

# evaluation

October 7, 2024

## 1 Evaluation

### 1.1 Scalability with data size

Tested on a 20-dimensional data set containing 5 clusters i 5 different subspaces. 10% of the data was added as noise records.

SUBCLU stopped to work after 200,000 points. CLIQUE stopped to work after 7mio points.

#### 1.1.1 Settings

CLIQUE: - xsi: 25 - tau: 0.08

SUBCLU: - epsilon 0.02 - minpts: 250 (for 10k points). Doubled for each step, so 500 minpts for 20k, 1000 minpts for 40k, etc.

MAFIA: -a 1.4 -b 0.35 -M 20 -n 1000 -V -timing

#### 1.1.2 Plot

```
[1]: import matplotlib.pyplot as plt
import matplotlib.ticker as ticker

# Data
size = [10000, 20000, 30000, 40000, 50000, 100000, 200000, 1000000, 2000000,
        ↪3000000, 4000000, 5000000, 6000000, 7000000, 8000000, 9000000, 10000000,
        ↪15000000]
clique = [62, 119, 155, 203, 235, 533, 1255, 8795, 16290,
          ↪38581, 50392, 62348, 127730, 142005]
mafia = [15, 25, 17, 22, 27, 31, 45, 155, 265,
         ↪414, 556, 637, 922, 1016, 1225, 1215, 1641]
subclu = [137, 477, 984, 1751, 3055, 12490, 67603]

# Figure with 3 subplots
fig, axs = plt.subplots(1, 3, figsize=(7.5, 2))

# CLIQUE (size in millions)
axs[0].plot([s / 1_000_000 for s in size[:len(clique)]], clique, 'ro-',
            ↪label='CLIQUE')
axs[0].set_xlabel('size (mio. of records)')
```

```

axs[0].set_ylabel('runtime (ms)')
axs[0].xaxis.set_major_locator(ticker.MaxNLocator(5))
axs[0].yaxis.set_major_locator(ticker.MaxNLocator(5))
axs[0].legend()

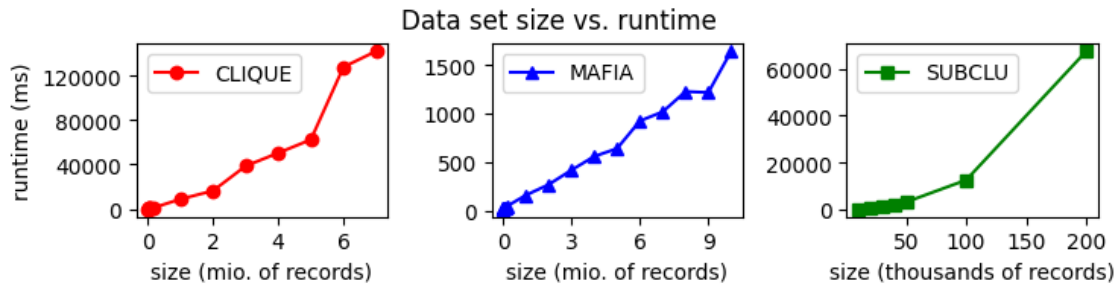
# MAFIA (size in millions)
axs[1].plot([s / 1_000_000 for s in size[:len(mafia)]], mafia, 'b^-',
            label='MAFIA')
axs[1].set_xlabel('size (mio. of records)')
axs[1].xaxis.set_major_locator(ticker.MaxNLocator(4))
axs[1].yaxis.set_major_locator(ticker.MaxNLocator(4))
axs[1].legend()

# SUBCLU (size in thousands)
axs[2].plot([s / 1000 for s in size[:len(subclu)]], subclu, 'gs-',
            label='SUBCLU')
axs[2].set_xlabel('size (thousands of records)')
axs[2].xaxis.set_major_locator(ticker.MaxNLocator(5))
axs[2].yaxis.set_major_locator(ticker.MaxNLocator(4))

axs[2].legend()

# Show plots
plt.suptitle('Data set size vs. runtime')
plt.tight_layout(rect=[0, 0, 1, 1.1])
plt.show()

```



## 1.2 Accuracy

Tested on a 10-dimensional data set containing 100,000 points for testing CLIQUE and MAFIA, however, SUBCLU was not able to handle this amount of data, so it was tested on a similar distribution of points with 20,000 points. 10% of the data was added as noise records.

The first case, has two clusters embedded in a different 4 dimensional subspace. Second case, has 4 clusters embedded in a different 4 dimensional subspace.

MAFIA reports the correct clusters in both cases with proper use of parameters. However, CLIQUE

reports...

### 1.2.1 2 clusters

Settings SUBCLU: - epsilon: 0.05 - minpts: 850

CLIQUE: - xsi: 20 - tau: 0.41 - prune: false

MAFIA: -a 3 -b 0.6 -M 20 -n 1000 -V -p -timing

#### Plot

```
[2]: import matplotlib.pyplot as plt
import matplotlib.image as mpimg

# Load images
clique_img = mpimg.imread('data-generator/datasets/mdcgen/accuracy/2clusters/
↳100k/clique/3d_plot.png')
mafia_img = mpimg.imread('data-generator/datasets/mdcgen/accuracy/2clusters/
↳100k/mafia/3d_plot.png')
subclu_img = mpimg.imread('data-generator/datasets/mdcgen/accuracy/2clusters/
↳20k/subclu/3d_plot.png')

# Figure with subplots
fig, axs = plt.subplots(1, 3, figsize=(7, 4))

# Plot images
axs[0].imshow(clique_img)
axs[0].axis('off')
axs[0].set_title('CLIQUE')

axs[1].imshow(mafia_img)
axs[1].axis('off')
axs[1].set_title('MAFIA')

axs[2].imshow(subclu_img)
axs[2].axis('off')
axs[2].set_title('SUBCLU')

# Title
plt.suptitle('Accuracy for 2 clusters in a 10-dimensional (3 dimensions shown)
↳data set')

# Show plot
plt.tight_layout(rect=[0, 0, 1, 1.1])
plt.show()
```

Accuracy for 2 clusters in a 10-dimensional (3 dimensions shown) data set



### 1.2.2 4 clusters

Settings SUBCLU: - epsilon: 0.02 - minpts: 500

CLIQUE: - xsi: 35 - tau: 0.2

MAFIA: -a 3 -b 0.6 -M 100 -n 1000 -V -p -timing

### Plot

```
[3]: import matplotlib.pyplot as plt
import matplotlib.image as mpimg

# Load images
clique_img = mpimg.imread('data-generator/datasets/mdcgen/accuracy/4clusters/
↳100k/clique/3d_plot.png')
mafia_img = mpimg.imread('data-generator/datasets/mdcgen/accuracy/4clusters/
↳100k/mafia/3d_plot.png')
subclu_img = mpimg.imread('data-generator/datasets/mdcgen/accuracy/4clusters/
↳20k/subclu/3d_plot.png')

# Figure with subplots
fig, axs = plt.subplots(1, 3, figsize=(7, 4))

# Plot images
axs[0].imshow(clique_img)
axs[0].axis('off')
axs[0].set_title('CLIQUE')

axs[1].imshow(mafia_img)
```

```

axs[1].axis('off')
axs[1].set_title('MAFIA')

axs[2].imshow(subclu_img)
axs[2].axis('off')
axs[2].set_title('SUBCLU')

# Title
plt.suptitle('Accuracy for 4 clusters in a 10-dimensional (3 dimensions shown) ↵
↳data set')

# Show plot
plt.tight_layout(rect=[0, 0, 1, 1.15])
plt.show()

```

Accuracy for 4 clusters in a 10-dimensional (3 dimensions shown) data set



### 1.2.3 Plus shape

Settings CLIQUE: - xsi: 10 - tau: 0.2

#### Plot

```

[4]: import matplotlib.pyplot as plt
import matplotlib.image as mpimg

# Load images
clique_img = mpimg.imread('data-generator/datasets/home_made/accuracy/
↳plus_shape/clique/plot.png')
mafia_img = mpimg.imread('data-generator/datasets/home_made/accuracy/plus_shape/
↳mafia/plot.png')

```

```

# Figure with subplots
fig, axs = plt.subplots(1, 2, figsize=(5, 4))

# Plot images
axs[0].imshow(clique_img)
axs[0].axis('off')
axs[0].set_title('CLIQUE')

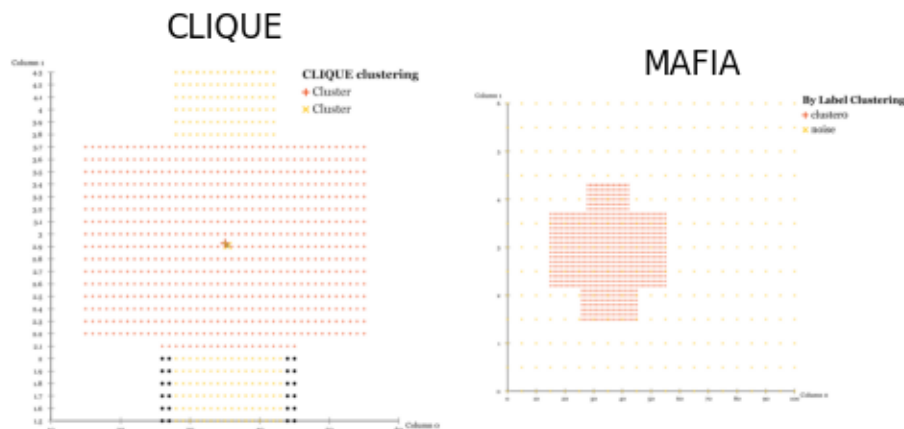
axs[1].imshow(mafia_img)
axs[1].axis('off')
axs[1].set_title('MAFIA')

# Title
plt.suptitle('Accuracy for a plus-shaped cluster in a 2-dimensional data set')

# Show plot
plt.tight_layout(rect=[0, 0, 1, 1.25])
plt.show()

```

## Accuracy for a plus-shaped cluster in a 2-dimensional data set



### 1.2.4 Bezier curve

Settings SUBCLU: - epsilon: 0.005 - minpts: 20

CLIQUE: - xsi: 20 - tau: 0.06

#### Plot

```

[5]: import matplotlib.pyplot as plt
import matplotlib.image as mpimg

# Load images

```

```

clique_img = mpimg.imread('data-generator/datasets/artificialCluster/accuracy/
↳bezier/clique/plot.png')
subclu_img = mpimg.imread('data-generator/datasets/artificialCluster/accuracy/
↳bezier/subclu/plot.png')

# Figure with subplots
fig, axs = plt.subplots(1, 2, figsize=(5, 4))

# Plot images
axs[0].imshow(clique_img)
axs[0].axis('off')
axs[0].set_title('CLIQUE')

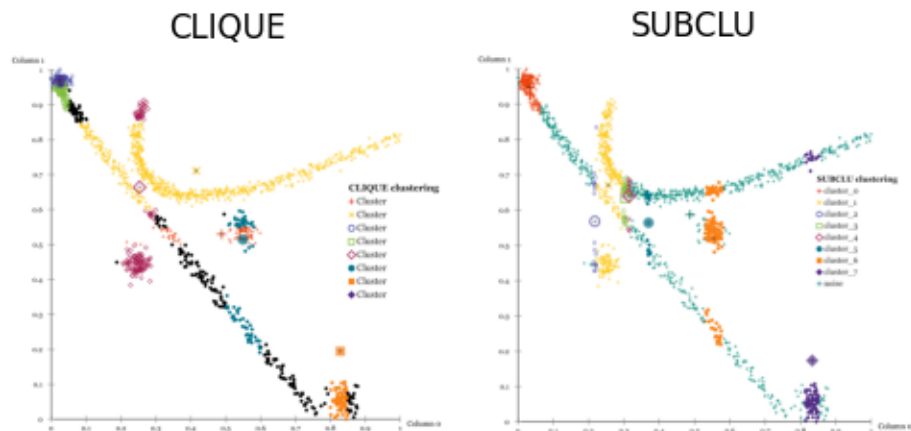
axs[1].imshow(subclu_img)
axs[1].axis('off')
axs[1].set_title('SUBCLU')

# Title
plt.suptitle('Accuracy for a bezier-shaped cluster in a 2-dimensional data set')

# Show plot
plt.tight_layout(rect=[0, 0, 1, 1.25])
plt.show()

