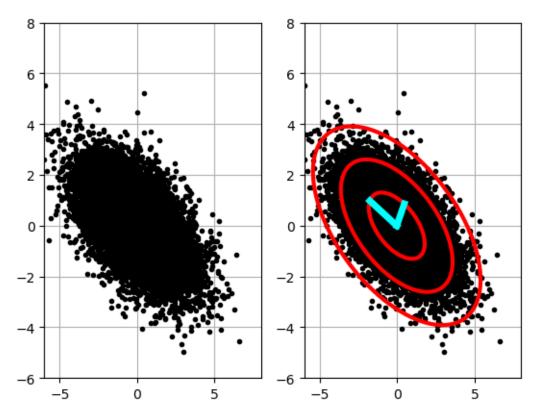
Metoda PCA na podstawie SVD

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Sprawozdanie
Matematyka Konkretna
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Laboratorium 4
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Wariant 10
Link do repozytorium: https://github.com/Maksiolo20/MK
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[5]: import matplotlib.pyplot as plt
     import numpy as np
     import pandas as pd
     # Read data from the CSV file
     data = pd.read_csv('10.csv', sep=',')
     # Extract the first row as the center of the data and principal axes
     center_and_axes = data.values.flatten()
     # Extracting values for the center of the data and principal axes
     center = center_and_axes[:1000] # Assuming the first 1000 values are for the_
      \hookrightarrow center
     axes = center_and_axes[1000:]
     theta = np.pi / 3 \# Rotate cloud by <math>pi/3
     R = np.array([[np.cos(theta), -np.sin(theta)], # Rotation matrix
     [np.sin(theta), np.cos(theta)]])
     nPoints = 10000 # Create 10,000 points
     sig = np.array([1.0, 2.0]) # Replace these values with your desired standard
      \rightarrow deviations
     xC = np.array([0.0, 0.0]) # Replace these values with your desired center
     X = R @ np.diag(sig) @ np.random.randn(2, nPoints) + np.diag(xC) @ np.ones((2, __
      ⇔nPoints))
     fig = plt.figure()
     ax1 = fig.add_subplot(121)
     ax1.plot(X[0, :], X[1, :], '.', color='k')
     ax1.grid()
     plt.xlim((-6, 8))
     plt.ylim((-6, 8))
     Xavg = np.mean(X, axis=1) # Compute mean
     B = X - np.tile(Xavg, (nPoints, 1)).T # Mean-subtracted data
     # Find principal components (SVD)
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U, S, VT = np.linalg.svd(B / np.sqrt(nPoints), full_matrices=0)
ax2 = fig.add_subplot(122)
ax2.plot(X[0, :], X[1, :], '.', color='k') # Plot data to overlay PCA
ax2.grid()
plt.xlim((-6, 8))
plt.ylim((-6, 8))
theta = 2 * np.pi * np.arange(0, 1, 0.01)
# 1-std confidence interval
Xstd = U @ np.diag(S) @ np.array([np.cos(theta), np.sin(theta)])
ax2.plot(Xavg[0] + Xstd[0, :], Xavg[1] + Xstd[1, :], '-', color='r',
linewidth=3)
ax2.plot(Xavg[0] + 2 * Xstd[0, :], Xavg[1] + 2 * Xstd[1, :], '-',
color='r', linewidth=3)
ax2.plot(Xavg[0] + 3 * Xstd[0, :], Xavg[1] + 3 * Xstd[1, :], '-',
color='r', linewidth=3)
# Plot principal components U[:,0]S[0] and U[:,1]S[1]
ax2.plot(np.array([Xavg[0], Xavg[0] + U[0, 0] * S[0]]),
np.array([Xavg[1], Xavg[1] + U[1, 0] * S[0]]), '-',
color='cyan', linewidth=5)
ax2.plot(np.array([Xavg[0], Xavg[0] + U[0, 1] * S[1]]),
np.array([Xavg[1], Xavg[1] + U[1, 1] * S[1]]), '-',
color='cyan', linewidth=5)
plt.show()
```



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