

Simulation

```
# example 2 of the thesis
generate = function(n = 200, # number of observations
                    p = 200, # 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)
list_names = c("y",
               paste("s", 1:ps, sep = "_"),
               paste("wbc", 1:pwbc, sep = "_"),
               paste("wai", 1:pwai, sep = "_"),
               paste("null", 1:(p - ps - pwbc - pwai), sep = "_"))
data = as.data.frame(cbind(Y, X))
colnames(data)[1] <- "y"
for (i in 2:(1 + ps)) {
  colnames(data)[i] <- paste("s", i - 1, sep = "_")
}
for (i in (2 + ps):(1 + ps + pwbc)) {
  colnames(data)[i] <- paste("wbc", i - 1 - ps, sep = "_")
}
for (i in (2 + ps + pwbc):(1 + ps + pwbc + pwai)) {
  colnames(data)[i] <- paste("wai", i - 1 - ps - pwbc, sep = "_")
}
for (i in (2 + ps + pwbc + pwai):p) {
  colnames(data)[i] <- paste("null", i - 1 - ps - pwbc - pwai, sep = "_")
}
return(data)
}

```

```

# retrieve data
data = generate()
df = data.frame(data)
n = 200

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

```

```

##
## Call:
## lm(formula = y ~ 1, data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -122.010  -31.912    2.924   30.477  110.671
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -1.406       2.972  -0.473   0.637
##
## Residual standard error: 42.03 on 199 degrees of freedom

```

```
# LASSO
X = as.matrix(data[2:n + 1])
Y = as.matrix(data[1])
fit_lasso = cv.glmnet(X, Y, nfolds = 10, type.measure = "mse")
param.best = fit_lasso$glmnet.fit$beta[, fit_lasso$lambda == fit_lasso$lambda.1se]
param.best
```

```
##      s_2      s_3      wbc_1      wbc_2      wbc_3      wbc_4
## 22.27692015 18.16185664 4.17326982 0.00000000 1.04021605 1.67275454
##      wbc_5      wbc_6      wbc_7      wbc_8      wbc_9      wbc_10
## 0.64560307 5.40336974 2.04787239 1.38428875 1.57131818 2.00732512
##      wbc_11      wbc_12      wbc_13      wbc_14      wbc_15      wbc_16
## 0.00000000 0.00000000 0.00000000 0.04294236 0.00000000 0.00000000
##      wbc_17      wbc_18      wbc_19      wbc_20      wbc_21      wbc_22
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      wbc_23      wbc_24      wbc_25      wbc_26      wbc_27      wbc_28
## 0.62002143 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      wbc_29      wbc_30      wai_1      wai_2      wai_3      wai_4
## 0.00000000 0.00000000 0.13116469 0.00000000 0.00000000 2.47426048
##      wai_5      wai_6      wai_7      wai_8      wai_9      wai_10
## 1.22959767 0.00000000 0.11098182 0.00000000 0.00000000 0.00000000
##      wai_11      wai_12      wai_13      wai_14      wai_15      wai_16
## 0.52406902 0.00000000 0.00000000 0.00000000 0.95305119 0.00000000
##      wai_17      wai_18      wai_19      wai_20      wai_21      wai_22
## 0.00000000 0.00000000 0.00000000 0.56816883 1.05370721 0.00000000
##      wai_23      wai_24      wai_25      wai_26      wai_27      wai_28
## 0.00000000 0.00000000 0.00000000 3.01450452 0.00000000 0.00000000
##      wai_29      wai_30      null_1      null_2      null_3      null_4
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      null_5      null_6      null_7      null_8      null_9      null_10
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      null_11      null_12      null_13      null_14      null_15      null_16
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      null_17      null_18      null_19      null_20      null_21      null_22
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      null_23      null_24      null_25      null_26      null_27      null_28
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      null_29      null_30      null_31      null_32      null_33      null_34
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      null_35      null_36      null_37      null_38      null_39      null_40
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      null_41      null_42      null_43      null_44      null_45      null_46
## 0.00000000 0.00000000 0.00000000 -0.40449752 0.00000000 0.00000000
##      null_47      null_48      null_49      null_50      null_51      null_52
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      null_53      null_54      null_55      null_56      null_57      null_58
## 0.00000000 0.03876890 0.00000000 0.00000000 0.00000000 0.00000000
##      null_59      null_60      null_61      null_62      null_63      null_64
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      null_65      null_66      null_67      null_68      null_69      null_70
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      null_71      null_72      null_73      null_74      null_75      null_76
## 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
##      null_77      null_78      null_79      null_80      null_81      null_82
```

##	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
##	null_83	null_84	null_85	null_86	null_87	null_88
##	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
##	null_89	null_90	null_91	null_92	null_93	null_94
##	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
##	null_95	null_96	null_97	null_98	null_99	null_100
##	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
##	null_101	null_102	null_103	null_104	null_105	null_106
##	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
##	null_107	null_108	null_109	null_110	null_111	null_112
##	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
##	null_113	null_114	null_115	null_116	null_117	null_118
##	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
##	null_119	null_120	null_121	null_122	null_123	null_124
##	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
##	null_125	null_126	null_127	null_128	null_129	null_130
##	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
##	null_131	null_132	null_133	null_134	null_135	null_136
##	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
##	V201					
##	0.00000000					