```

Accuracy for a bezier-shaped cluster in a 2-dimensional data set



## 1.3 Cluster dimensionality

### 1.3.1 Settings

CLIQUE: - xsi: 30 - tau: 0.25

MAFIA: -a 2.2 -b 0.35 -M 40 -n 1000 -p -V -timing

### 1.3.2 Plot

```
[6]: import matplotlib.pyplot as plt
import matplotlib.ticker as ticker

# Data
dims = [2, 3, 4, 5, 6, 7, 8, 9, 10, 15, 20]
mafia = [45, 63, 55, 45, 63, 55, 60, 79, 86, 1387, 62607]
clique = [4440, 4347, 6017, 5740, 8093, 12532, 26373]

# Figure size
plt.figure(figsize=(2.5, 2.1))

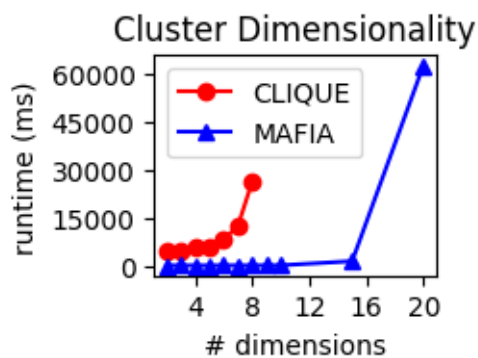
# Plot both in same figure
plt.plot(dims[:len(clique)], clique, 'ro-', label='CLIQUE')
plt.plot(dims, mafia, 'b^-', label='MAFIA')

# Set number of x-ticks
plt.gca().xaxis.set_major_locator(ticker.MaxNLocator(6))
plt.gca().yaxis.set_major_locator(ticker.MaxNLocator(6))

# Add labels and title
plt.xlabel('# dimensions')
plt.ylabel('runtime (ms)')
plt.title('Cluster Dimensionality')

# Add legend
plt.legend()

# Show plot
plt.tight_layout()
plt.show()
```





## 1.4 Data dimensionality

MAFIA and CLIQUE was on a 100 k dataset.

### 1.4.1 Settings

CLIQUE: - xsi: 30 - tau: 0.3 - prune: false

MAFIA: -a 2.2 -b 0.35 -M 40 -n 1000 -p -V -timing

### 1.4.2 Plot

```
[7]: import matplotlib.pyplot as plt
import matplotlib.ticker as ticker

# Data
dims = [10, 20, 30, 40, 50, 100]
mafia_runtime = [49, 72, 85, 104, 133, 317]
clique_runtime = [35688, 44180, 50275, 47682, 67772]

# CLIQUE:
# 10d = 66 clusters found
# 20d = 84
# 30d = 93
# 40d = 77
# 50d = 93
# 100d, not able to run.

# MAFIA:
# 10d = 3 clusters found
# 20d = 4 clusters found
# 30d = 3 clusters found
# 40d = 4 clusters found
# 50d = 3 clusters found
# 100d = 4 clusters found

# Figure with 2 subplots
fig, axs = plt.subplots(1, 2, figsize=(4, 2))

# CLIQUE
axs[0].plot(dims[:len(clique_runtime)], clique_runtime, 'ro-', label='CLIQUE')
axs[0].legend()
axs[0].xaxis.set_major_locator(ticker.MaxNLocator(5))

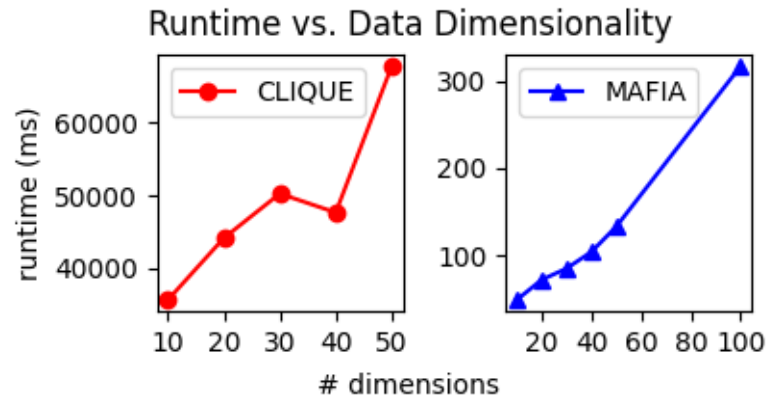
# MAFIA
axs[1].plot(dims, mafia_runtime, 'b^-', label='MAFIA')
axs[1].legend()
axs[1].xaxis.set_major_locator(ticker.MaxNLocator(5))
```

```

# Common x and y labels
fig.text(0.5, -0.02, '# dimensions', ha='center')
fig.text(-0.01, 0.5, 'runtime (ms)', va='center', rotation='vertical')

# Show plot
plt.suptitle('Runtime vs. Data Dimensionality')
plt.tight_layout(rect=[0, 0, 1, 1.1])
plt.show()

```



## 1.5 Sensitivity of alpha

As observed in article. Alpha and beta controls the number of clusters and their quality.

- 1,000,000 data points
- 20 dims
- 5 clusters
- 10% outliers
- noiseMatrix: 1 3 5 7 8

### 1.5.1 Settings

-b 0.35 -M 40 -n 1000

-a 0.8 -> 5.2 (step size: 0.4)

### 1.5.2 Plot

```

[8]: import matplotlib.pyplot as plt
import numpy as np

# Data
alpha = np.arange(1.2, 9.6, 0.4)
clusters_found = [23, 11, 5, 5, 5, 5, 5, 7, 7, 7, 7, 7, 7, 7, 7, 7, 22, 22, 22, 22,
                  22, 22]

```

```

# Figure size
plt.figure(figsize=(2, 2))

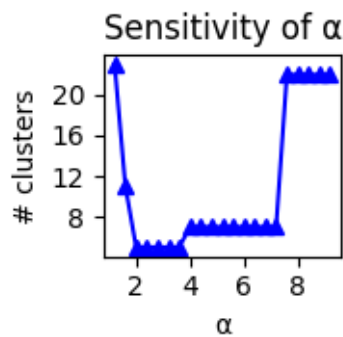
# Plot
plt.plot(alpha, clusters_found, 'b^-')

# Set number of x- and y-ticks
plt.gca().xaxis.set_major_locator(ticker.MaxNLocator(5))
plt.gca().yaxis.set_major_locator(ticker.MaxNLocator(5))

# Add labels and title
plt.xlabel(' ')
plt.ylabel('# clusters')
plt.title('Sensitivity of ')

# Show plot
plt.tight_layout()
plt.show()

```



## 2 Export to pdf

1. Install a LaTeX distribution.
  - Windows: MikTeX
  - Mac: MacTeX
  - Linux: TeX Live
2. Install pandoc (<https://pandoc.org/>), make sure it can be found using `pandoc --version`. Otherwise, try to re-open terminal.
3. Convert to pdf by running the following command: `jupyter nbconvert --to pdf evaluation.ipynb`.