

▼ CCA

In PCA we looked at only the input X to carry out dimensionality reduction and is unsupervised in that it does not incorporate the labels y into the analysis.

```
import numpy as np
from sklearn.cross_decomposition import CCA

# some dummy data for showing the process of running CCA
X = np.array([[0., 0., 1.], [1., 0., 0.], [2., 2., 2.], [3., 5., 4.]])
y = np.array([[0.1, -0.2], [0.9, 1.1], [6.2, 5.9], [11.9, 12.3]])
print(X)
print(y)
```

```
↳ [[0. 0. 1.]
    [1. 0. 0.]
    [2. 2. 2.]
    [3. 5. 4.]]
   [[ 0.1 -0.2]
    [ 0.9  1.1]
    [ 6.2  5.9]
    [11.9 12.3]]
```

```
cca = CCA(n_components=1)
cca.fit(X, y)
```

```
↳ CCA(copy=True, max_iter=500, n_components=1, scale=True, tol=1e-06)
```

```
# more useful than simply the fit, the actual transformation from our computed CCA
X_c, y_c = cca.fit_transform(X, y)
print(X_c)
print(y_c)
```

```
↳ [[-1.3373174 ]
    [-1.10847164]
    [ 0.40763151]
    [ 2.03815753]]
   [[-0.85511537]
    [-0.70878547]
    [ 0.26065014]
    [ 1.3032507 ]]
```

To demonstrate why CCA is more powerful, let's examine scaling the data.

```
X_c, y_c = cca.fit_transform(2*X, 2*y)
print(X_c)
print(y_c)
```

```

↳ [[ -1.3373174 ]
    [-1.10847164]
    [ 0.40763151]
    [ 2.03815753]]
   [[ -0.85511537]
    [-0.70878547]
    [ 0.26065014]
    [ 1.3032507 ]]

```

As we see, scaling both by 2 doesn't change the transform at all. Going further:

```

X_c, y_c = cca.fit_transform(2*X, 3*y)
print(X_c)
print(y_c)

```

```

↳ [[ -1.3373174 ]
    [-1.10847164]
    [ 0.40763151]
    [ 2.03815753]]
   [[ -0.85511537]
    [-0.70878547]
    [ 0.26065014]
    [ 1.3032507 ]]

```

Even scaling separately, the transformation remains the same. This shows a very important property of linear transformations (scaling or addition)

Proof:

```

from IPython.display import Image
Image(filename="/content/data/screenshot.PNG")

```

$$\begin{aligned}
 \rho(aX + b, cY + d) &= \frac{\text{Cov}(aX + b, cY + d)}{\sqrt{\text{Var}(aX + b) \text{Var}(cY + d)}} \\
 &= \frac{\text{Cov}(aX, cY)}{\sqrt{\text{Var}(aX) \text{Var}(cY)}} \\
 &= \frac{a \cdot c \cdot \text{Cov}(X, Y)}{\sqrt{a^2 \text{Var}(X) \cdot c^2 \text{Var}(Y)}} \\
 &= \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X) \text{Var}(Y)}} \\
 &= \rho(X, Y)
 \end{aligned}$$

Source: EECS 189 Course Notes for proof