

מעבדה לבינה מלאכותית שי בושנסקי – כלי ויזואליזציה למנוע הגנטי

הדוגמאות הבאות בפייטון ומתבססות על הספריות הבאות

```
import numpy as np
import matplotlib.pyplot as plt
import networkx as nx
import plotly.graph_objects as go
from plotly.subplots import make_subplots
import pandas as pd
from pandas.plotting import parallel_coordinates
from scipy.cluster.hierarchy import dendrogram, linkage
```

1. דוגמא להצגת פונקציית הפיטנס:

המטרה: מציאת מינימום לפרבולה (פ' בשני משתנים X ו Y)

כרומוזום מורכב משני גנים

```
# Define the fitness function
```

```
def fitness_function(x, y):
```

```
    return x**2 + y**2
```

כל אלל בין 10- ל 10:

```
# Generate grid data for visualization
```

```
x = np.linspace(-10, 10, 400)
```

```
y = np.linspace(-10, 10, 400)
```

```
x, y = np.meshgrid(x, y)
```

```
z = fitness_function(x, y)
```

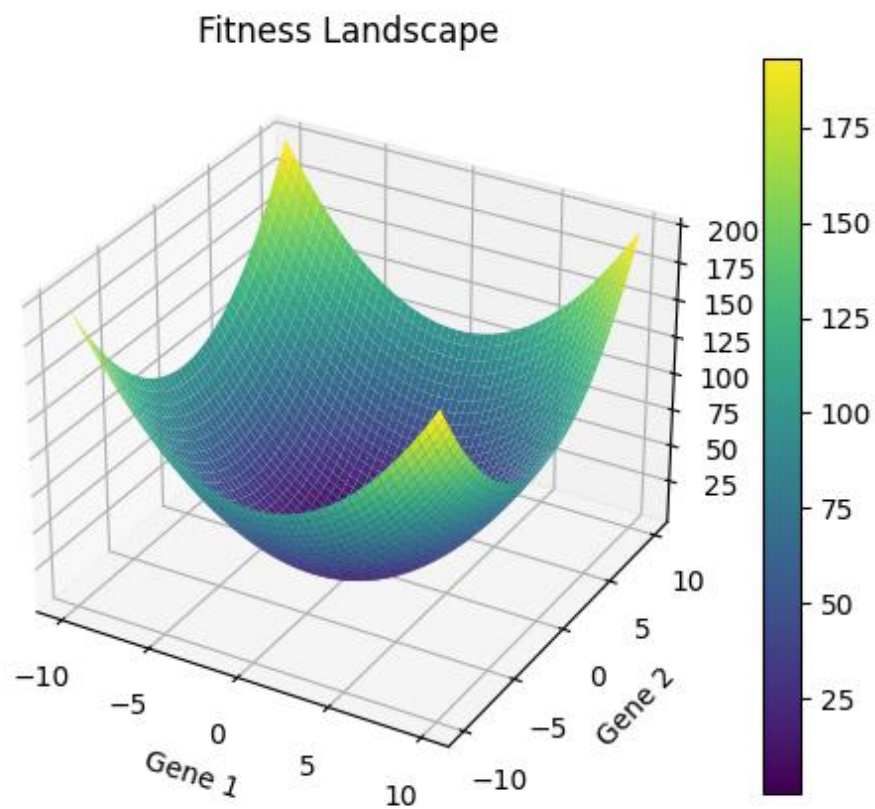
```
# Plot the fitness landscape
```

```
fig = plt.figure()
```

```
ax = fig.add_subplot(111, projection='3d')
```

