Parallel all papers PLS for corn data

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1 A Partial Least Squares-Based Consensus Regression Method for the Analysis of Near-Infrared Complex Spectral Data of Plant Samples

```
60-20 10

1~mp6 0.159

## 0.1323149 0.1537739 0.1752328

## user system elapsed

## 0.004 0.000 8.208
```

1 A Partial Least Squares-Based Consensus Regression Method for the Analysis of Near-Infrared Complex Spectral Data of Plant Samples

```
60-20 10
2~mp6 0.107

## 0.08103781 0.09794708 0.1148564

## user system elapsed
## 0.002 0.001 7.697
```

1 A Partial Least Squares-Based Consensus Regression Method for the Analysis of Near-Infrared Complex Spectral Data of Plant Samples

```
60-20 10

3~mp6 0.150

## 0.125667 0.1434097 0.1611523

## user system elapsed

## 0.003 0.000 8.698
```

1 A Partial Least Squares-Based Consensus Regression Method for the Analysis of Near-Infrared Complex Spectral Data of Plant Samples

```
60-20 10

4~mp6 0.370

## 0.2868808 0.3511157 0.4153506

## user system elapsed

## 0.002 0.000 8.302
```

2 A strategy that iteratively retains informative variables for selecting optimal variable subset in multivariate calibration

```
64-16 10

1~m5 RMSEC = 0.0149; RMSEP = 0.0201

## 0.01972323 0.02083414 0.02194506

## 0.01593558 0.01855212 0.02116865

## user system elapsed
```

3 Cross-validation for the selection of spectral variables using the successive projections algorithm

```
60-20 SavitzkyGolay filler (in)

1~m5 0.040

## 0.03537567 0.04127559 0.04717551

## user system elapsed
## 0.003 0.001 6.434
```

0.003 0.000 9.759

##

3 Cross-validation for the selection of spectral variables using the successive projections algorithm

```
60-20 SavitzkyGolay filler (in)
2~m5 0.029
## 0.03536274 0.04220874 0.04905474
```

```
## user system elapsed
## 0.003 0.000 9.512
```

3 Cross-validation for the selection of spectral variables using the successive projections algorithm

```
60-20 SavitzkyGolay filler (in)
3~m5 0.119

## 0.08316763 0.1050105 0.1268535

## user system elapsed
## 0.002 0.000 5.791
```

3 Cross-validation for the selection of spectral variables using the successive projections algorithm

```
60-20 SavitzkyGolay filler (in)

4~m5 0.196

## 0.2029086 0.2454358 0.2879631

## user system elapsed

## 0.003 0.000 5.738
```

4 Reduced PCR/PLSR models by subspace projections

```
40-40 Scale 4

1~m5 0.36

## 0.2607245 0.3516627 0.4426009

## user system elapsed
## 0.002 0.000 7.716
```

4 Reduced PCR/PLSR models by subspace projections

```
2~m5 0.97

## 0.6307844 0.6980552 0.765326

## user system elapsed

## 0.003 0.000 7.734
```

40-40 Scale 4

4 Reduced PCR/PLSR models by subspace projections

40-40 Scale 4

```
3~m5 0.44

## 0.5263017 0.6390286 0.7517555

## user system elapsed
## 0.002 0.001 7.718
```

4 Reduced PCR/PLSR models by subspace projections

```
40-40 Scale 4

4~m5 0.49

## 0.6705351 0.7936221 0.9167091

## user system elapsed
```

user system elapsed 0.002 0.001 7.718

5 Stability competitive adaptive reweighted sampling (SCARS) and its applications to multivariate calibration of NIR spectra

```
40-40 Scale 10

1~mp5 0.357

## 0.3675207 0.4148987 0.4622767

## user system elapsed

## 0.002 0.000 9.529
```

6 Pretreating near infrared spectra with fractional order Savitzky–Golay differentiation (FOSGD)

```
64-16 None 10

2~m5 RMSECV=0.0363; RMSECP=0.04

## 0.05392444 0.06338411 0.07284378

## user system elapsed

## 0.002 0.000 8.489
```

6 Pretreating near infrared spectra with fractional order Savitzky–Golay differentiation (FOSGD)

```
64-16 None 10

4~m5 RMSECV=0.24; RMSECP=0.219

## 0.2147657 0.2653311 0.3158965

## user system elapsed
## 0.002 0.000 8.636
```

7 A variable elimination method to improve the parsimony of MLR models using the successive projections algorithm

```
60-20 None (in)

1~m5 0.045(06)

## 0.03352668 0.04006216 0.04659764

## user system elapsed

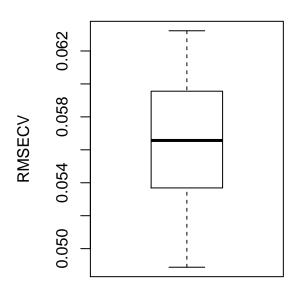
## 0.003 0.000 5.661
```

7 A variable elimination method to improve the parsimony of MLR models using the successive projections algorithm

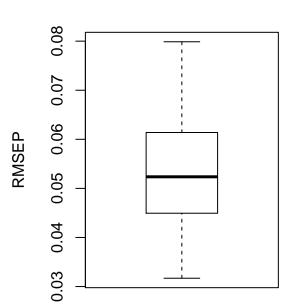
60-20 None (in)

 $2 \sim m5 \ 0.028(10)$

PLS after scaled



PLS after scaled



Number of Calibration

Number of Calibration

0.03352668 0.04006216 0.04659764

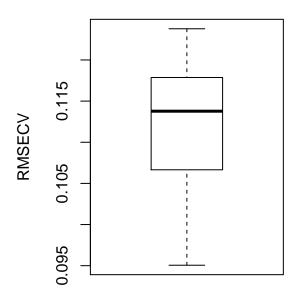
user system elapsed ## 0.008 0.000 8.171

7 A variable elimination method to improve the parsimony of MLR models using the successive projections algorithm

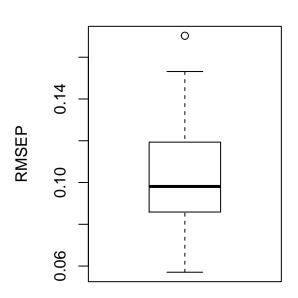
60-20 None (in)

3~m5 0.110(07)

PLS after scaled



PLS after scaled



Number of Calibration

Number of Calibration

0.03352668 0.04006216 0.04659764

user system elapsed ## 0.005 0.000 6.660

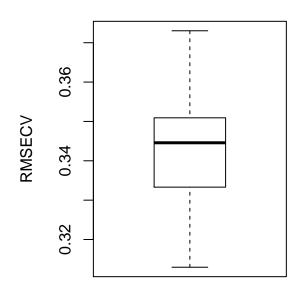
7 A variable elimination method to improve the parsimony of MLR models using the successive projections algorithm

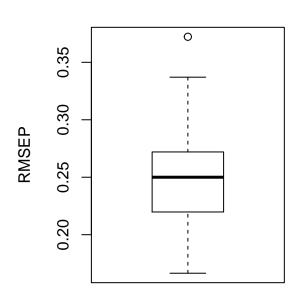
60-20 None (in)

 $4 \sim m5 \ 0.228(05)$









Number of Calibration

Number of Calibration

```
## 0.03352668 0.04006216 0.04659764
##
      user system elapsed
##
    0.005
           0.000
                    5.494
```

8 Using consensus interval partial least square in near infrared spectra analysis

```
52\text{-}26 Delete 75 , 77;\,10
```

1~m5 RMSECV=0.0124; RMSEP=0.0157

```
## 0.0200874 0.02142069 0.02275398
## 0.01678213 0.01947068 0.02215922
     user system elapsed
    0.003
           0.000
##
                    6.391
```

8 Using consensus interval partial least square in near infrared spectra analysis

```
52-26 Delete 75 , 77 10

2~m5 RMSECV=0.0613; RMSEP=0.0673

## 0.06526904 0.06992496 0.07458089

## 0.05731671 0.06727828 0.07723985

## user system elapsed

## 0.002 0.001 6.217
```

8 Using consensus interval partial least square in near infrared spectra analysis

```
52-26 Delete 75 , 77 10

3~m5 RMSECV=0.1080; RMSEP=0.1353

## 0.1247879 0.1383133 0.1518387

## 0.105457 0.1255203 0.1455836

## user system elapsed

## 0.002 0.000 6.136
```

8 Using consensus interval partial least square in near infrared spectra analysis

```
52-26 Delete 75 , 77 10

4~m5 RMSECV=0.2579; RMSEP=0.2356

## 0.2545005 0.2833188 0.312137

## 0.2512721 0.2812153 0.3111585

## user system elapsed

## 0.002 0.001 6.098
```