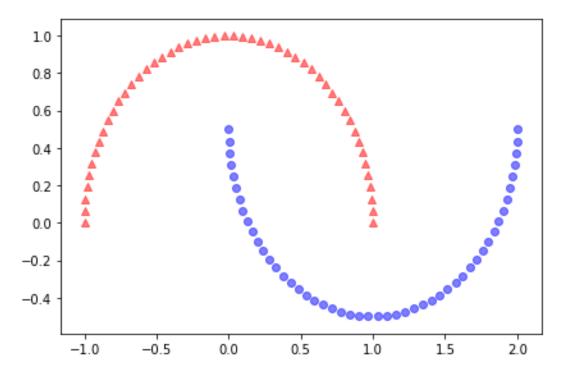
# **KPCAimplementation**

#### 2019年3月5日

### 1 KPCA 実装

```
In [37]: from scipy.spatial.distance import pdist,squareform
         from scipy import exp
         from scipy.linalg import eigh
         import numpy as np
         from sklearn.datasets import make_moons
         from sklearn.datasets import make_circles
         from sklearn.decomposition import PCA
         from matplotlib.ticker import FormatStrFormatter
         import matplotlib.pyplot as plt
         from sklearn.decomposition import KernelPCA
In [38]: # difine Kernel PCA
         def rbf_kernel_pca(x,gamma,n_components):
             # difine Kernel matrix
             sq_dists=pdist(x,'sqeuclidean')
             mat_sq_dists=squareform(sq_dists)
             K=exp(-gamma*mat_sq_dists)
             # centralize Kernal matrix
             N=K.shape[0]
             one_n=np.ones((N,N))/N
             K=K-one_n.dot(K)-K.dot(one_n)+one_n.dot(K).dot(one_n)
             # calculate eigen fancots & sort into descending order
             eigvals,eigvecs=eigh(K)
             eigvals,eigvecs=eigvals[::-1],eigvecs[:,::-1]
             # select k eigen factors
             alphas=np.column_stack((eigvecs[:,i] for i in range(n_components)))
             # select eigen values
             lambdas=[eigvals[i] for i in range(n_components)]
             return alphas, lambdas
```

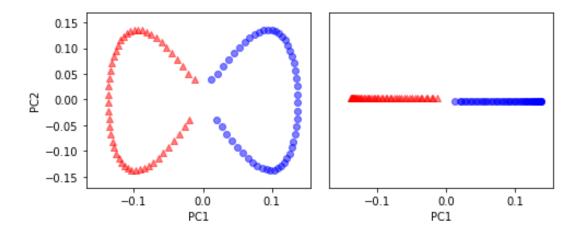
### 1.1 例 1 半月形データの射影 KPCA vs PCA



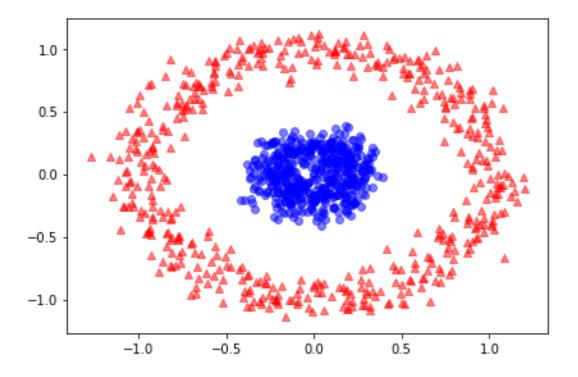
### 1.1.1 PCA ではどうなるか

```
ax[1].set_ylim([-1,1])
 ax[1].set_yticks([])
 ax[1].set_xlabel('PC1')
 plt.tight_layout()
 plt.show()
 0.75
 0.50
 0.25
 0.00
-0.25
-0.50
-0.75
           -1
                       ò
                                  i
                                                 -1
                                                             ò
                                                                        i
                      PC1
                                                            PC1
```