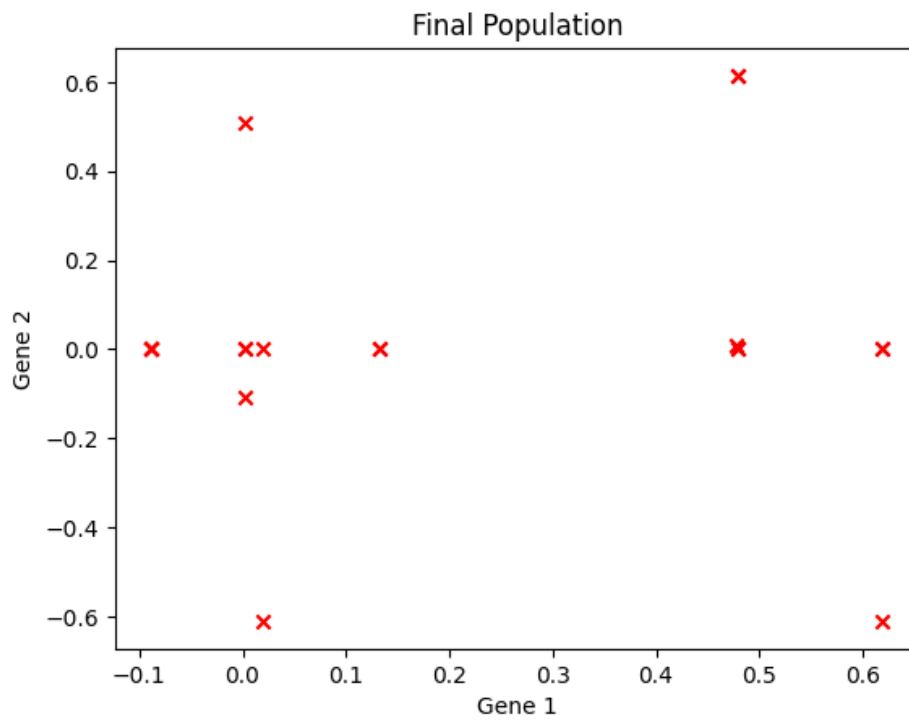
```
surface = ax.plot_surface(x, y, z, cmap='viridis')
```

```
fig.colorbar(surface)
ax.set_title('Fitness Landscape')
ax.set_xlabel('Gene 1')
ax.set_ylabel('Gene 2')
ax.set_zlabel('Fitness')
plt.show()
```



2. דיאגרמת פיזור ערכי האללים בדור האחרון

```
# Plot the final population
plt.scatter(population[:, 0], population[:, 1], c='red', marker='x')
plt.title('Final Population')
plt.xlabel('Gene 1')
plt.ylabel('Gene 2')
plt.show()
```



1. Average Fitness Over Generations

```
plt.figure()

plt.plot(range(num_generations), avg_fitness_over_time, label='Average Fitness')

plt.title('Average Fitness Over Generations')

plt.xlabel('Generation')

plt.ylabel('Average Fitness')

plt.legend()

plt.show()
```

