## outliers\_part\_one

## January 20, 2019

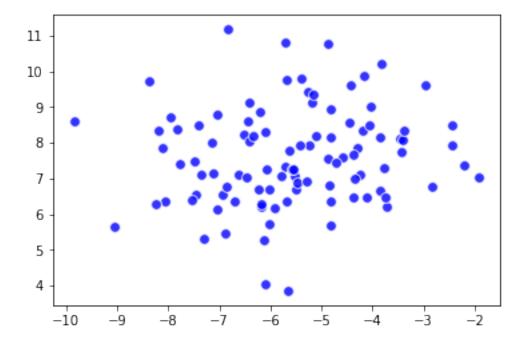
```
In [1]: # outliers values (one dimension)
                  from sklearn.datasets import load_boston
                  boston = load_boston()
                  continuous_variables = [n for n in range(boston.data.shape[1]) if n!=3]
                  print(continuous_variables)
[0, 1, 2, 4, 5, 6, 7, 8, 9, 10, 11, 12]
In [2]: import numpy as np
                  from sklearn import preprocessing
                  normalize_data = preprocessing.StandardScaler().fit_transform(boston.data[:,continuous]
                  ouliers_rows, outliter_columns = np.where(np.abs(normalize_data)>3)
                  print(ouliers_rows)
                  print(outliter_columns)
[ 55 56 57 102 141 199 200 201 202 203 204 225 256 257 262 283 284 347
  351 352 353 353 354 355 364 365 367 373 374 374 380 398 404 405 406 410
  410 411 412 412 414 414 415 416 418 418 419 423 424 425 426 427 427 429
 431 436 437 438 445 450 454 455 456 457 466]
[1 1 1 10 11 1 1 1 1 1 1 4 1 4 4 1 1 1 6 6 1
    In [3]: print(list(zip(ouliers_rows, outliter_columns)))
[(55, 1), (56, 1), (57, 1), (102, 10), (141, 11), (199, 1), (200, 1), (201, 1), (202, 1), (203, 1), (203, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204, 1), (204
In [4]: # EllipticEcvelop distribution
                  from sklearn.datasets import make_blobs
                  from sklearn.covariance import EllipticEnvelope
                  from matplotlib import pyplot as plt
                  def generate_distribiution(blobs):
                            blob = make_blobs(n_samples=100, n_features=2, centers=blobs, cluster_std=1.5, shu
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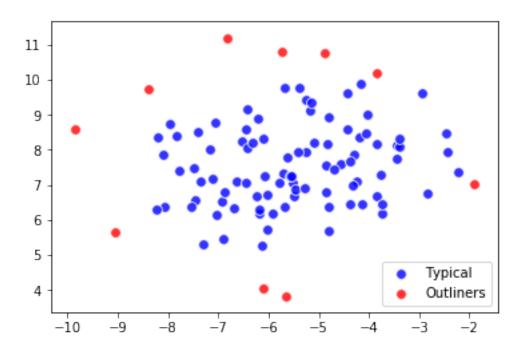
```
robust_covariance_est = EllipticEnvelope(contamination=.1).fit(blob[0])# contamina
    detection = robust_covariance_est.predict(blob[0])
    outliers = np.where(detection == -1)[0]
    inliers = np.where(detection == 1)[0]

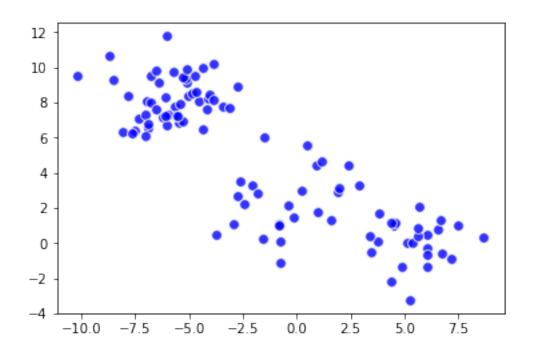
    plt.scatter(blob[0][:,0],blob[0][:,1], c='blue', alpha=0.8, s=60, marker='o', edged
    plt.show()
    return (outliers, inliers, blob)

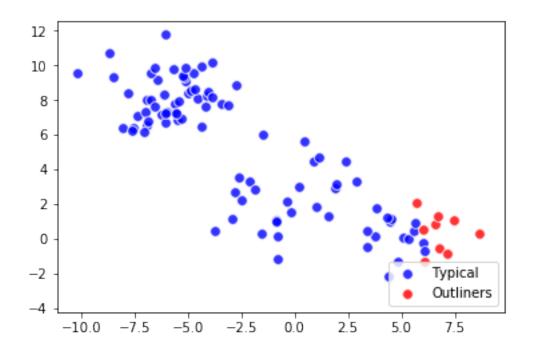
In [5]: # EllipticEcvelop distribution and outliers
    def plot outliers(outliers, inliers, blob):
```

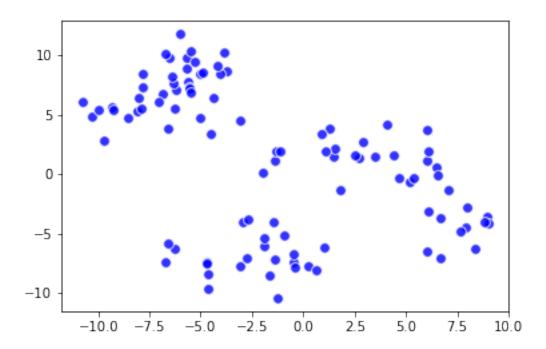
## 

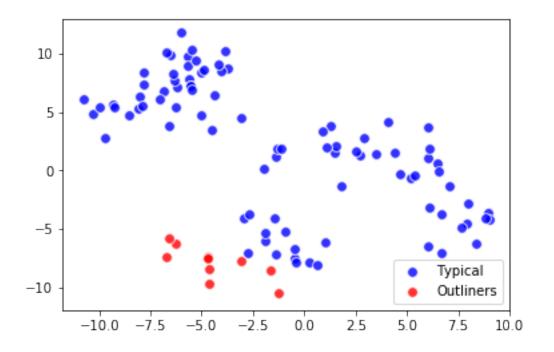


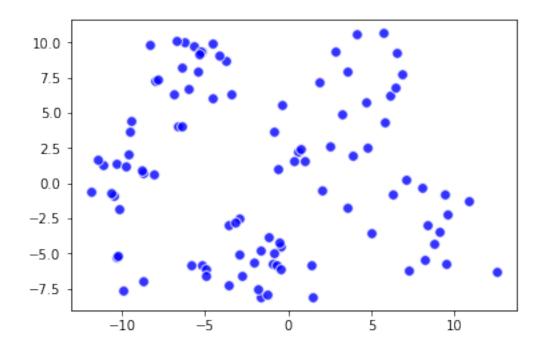


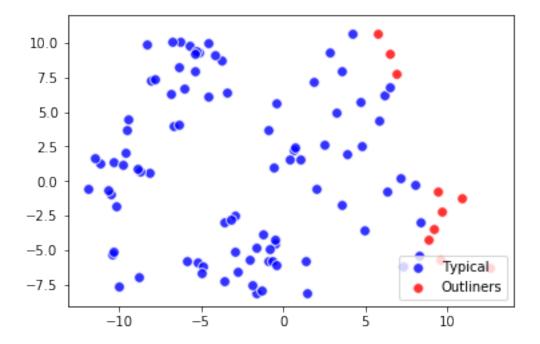






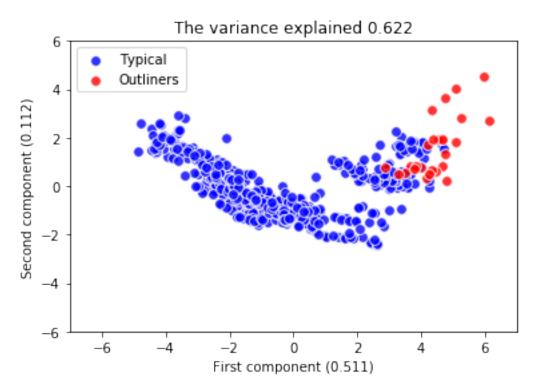






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In [11]: from sklearn.decomposition import PCA
         boston = load_boston()
         # outliers in boston data 1
         continuous_variables = [n for n in range(boston.data.shape[1]) if n!=3]
         normalized_data = preprocessing.StandardScaler().fit_transform(boston.data[:,continuo
         pca = PCA(n_components=2)
         Zscore_components = pca.fit_transform(normalize_data)
         vtot = "The variance explained " + str(round(np.sum(pca.explained_variance_ratio_),3)
         v1 = str(round(np.sum(pca.explained_variance_ratio_[0]),3))
               str(round(np.sum(pca.explained_variance_ratio_[1]),3))
In [12]: # outliers in boston data 2
         robust_covariance_est = EllipticEnvelope(store_precision=False, assume_centered = False)
         robust_covariance_est.fit(normalized_data)
         detection = robust_covariance_est.predict(normalized_data)
         outliers = np.where(detection == -1)
         regular = np.where(detection == 1)
In [13]: # outliers in boston data 3
         from matplotlib import pyplot as plt
         in_points = plt.scatter(Zscore_components[regular, 0], Zscore_components[regular, 1],
                                 color='blue', alpha=0.8, s=60, marker='o', edgecolors='white'
         out_points = plt.scatter(Zscore_components[outliers,0],Zscore_components[outliers,1],
                                 color='red', alpha=0.8, s=60, marker='o', edgecolors='white')
```

```
plt.legend((in_points,out_points),('Typical','Outliners'),scatterpoints=1, loc='best'
plt.xlabel('First component (%s)' % v1)
plt.ylabel('Second component (%s)' % v2)
plt.xlim([-7,7])
plt.ylim([-6,6])
plt.title(vtot) # EllipticEnvelope only shows outliers data in the one, smallest clus
plt.show()
```



## In []: