01. 비지도 학습 예제 - 붓꽃

붓꽃 데이터는 몇 차원인가?
(1) 우리는 4차원이나 그 이상보다 2차원 데이터를 플로팅(그래프)하는 것이 쉽다.

Principal component analysis (PCA)

http://scikit-learn.org/stable/modules/generated/sklearn.decomposition.PCA.html (http://scikit-learn.org/stable/modules/generated/sklearn.decomposition.PCA.html)

In [20]: from sklearn.decomposition import PCA #모델 클래스 선택 import seaborn as sns %matplotlib inline

```
In [25]: iris = sns.load_dataset('iris')
    iris.head()
    print(type(iris))
    X_iris = iris.drop('species', axis=1)
    y_iris = iris['species']
    X_iris
```

<class 'pandas.core.frame.DataFrame'>

Ο.	. т.	$\Gamma \cap \Gamma$	١.
UL	ΙŢ	25	

25]:		sepal_length	sepal_width	petal_length	petal_width
	0	5.1	3.5	1.4	0.2
	1	4.9	3.0	1.4	0.2
	2	4.7	3.2	1.3	0.2
	3	4.6	3.1	1.5	0.2
	4	5.0	3.6	1.4	0.2
	5	5.4	3.9	1.7	0.4
	6	4.6	3.4	1.4	0.3
	7	5.0	3.4	1.5	0.2
	8	4.4	2.9	1.4	0.2
	9	4.9	3.1	1.5	0.1
	10	5.4	3.7	1.5	0.2
	11	4.8	3.4	1.6	0.2
	12	4.8	3.0	1.4	0.1
	13	4.3	3.0	1.1	0.1
	14	5.8	4.0	1.2	0.2
	15	5.7	4.4	1.5	0.4
	16	5.4	3.9	1.3	0.4
	17	5.1	3.5	1.4	0.3
	18	5.7	3.8	1.7	0.3
	19	5.1	3.8	1.5	0.3

	sepal_length	sepal_width	petal_length	petal_width
20	5.4	3.4	1.7	0.2
21	5.1	3.7	1.5	0.4
22	4.6	3.6	1.0	0.2
23	5.1	3.3	1.7	0.5
24	4.8	3.4	1.9	0.2
25	5.0	3.0	1.6	0.2
26	5.0	3.4	1.6	0.4
27	5.2	3.5	1.5	0.2
28	5.2	3.4	1.4	0.2
29	4.7	3.2	1.6	0.2
120	6.9	3.2	5.7	2.3
121	5.6	2.8	4.9	2.0
122	7.7	2.8	6.7	2.0
123	6.3	2.7	4.9	1.8
124	6.7	3.3	5.7	2.1
125	7.2	3.2	6.0	1.8
126	6.2	2.8	4.8	1.8
127	6.1	3.0	4.9	1.8
128	6.4	2.8	5.6	2.1
129	7.2	3.0	5.8	1.6
130	7.4	2.8	6.1	1.9
131	7.9	3.8	6.4	2.0
132	6.4	2.8	5.6	2.2
133	6.3	2.8	5.1	1.5
134	6.1	2.6	5.6	1.4

	sepal_length	sepal_width	petal_length	petal_width
135	7.7	3.0	6.1	2.3
136	6.3	3.4	5.6	2.4
137	6.4	3.1	5.5	1.8
138	6.0	3.0	4.8	1.8
139	6.9	3.1	5.4	2.1
140	6.7	3.1	5.6	2.4
141	6.9	3.1	5.1	2.3
142	5.8	2.7	5.1	1.9
143	6.8	3.2	5.9	2.3
144	6.7	3.3	5.7	2.5
145	6.7	3.0	5.2	2.3
146	6.3	2.5	5.0	1.9
147	6.5	3.0	5.2	2.0
148	6.2	3.4	5.4	2.3
149	5.9	3.0	5.1	1.8

150 rows × 4 columns

```
In [36]: model = PCA(n_components=2) # 유지할 구성 요소수. 없으면 전체 선택 model.fit(X_iris) # y는 지정 안함. 데이터 학습 X_2D = model.transform(X_iris) # 데이터를 2차원으로 변환
```

```
In [37]: print(X_iris.head())
         print(X_2D)
            sepal_length sepal_width petal_length petal_width
                                                            0.2
         0
                     5.1
                                  3.5
                                               1.4
                     4.9
                                 3.0
                                               1.4
                                                            0.2
         2
                     4.7
                                 3.2
                                               1.3
                                                            0.2
                                               1.5
                                                            0.2
                     4.6
                                 3.1
                                                            0.2
                     5.0
                                 3.6
                                               1.4
         [[-2.68412563 0.31939725]
          [-2.71414169 -0.17700123]
          [-2.88899057 -0.14494943]
          [-2.74534286 -0.31829898]
          [-2.72871654 0.32675451]
          [-2.28085963 0.74133045]
          [-2.82053775 -0.08946138]
          [-2.62614497 0.16338496]
          [-2.88638273 -0.57831175]
          [-2.6727558 -0.11377425]
          [-2.50694709 0.6450689 ]
          [-2.61275523 0.01472994]
          [-2.78610927 -0.235112
          [ 0 00000074 0 54400450]
```