2. Diversity Over Generations

```
plt.figure()

plt.plot(range(num_generations), diversity_over_time, label='Diversity')

plt.title('Diversity Over Generations')

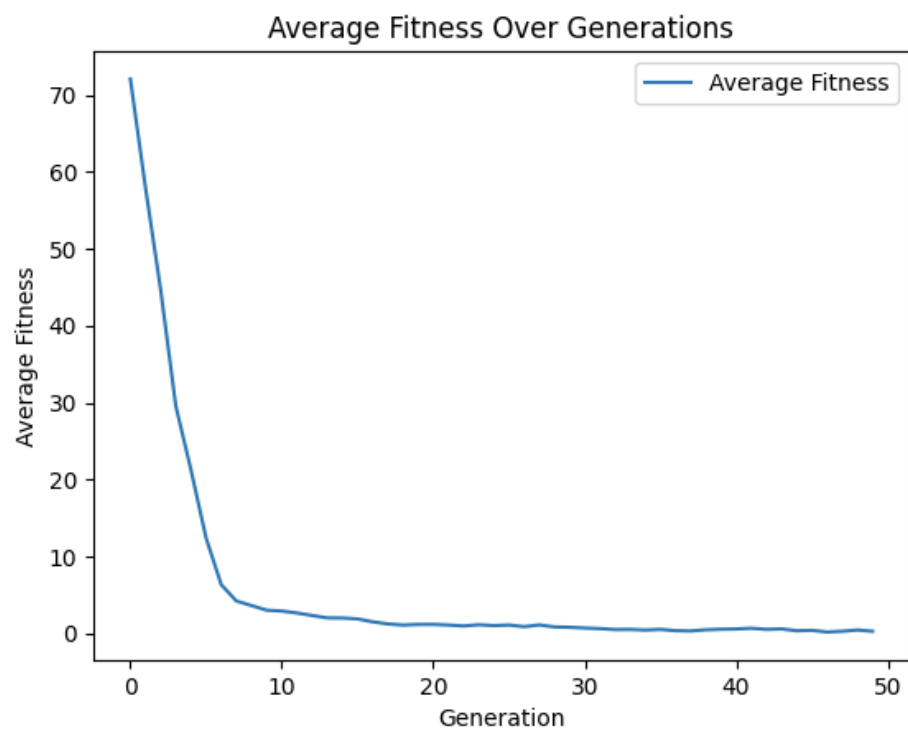
plt.xlabel('Generation')

plt.ylabel('Diversity (Standard Deviation)')

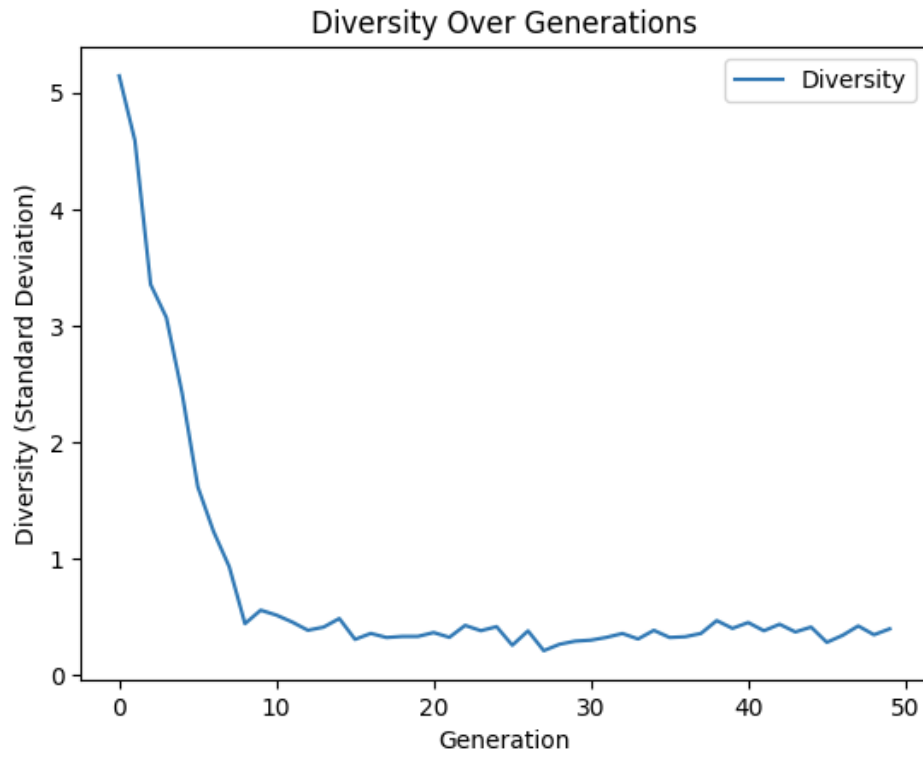
plt.legend()

plt.show()
```