#### 1.1.2 KPCA ではどうなるか



### 1.2 例 2 同心円状データの射影 KPCA vs PCA



#### 1.2.1 PCA ではどうなるか

```
In [43]: # how about PCA?
         scikit_pca=PCA(n_components=2)
         x_spca=scikit_pca.fit_transform(x)
         fig, ax=plt.subplots(nrows=1,ncols=2,figsize=(7,3))
         fig.set_facecolor('white')
         ax[0].scatter(x_spca[y==0,0],x_spca[y==0,1],color='red',marker='\^',alpha=0.5)
         ax[0].scatter(x_spca[y==1,0],x_spca[y==1,1],color='blue',marker='o',alpha=0.5)
         ax[1].scatter(x_spca[y=0,0],np.zeros((500,1))+0.02,color='red',marker='^',alpha=0.5)
         ax[1].scatter(x_spca[y==1,0],np.zeros((500,1))-0.02,color='blue',marker='o',alpha=0.5)
         ax[0].set_xlabel('PC1')
         ax[0].set_ylabel('PC2')
         ax[1].set_ylim([-1,1])
         ax[1].set_yticks([])
         ax[1].set_xlabel('PC1')
         plt.tight_layout()
         plt.show()
         1.0
         0.5
         0.0
       -0.5
```

#### 1.2.2 KPCA ではどうなるか

-1.0

0.0

PC1

0.5

1.0

-1.0

-0.5

0.0

PC1

0.5

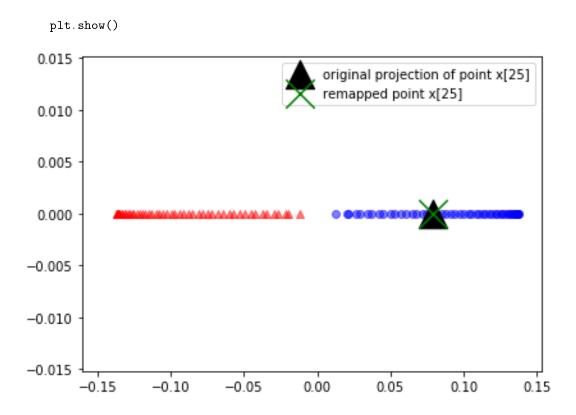
1.0

```
ax[1].set_ylim([-1,1])
 ax[1].set_yticks([])
 ax[1].set_xlabel('PC1')
 plt.tight_layout()
 plt.show()
 0.075
 0.050
 0.025
 0.000
-0.025
-0.050
-0.075
                                                -0.04 -0.02 0.00
         -0.04 -0.02 0.00
                          0.02
                                 0.04
                                      0.06
                                                                  0.02
                                                                        0.04
                                                                              0.06
                                                               PC1
```

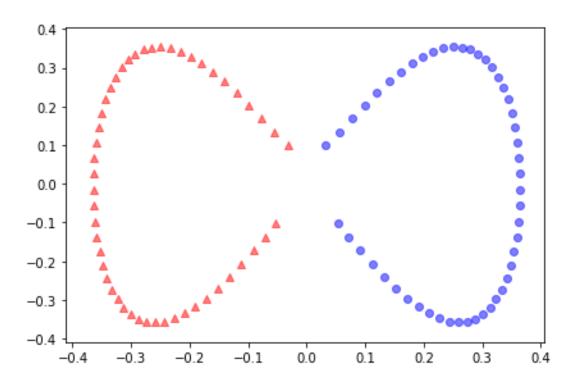
### 2 KPCA での New data に対する扱い方

トレーニングデータとのカーネル類似関数と固有値による正規化

```
In [45]: # How to deal with New data? --> caluculate K between training and new ones
         x,y=make_moons(n_samples=100,random_state=123)
         alphas,lambdas=rbf_kernel_pca(x,gamma=15,n_components=1)
         x_new=x[25]
         x_proj=alphas[25]
         def project_x(x_new,x,gamma,alphas,lambdas):
             pair_dist=np.array([np.sum((x_new-row)**2)for row in x])
             k=np.exp(-gamma*pair_dist)
             return k.dot(alphas/lambdas)
         x_reproj=project_x(x_new,x,gamma=15,alphas=alphas,lambdas=lambdas)
         fig, ax=plt.subplots()
         fig.set_facecolor('white')
         ax.scatter(alphas[y==0,0],np.zeros((50)),color='red',marker='^',alpha=0.5)
         ax.scatter(alphas[y==1,0],np.zeros((50)),color='blue',marker='o',alpha=0.5)
         ax.scatter(x_proj,0,color='black',label='original projection of point x[25]',marker='^',s=5
         ax.scatter(x_reproj,0,color='green',label='remapped point x[25]',marker='x',s=500)
         plt.legend(scatterpoints=1)
         plt.tight_layout()
```



## 3 Sklearn での実装



In [47]: