4
## 0.41167291 0.49015160 0.62984926 0.51652406 0.52576780 0.43759612
##      wai_25      wbc_2      wai_11      wai_24      wai_10      wbc_9
## 0.54633943 0.59052412 0.57742149 0.46560369 0.52112718 0.52885333
##      wai_15      wbc_1      wai_3      wai_30      wbc_7      wai_2
## 0.52317130 0.43169706 0.54172468 0.51709146 0.47572033 0.50655298
##      wai_8      wai_29      wbc_10      wai_19      wai_21      wai_6
## 0.40777084 0.47095149 0.47495990 0.47588936 0.46389793 0.46635070
##      wai_7      wbc_6      wai_14      wai_5      wai_28      wai_16
## 0.47067295 0.50801433 0.49614075 0.41016158 0.43347407 0.39626006
##      wbc_4      null_4      null_133      null_32      null_19      null_20
## 0.36150568 0.11389003 0.10699859 0.10813701 -0.10441898 0.08178900
##      null_17      null_41      null_78      null_77      wbc_13      null_87
## -0.08808231 0.07903326 -0.08612236 0.07660582 0.13857490 0.06751004
##      null_115      null_68      null_66      null_111      null_29      null_30
## 0.07367699 -0.06495149 0.06213767 -0.05082599 -0.05898074 -0.05221774
```

```
##      null_99      null_120      null_91      null_52      null_98      null_90
## 0.05934190 -0.05709011 -0.05559848 -0.05028548 -0.04873925 0.04701043
##      null_12      wbc_22      null_46      null_9
## -0.04873856 0.07819059 -0.04811093 -0.04727846
```

```
# LASSO
fit_lasso = cv.glmnet(X, Y,
                      nfolds = 10,
                      type.measure = "mse")
param_lasso = fit_lasso$glmnet.fit$beta[, fit_lasso$lambda == fit_lasso$lambda.1se]
param_lasso
```

```
## strong_1 strong_2 strong_3 wbc_1 wbc_2 wbc_3 wbc_4
## 20.0052585 25.0354613 19.5689981 0.3141667 0.5464197 0.4739149 0.2594344
## wbc_5 wbc_6 wbc_7 wbc_8 wbc_9 wbc_10 wbc_11
## 0.4793141 0.3518509 0.3719128 0.6216419 0.4699600 0.4502717 0.0000000
## wbc_12 wbc_13 wbc_14 wbc_15 wbc_16 wbc_17 wbc_18
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## wbc_19 wbc_20 wbc_21 wbc_22 wbc_23 wbc_24 wbc_25
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## wbc_26 wbc_27 wbc_28 wbc_29 wbc_30 wai_1 wai_2
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.2884951 0.4941180
## wai_3 wai_4 wai_5 wai_6 wai_7 wai_8 wai_9
## 0.5547449 0.5403319 0.4495702 0.4451067 0.4257245 0.4356972 0.3656373
## wai_10 wai_11 wai_12 wai_13 wai_14 wai_15 wai_16
## 0.4888274 0.4913452 0.6501002 0.5062421 0.4901149 0.4835426 0.3058103
## wai_17 wai_18 wai_19 wai_20 wai_21 wai_22 wai_23
## 0.6136230 0.4719756 0.5104335 0.5152844 0.5240354 0.5285334 0.4928308
## wai_24 wai_25 wai_26 wai_27 wai_28 wai_29 wai_30
## 0.4182108 0.4649379 0.6231632 0.5401755 0.3842580 0.4920385 0.4629532
## null_1 null_2 null_3 null_4 null_5 null_6 null_7
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## null_8 null_9 null_10 null_11 null_12 null_13 null_14
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## null_15 null_16 null_17 null_18 null_19 null_20 null_21
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## null_22 null_23 null_24 null_25 null_26 null_27 null_28
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## null_29 null_30 null_31 null_32 null_33 null_34 null_35
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## null_36 null_37 null_38 null_39 null_40 null_41 null_42
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## null_43 null_44 null_45 null_46 null_47 null_48 null_49
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## null_50 null_51 null_52 null_53 null_54 null_55 null_56
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## null_57 null_58 null_59 null_60 null_61 null_62 null_63
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## null_64 null_65 null_66 null_67 null_68 null_69 null_70
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## null_71 null_72 null_73 null_74 null_75 null_76 null_77
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
## null_78 null_79 null_80 null_81 null_82 null_83 null_84
## 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000 0.0000000
```

##	null_85	null_86	null_87	null_88	null_89	null_90	null_91
##	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000
##	null_92	null_93	null_94	null_95	null_96	null_97	null_98
##	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000
##	null_99	null_100	null_101	null_102	null_103	null_104	null_105
##	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000
##	null_106	null_107	null_108	null_109	null_110	null_111	null_112
##	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000
##	null_113	null_114	null_115	null_116	null_117	null_118	null_119
##	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000
##	null_120	null_121	null_122	null_123	null_124	null_125	null_126
##	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000
##	null_127	null_128	null_129	null_130	null_131	null_132	null_133
##	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000
##	null_134	null_135	null_136	null_137			
##	0.0000000	0.0000000	0.0000000	0.0000000			