3. גרף ממוצע הפיטנס לאורך הדורות



4. גרף הגיון הגנטי לאורך הדורות



5. מעקב גנאלוגי אחרי חמשת הדורות האחרונים

```
# Limit the number of generations to show
num_generations_to_show = 5
limited_genealogy = {k: v for k, v in genealogy.items() if k // population_size <
num_generations_to_show}

G = nx.DiGraph()
for key, values in limited_genealogy.items():
    for value in values:
        G.add_edge(value, key)

plt.figure(figsize=(12, 8))
pos = nx.spring_layout(G)
nx.draw(G, pos, with_labels=True, node_size=300, node_color='skyblue', font_size=6,
font_weight='bold', arrows=True)
plt.title('Genealogy Tracking (Limited to First Few Generations)')
plt.show()

# Interactive Plotting with Plotly
edge_x = []
edge_y = []
for edge in G.edges():
    x0, y0 = pos[edge[0]]
    x1, y1 = pos[edge[1]]
    edge_x.extend([x0, x1, None])
    edge_y.extend([y0, y1, None])

edge_trace = go.Scatter(
    x=edge_x, y=edge_y,
    line=dict(width=0.5, color='#888'),
    hoverinfo='none',
    mode='lines')

node_x = []
node_y = []
node_text = []
for node in G.nodes():
    x, y = pos[node]
    node_x.append(x)
    node_y.append(y)
    node_text.append(str(node))

node_trace = go.Scatter(
    x=node_x, y=node_y,
    mode='markers+text',
    text=node_text,
    textposition="bottom center",
```

```

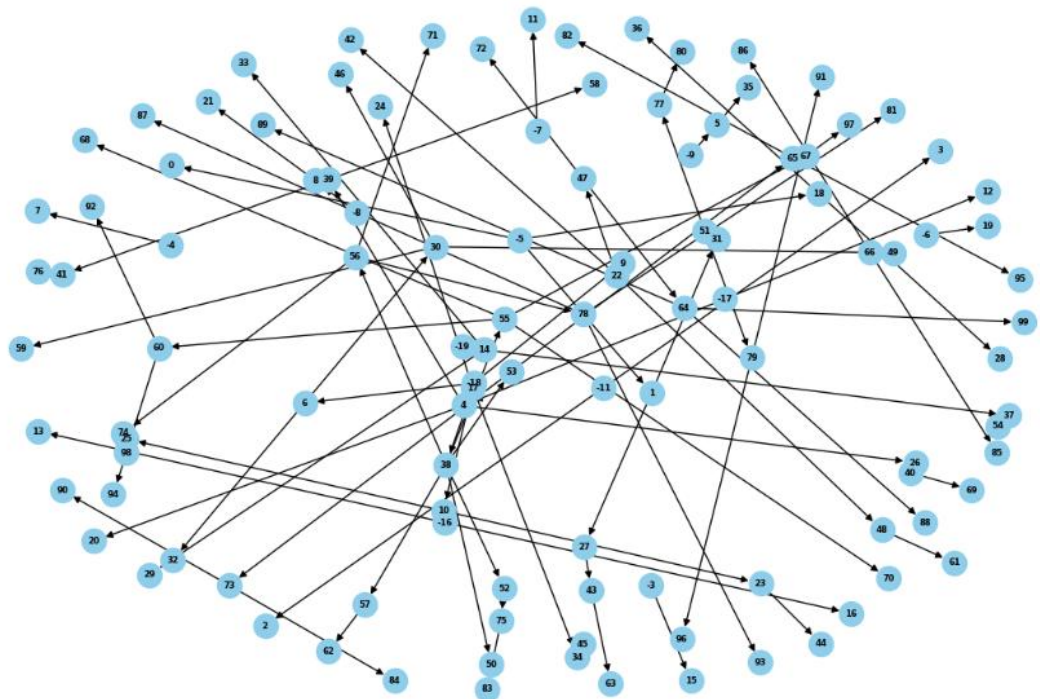
hoverinfo='text',
marker=dict(
    showscale=False,
    color='skyblue',
    size=10,
    line_width=2))

```

```

fig = go.Figure(data=[edge_trace, node_trace],
    layout=go.Layout(
        title='<br>Interactive Genealogy Tracking',
        titlefont_size=16,
        showlegend=False,
        hovermode='closest',
        margin=dict(b=20,l=5,r=5,t=40),
        annotations=[ dict(
            text="Interactive plot of the genealogy",
            showarrow=False,
            xref="paper", yref="paper" ) ],
        xaxis=dict(showgrid=False, zeroline=False),
        yaxis=dict(showgrid=False, zeroline=False))
    )
fig.show()

```



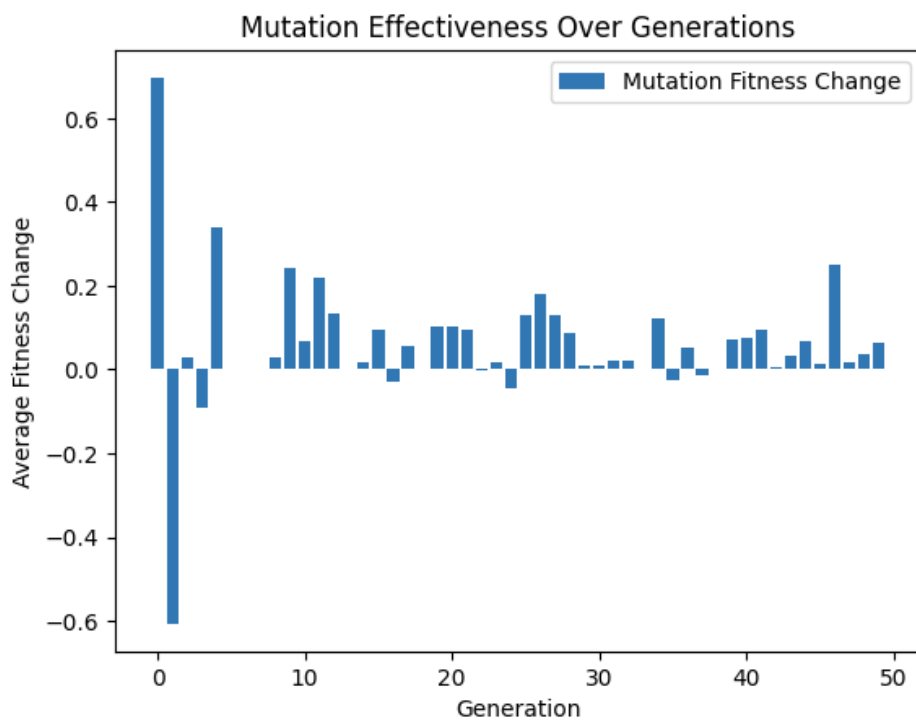
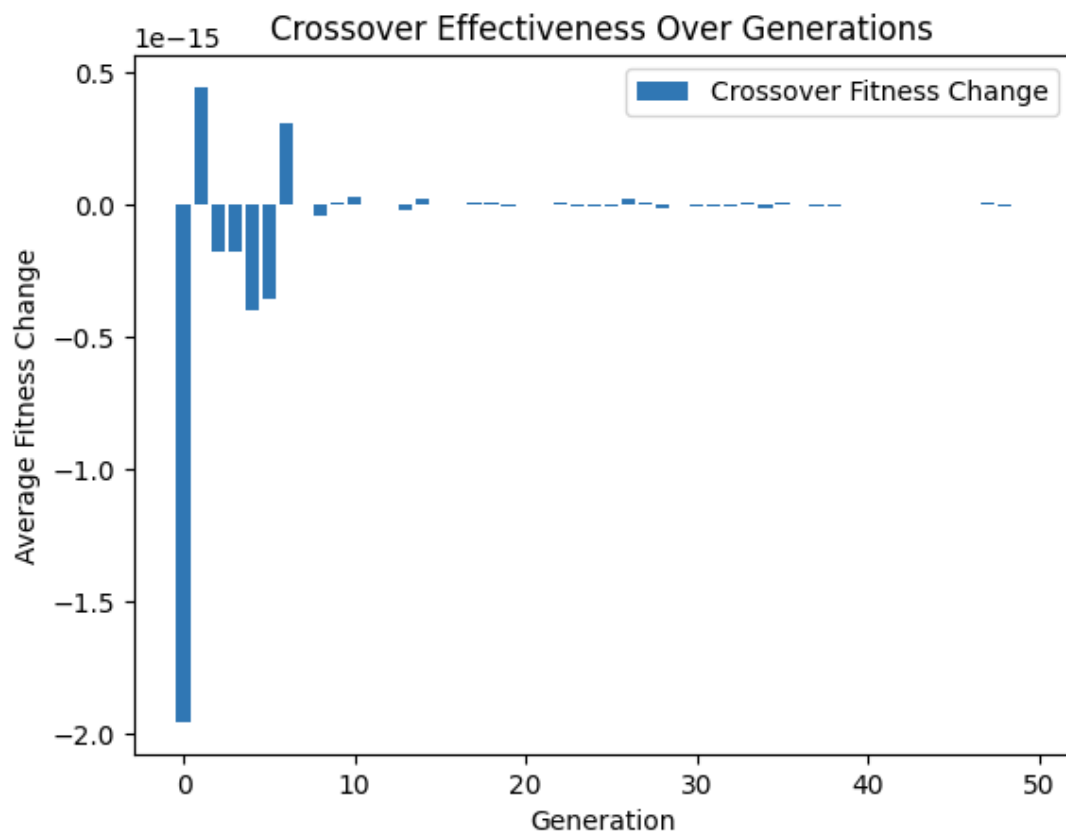
6. מעקב אחר השפעת האופרטורים הגנטיים לאורך הדורות:

```
crossover_fitness_changes = np.array(crossover_fitness_changes)
mutation_fitness_changes = np.array(mutation_fitness_changes)

avg_crossover_fitness_changes = np.mean(crossover_fitness_changes, axis=1)
avg_mutation_fitness_changes = np.mean(mutation_fitness_changes, axis=1)

# Bar chart for crossover effectiveness
plt.figure()
plt.bar(range(num_generations), avg_crossover_fitness_changes, label='Crossover Fitness Change')
plt.title('Crossover Effectiveness Over Generations')
plt.xlabel('Generation')
plt.ylabel('Average Fitness Change')
plt.legend()
plt.show()

# Bar chart for mutation effectiveness
plt.figure()
plt.bar(range(num_generations), avg_mutation_fitness_changes, label='Mutation Fitness Change')
plt.title('Mutation Effectiveness Over Generations')
plt.xlabel('Generation')
plt.ylabel('Average Fitness Change')
plt.legend()
plt.show()
```

7. בדיקת רגישות לפרמטרים שונים של ההרצה:

```
# Run GA with different parameter settings

parameter_sets = [

    {'population_size': 20, 'num_generations': 50, 'mutation_rate': 0.05, 'crossover_rate': 0.5},
    {'population_size': 20, 'num_generations': 50, 'mutation_rate': 0.1, 'crossover_rate': 0.5},
    {'population_size': 20, 'num_generations': 50, 'mutation_rate': 0.2, 'crossover_rate': 0.5},
    {'population_size': 20, 'num_generations': 50, 'mutation_rate': 0.1, 'crossover_rate': 0.3},
    {'population_size': 20, 'num_generations': 50, 'mutation_rate': 0.1, 'crossover_rate': 0.7},

]

results = []

for params in parameter_sets:

    avg_fitness_over_time, diversity_over_time, genealogy, crossover_fitness_changes,
    mutation_fitness_changes, final_population = run_ga(

        params['population_size'], params['num_generations'], params['mutation_rate'],
        params['crossover_rate'])

    final_avg_fitness = avg_fitness_over_time[-1]

    final_diversity = diversity_over_time[-1]

    results.append([params['mutation_rate'], params['crossover_rate'], final_avg_fitness,
    final_diversity, final_population])

# Create DataFrame for parallel coordinates plot

results_df = pd.DataFrame(results, columns=['Mutation Rate', 'Crossover Rate', 'Final Avg Fitness',
'Final Diversity', 'Final Population'])

# Parallel Coordinates Plot

plt.figure(figsize=(10, 6))

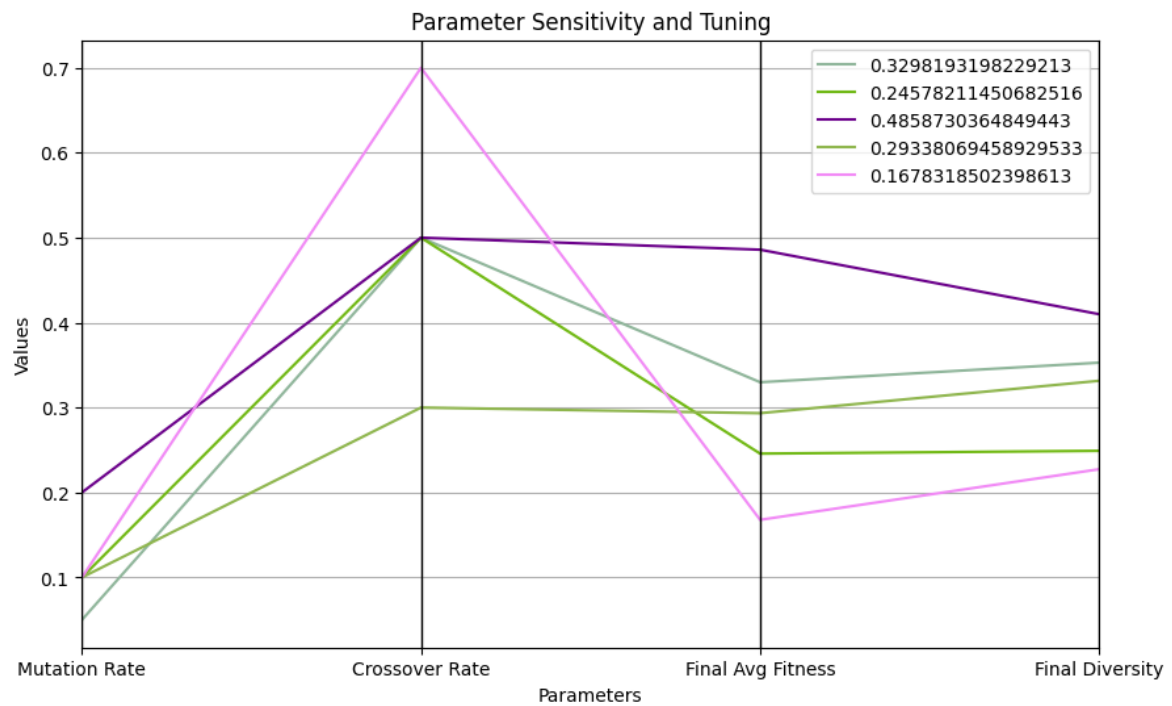
parallel_coordinates(results_df.drop(columns=['Final Population']), class_column='Final Avg Fitness',
cols=['Mutation Rate', 'Crossover Rate', 'Final Avg Fitness', 'Final Diversity'])

plt.title('Parameter Sensitivity and Tuning')

plt.xlabel('Parameters')

plt.ylabel('Values')
```

