

HD_regression

Haotian Xu

9/2/2021

This is a simple guide for changepoint detection in high-dimensional linear regression.

Function *simu.change.regression* simulates sparse regression model with changepoints in coefficients.

Dynamic programming is implemented for regression changepoint detection:

- *DP.regression*: performs dynamic programming for regression changepoint detection through l0 penalty.
 - *CV.search.DP.regression*: perform grid search to select the tuning parameters (gamma for l0, lambda for l1) through Cross-Validation.

In addition, function *local.refine.regression* performs local refinement for an initial changepoint estimation.

Simulate data

```
library(changepoints)

## Loading required package: gglasso
## Loading required package: glmnet
## Loading required package: Matrix
## Loaded glmnet 4.1-2
## Loading required package: penalized
## Loading required package: survival
##
## Attaching package: 'survival'
## The following object is masked from 'package:gglasso':
##
##      colon
## Welcome to penalized. For extended examples, see vignette("penalized").
## Loading required package: ks
## Loading required package: MASS
## parameters for simulating data
d0 = 5 # the number of nonzero elements
sigma = 1 # error standard deviation
kappa = 5 # minimum jump size in l2 norm
delta = 5 # minimal gap between boundaries
n = 200 # sample size
p = 50 # dimensionality
change.point = c(80, 170)
```

```

set.seed(0)
data = simu.change.regression(d0, change.point, p, n, sigma, kappa)
X = data$X
y = data$y
cpt_true = data$cpt.true

```

Perform dynamic programming

```

gamma.dp.set = c(0.01, 0.5, 1, 5, 10, 50) # a set of tuning parameters for DP
lambda.dp.set = c(0.01, 0.1, 1, 1.5) # a set of tuning parameters for lasso
DP_result = CV.search.DP.regression(y, X, gamma.dp.set, lambda.dp.set, delta) # grid search through cross-validation

```

```

##      [,1]      [,2]      [,3]      [,4]
## [1,] numeric,3 numeric,11 numeric,9 numeric,8
## [2,] 3         11         9         8
## [3,] 881.6197  780.3093  750.3877  791.0653
## [4,] 0.06585709 1.584812  110.3635  194.1428
## [5,] numeric,2 numeric,3  numeric,6 numeric,8
## [6,] 2         3         6         8
## [7,] 764.6688  554.4619  608.9422  791.0653
## [8,] 0.1440713 4.898734  111.2579  194.1428
## [9,] numeric,2 numeric,2  numeric,5 numeric,5
## [10,] 2        2         5         5
## [11,] 764.6688 950.2495  552.9739  483.8277
## [12,] 0.1440713 8.21851  114.369  204.563
## [13,] 97        numeric,2 numeric,2 numeric,2
## [14,] 1         2         2         2
## [15,] 7033.705  950.2495  307.2387  403.3635
## [16,] 6.969676  8.21851  134.5234  226.9008
## [17,] 97        83        numeric,2 numeric,2
## [18,] 1         1         2         2
## [19,] 7033.705  787.1442  307.2387  403.3635
## [20,] 6.969676  36.18725  134.5234  226.9008
## [21,] 97        numeric,0  numeric,0 numeric,0
## [22,] 1         0         0         0
## [23,] 7033.705  1019.984  689.6947  697.7489
## [24,] 6.969676  249.1271  409.5372  469.2157

```

```

min_idx = as.vector(arrayInd(which.min(DP_result$test_error), dim(DP_result$test_error))) # select gamma
cpt_DP_hat = unlist(DP_result$cpt_hat[min_idx[1], min_idx[2]]) # estimated changepoints by DP
cpt_DP_hat

```

```
## [1] 79 167
```

```
Hausdorff.dist(cpt_DP_hat, cpt_true)
```

```
## [1] 3
```

```

zeta = 1 # tuning parameter for group lasso
cpt_DPlr_hat = local.refine.regression(cpt_DP_hat, y, X, zeta, w = 1/3) # perform local refinement
cpt_DPlr_hat

```

```
## [1] 79 170
```

```
Hausdorff.dist(cpt_DPlr_hat, cpt_true)
```

```
## [1] 1
```