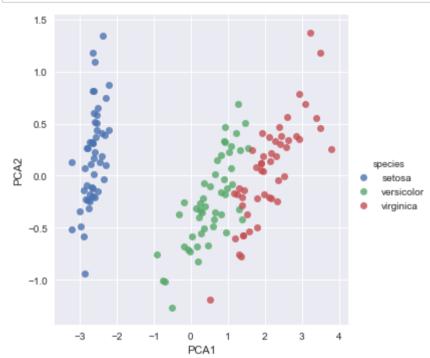
```
# 첫번째 열 데이터
In [23]: X_2D[: , 0]
Out [23]: array([-2.68412563, -2.71414169, -2.88899057, -2.74534286, -2.72871654,
                 -2.28085963. -2.82053775. -2.62614497. -2.88638273. -2.6727558
                 -2.50694709. -2.61275523. -2.78610927. -3.22380374. -2.64475039.
                 -2.38603903, -2.62352788, -2.64829671, -2.19982032, -2.5879864
                 -2.31025622, -2.54370523, -3.21593942, -2.30273318, -2.35575405
                 -2.50666891, -2.46882007, -2.56231991, -2.63953472, -2.63198939
                 -2.58739848, -2.4099325 , -2.64886233, -2.59873675, -2.63692688
                 -2.86624165, -2.62523805, -2.80068412, -2.98050204, -2.59000631
                 -2.77010243. -2.84936871. -2.99740655. -2.40561449. -2.20948924
                 -2.71445143, -2.53814826, -2.83946217, -2.54308575, -2.70335978
                  1.28482569.
                               0.93248853,
                                            1.46430232.
                                                          0.18331772.
                                                                      1.08810326
                  0.64166908.
                               1.09506066, -0.74912267,
                                                          1.04413183, -0.0087454
                 -0.50784088,
                               0.51169856.
                                             0.26497651.
                                                          0.98493451, -0.17392537
                  0.92786078,
                               0.66028376,
                                            0.23610499,
                                                          0.94473373.
                                                                       0.04522698
                  1.11628318.
                               0.35788842,
                                             1.29818388,
                                                          0.92172892.
                                                                       0.71485333
                  0.90017437.
                               1.33202444.
                                             1.55780216.
                                                          0.81329065, -0.30558378
                 -0.06812649, -0.18962247,
                                            0.13642871,
                                                          1.38002644.
                                                                       0.58800644
                  0.80685831.
                               1.22069088.
                                            0.81509524.
                                                          0.24595768,
                                                                       0.16641322
                  0.46480029.
                               0.8908152
                                            0.23054802. -0.70453176.
                                                                       0.35698149
                  0.33193448,
                               0.37621565.
                                             0.64257601, -0.90646986,
                                                                       0.29900084
                                            2.61667602,
                  2.53119273,
                               1.41523588,
                                                          1.97153105.
                                                                       2.35000592
                  3.39703874.
                               0.52123224,
                                             2.93258707.
                                                          2.32122882.
                                                                       2.91675097.
                  1.66177415.
                               1.80340195.
                                             2.1655918 .
                                                          1.34616358.
                                                                       1.58592822
                  1.90445637,
                               1.94968906.
                                             3.48705536.
                                                          3.79564542,
                                                                       1.30079171.
                                                          1.38876613,
                               1.19900111,
                  2.42781791,
                                             3.49992004
                                                                       2.2754305
                  2.61409047,
                               1.25850816,
                                             1.29113206.
                                                          2.12360872,
                                                                       2.38800302
                                             2.15943764.
                  2.84167278,
                               3.23067366,
                                                          1.44416124.
                                                                       1.78129481.
                  3.07649993.
                               2.14424331,
                                             1.90509815.
                                                          1.16932634.
                                                                       2.10761114,
                  2.31415471,
                               1.9222678
                                            1.41523588,
                                                          2.56301338.
                                                                       2.41874618
                                                         1.90094161.
                  1.94410979.
                               1.52716661, 1.76434572,
                                                                       1.390188861)
```

sns.Implot

```
sns.lmplot(x, y, # 입력 variable
hue="" # 데이터의 일부 집합을 정의
data=iris # 데이터 프레임
fit_reg=False # TRUE : 회귀 모델을 추정하고 플롯한다.
```

https://seaborn.pydata.org/generated/seaborn.lmplot.html (https://seaborn.pydata.org/generated/seaborn.lmplot.html)

```
In [48]: iris['PCA1'] = X_2D[: ,0] # feature 생성 iris['PCA2'] = X_2D[: ,1] # feature 생성 sns.Implot("PCA1", "PCA2", hue="species", data=iris, fit_reg=False);
```



붓꽃(Iris)에 대한 정보가 없음에도 2차원 표현에서 종(Species)가 매우 잘 분리되어 있다.

In []: