

# Programming assignment 10: Dimensionality Reduction¶

In [40]:

```
import numpy as np
import matplotlib.pyplot as plt

%matplotlib inline
```

## PCA Task¶

Given the data in the matrix  $X$  your tasks is to:

- Calculate the covariance matrix  $\Sigma$ .
- Calculate eigenvalues and eigenvectors of  $\Sigma$ .
- Plot the original data  $X$  and the eigenvectors to a single diagram. What do you observe? Which eigenvector corresponds to the smallest eigenvalue?
- Determine the smallest eigenvalue and remove its corresponding eigenvector. The remaining eigenvector is the basis of a new subspace.
- Transform all vectors in  $X$  in this new subspace by expressing all vectors in  $X$  in this new basis.

## The given data $X$ ¶

In [41]:

```
X = np.array([(-3, -2), (-2, -1), (-1, 0), (0, 1),
              (1, 2), (2, 3), (-2, -2), (-1, -1),
              (0, 0), (1, 1), (2, 2), (-2, -3),
              (-1, -2), (0, -1), (1, 0), (2, 1), (3, 2)])
```

## Task 1: Calculate the covariance matrix $\Sigma$ ¶

In [42]:

```
def get_covariance(X):
    """Calculates the covariance matrix of the input data.

    Parameters
    -----
    X : array, shape [N, D]
        Data matrix.

    Returns
    -----
    Sigma : array, shape [D, D]
```

Covariance matrix

```
"""  
# TODO  
return np.cov(X.transpose())
```

## Task 2: Calculate eigenvalues and eigenvectors of $\Sigma$ .

In [43]:

```
def get_eigen(S):  
    """Calculates the eigenvalues and eigenvectors of the input matrix.  
  
    Parameters  
    -----  
    S : array, shape [D, D]  
        Square symmetric positive definite matrix.  
  
    Returns  
    -----  
    L : array, shape [D]  
        Eigenvalues of S  
    U : array, shape [D, D]  
        Eigenvectors of S  
  
    """  
    # TODO  
    return np.linalg.eigh(S)
```

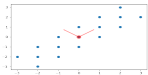
## Task 3: Plot the original data X and the eigenvectors to a single diagram.

In [44]:

```
# plot the original data  
plt.scatter(X[:, 0], X[:, 1])  
  
# plot the mean of the data  
mean_d1, mean_d2 = X.mean(0)  
plt.plot(mean_d1, mean_d2, 'o', markersize=10, color='red', alpha=0.5)  
  
# calculate the covariance matrix  
Sigma = get_covariance(X)  
  
# calculate the eigenvector and eigenvalues of Sigma
```

```
L, U = get_eigen(Sigma)
```

```
plt.arrow(mean_d1, mean_d2, U[0, 0], U[0, 1], width=0.01, color='red', alpha=0.5)
plt.arrow(mean_d1, mean_d2, U[1, 0], U[1, 1], width=0.01, color='red', alpha=0.5);
```



What do you observe in the above plot? Which eigenvector corresponds to the smallest eigenvalue?

Write your answer here:

[YOUR ANSWER]

## Task 4: Transform the data¶

Determine the smallest eigenvalue and remove its corresponding eigenvector. The remaining eigenvector is the basis of a new subspace. Transform all vectors in  $X$  in this new subspace by expressing all vectors in  $X$  in this new basis.

In [45]:

```
def transform(X, U, L):
    """Transforms the data in the new subspace spanned by the eigenvector
    corresponding to the largest eigenvalue.
```

Parameters

-----

$X$  : array, shape  $[N, D]$

Data matrix.

$L$  : array, shape  $[D]$

Eigenvalues of  $\text{Sigma}_X$

$U$  : array, shape  $[D, D]$

Eigenvectors of  $\text{Sigma}_X$

Returns

-----

$X_t$  : array, shape  $[N, 1]$

Transformed data

"""

# TODO

return None

In [46]:

```
 $X_t$  = transform( $X$ ,  $U$ ,  $L$ )
```

# Task SVD¶

**Task 5: Given the matrix  $M$  find its SVD decomposition  $M = U \cdot \Sigma \cdot V$  and reduce it to one dimension using the approach described in the lecture.¶**

In [47]:

```
M = np.array([[1, 2], [6, 3],[0, 2]])
```

In [48]:

```
def reduce_to_one_dimension(M):
    """Reduces the input matrix to one dimension using its SVD decomposition.

    Parameters
    -----
    M : array, shape [N, D]
        Input matrix.

    Returns
    -----
    M_t: array, shape [N, 1]
        Reduce matrix.

    """
    # TODO
    return None
```

In [49]:

```
M_t = reduce_to_one_dimension(M)
```