

# 主成分与偏最小二乘回归的效果对比

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2020-04-17

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## 1 简介

此篇意在比较 PCR（主成分分析）和 PLS（偏最小二乘回归）的效果，得出在什么情况下哪种方法更为合适。

模拟了一个  $n$  个观测值， $p$  个变量，变量之间相关系数为  $\rho$  的数据集，通过  $\beta_0$  和  $\beta_1$  加上一个标准正态分布的残差模拟出被解释变量。

## 2 说明

在 `comparison.py` 中定义了一个 `comparison` 函数，用于输出 PCR 和 PLS 的指标对比，分别包括：

- 交叉验证中的测试误差

- 成分的个数（交叉验证中的测试误差取到最小时）
- 对 Y 的解释程度（在此成分个数下）

### 3 结论

相比于 PCR, PLS 在以下情况的表现更佳:

- 变量个数更多
- 变量之间相关系数较小
- 各个变量的系数较大（变量对结果的影响较大）

### 4 模拟过程

```
library(knitr)
library(reticulate)
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.2.1 --
```

```
## v ggplot2 3.2.1      v purrr  0.3.4
## v tibble  2.1.1      v dplyr  0.8.3
## v tidyr   1.0.2      v stringr 1.4.0
## v readr   1.3.1      v forcats 0.4.0
```

```
## Warning: package 'purrr' was built under R version 3.6.2
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

```
use_python("/usr/local/bin/python3")
py_config()
```

```
## python:          /usr/local/bin/python3
```

```
## libpython:       /usr/local/opt/python@3.9/Frameworks/Python.framework/Versions/3.9/libpython3.9.dylib
```

```
## pythonhome:      /usr/local/opt/python@3.9/Frameworks/Python.framework/Versions/3.9:/
## version:         3.9.1 (default, Jan  6 2021, 06:05:23) [Clang 12.0.0 (clang-1200.0.
## numpy:           /usr/local/lib/python3.9/site-packages/numpy
## numpy_version:   1.21.1
##
## python versions found:
## /usr/local/bin/python3
## /Users/ethan/.virtualenvs/r-reticulate/bin/python
## /usr/bin/python
## /usr/bin/python3
```

```
import numpy as np
import pandas as pd
from scipy.stats import norm
from src.scale import scale
from src.sim import sim
from model.comparison import comparison
```

#### 4.1 变化 - p

```
n, p, rho = 1000, 10, 0.5
mu = norm.rvs(size=p, scale=1)
beta0, beta1 = 0.5, 0.5 * np.ones(p, dtype=float)
comparison(n, p, rho, mu, beta0, beta1)
```

```
##   methods  n components  test error  variation explanation
## 0      PCR              6    1.040492          0.869771
## 1      PLS              2    1.040659          0.869798
```

```
n, p, rho = 1000, 30, 0.5
mu = norm.rvs(size=p, scale=1)
beta0, beta1 = 0.5, 0.5 * np.ones(p, dtype=float)
comparison(n, p, rho, mu, beta0, beta1)
```

```
## methods n components test error variation explanation
## 0 PCR 28 1.053451 0.956126
## 1 PLS 3 1.052155 0.956072
```

```
n, p, rho = 1000, 50, 0.5
mu = norm.rvs(size=p, scale=1)
beta0, beta1 = 0.5, 0.5 * np.ones(p, dtype=float)
comparison(n, p, rho, mu, beta0, beta1)
```

```
## methods n components test error variation explanation
## 0 PCR 34 1.016787 0.973011
## 1 PLS 4 1.032042 0.973507
```

## 4.2 变化 - rho

```
n, p, rho = 1000, 30, 0.25
mu = norm.rvs(size=p, scale=1)
beta0, beta1 = 0.5, 0.5 * np.ones(p, dtype=float)
comparison(n, p, rho, mu, beta0, beta1)
```

```
## methods n components test error variation explanation
## 0 PCR 29 1.038088 0.923840
## 1 PLS 2 1.033148 0.923736
```

```
n, p, rho = 1000, 30, 0.5
mu = norm.rvs(size=p, scale=1)
beta0, beta1 = 0.5, 0.5 * np.ones(p, dtype=float)
comparison(n, p, rho, mu, beta0, beta1)
```

```
## methods n components test error variation explanation
## 0 PCR 29 0.992817 0.959971
## 1 PLS 4 0.991414 0.959945
```

```
n, p, rho = 1000, 30, 0.75
mu = norm.rvs(size=p, scale=1)
beta0, beta1 = 0.5, 0.5 * np.ones(p, dtype=float)
comparison(n, p, rho, mu, beta0, beta1)
```

```
## methods n components test error variation explanation
## 0 PCR 18 0.968682 0.979205
## 1 PLS 4 0.971065 0.979280
```

### 4.3 变化 - beta

```
n, p, rho = 1000, 30, 0.5
mu = norm.rvs(size=p, scale=1)
beta0, beta1 = 0.1, 0.1 * np.ones(p, dtype=float)
comparison(n, p, rho, mu, beta0, beta1)
```

```
## methods n components test error variation explanation
## 0 PCR 8 1.008096 0.475111
## 1 PLS 0 1.012978 0.474091
```

```
n, p, rho = 1000, 30, 0.5
mu = norm.rvs(size=p, scale=1)
beta0, beta1 = 0.5, 0.5 * np.ones(p, dtype=float)
comparison(n, p, rho, mu, beta0, beta1)
```

```
## methods n components test error variation explanation
## 0 PCR 23 0.968686 0.958227
## 1 PLS 4 0.970101 0.958516
```

```
n, p, rho = 1000, 30, 0.5
mu = norm.rvs(size=p, scale=1)
beta0, beta1 = 1, 1 * np.ones(p, dtype=float)
comparison(n, p, rho, mu, beta0, beta1)
```

##	methods	n components	test error	variation explanation
## 0	PCR	26	0.950735	0.990006
## 1	PLS	4	0.953445	0.990016