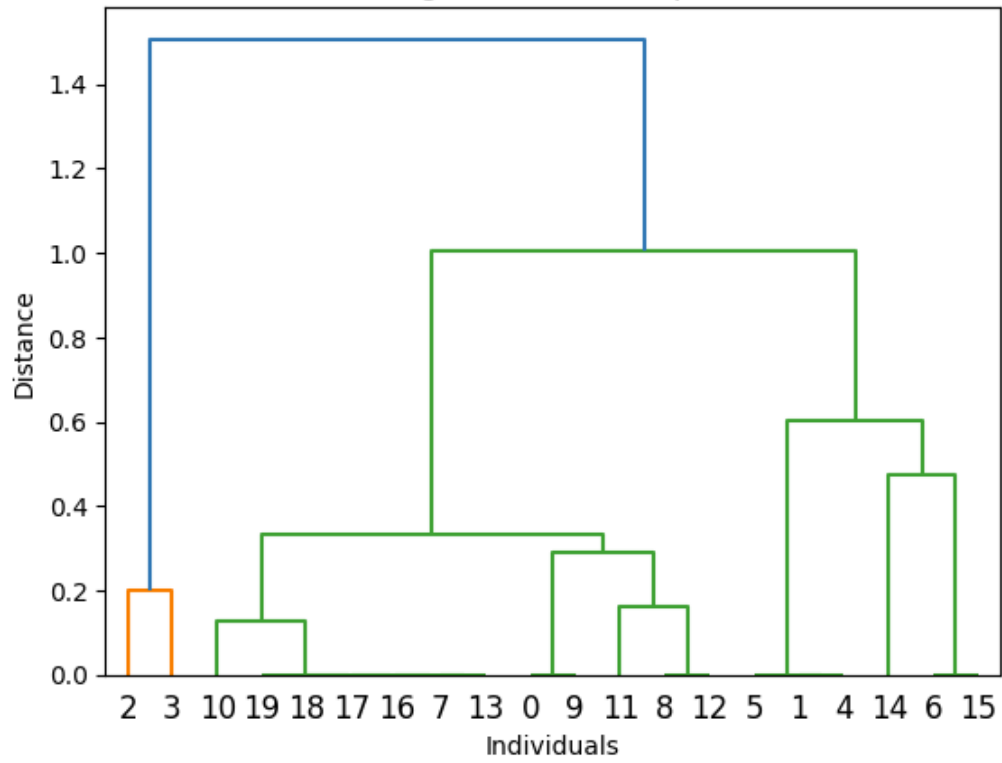
plt.show()



8. חלוקה היררכית של האוכלוסיה לצבירים לפי דמיון בין הפרטים באוכלוסייה הסופית

```
# Dendrogram for cluster analysis
Z = linkage(final_population, 'ward')
plt.figure()
dendrogram(Z)
plt.title('Dendrogram for Final Population')
plt.xlabel('Individuals')
plt.ylabel('Distance')
plt.show()
```

Dendrogram for Final Population



למי שממש ב C++ יש ספריה בשם Qt שבאמצעותה ניתן ליצר ויזואליזציה מקבילה

דוגמא לגרף של ערכים אקראיים לאורך סדרת זמן:

```
#include <QtCharts>

#include <QChartView>

#include <QLineSeries>

#include <QApplication>

int main(int argc, char *argv[]) {
    QApplication app(argc, argv);

    // Create the data series
    QtCharts::QLineSeries *series = new QtCharts::QLineSeries();

    for (int i = 1; i <= 50; ++i) {
        series->append(i, qrand() % 50); // Random data for illustration
    }

    // Create a chart
    QtCharts::QChart *chart = new QtCharts::QChart();

    chart->legend()->hide();
    chart->addSeries(series);
    chart->createDefaultAxes();
    chart->setTitle("Simple Fitness Chart Example");

    // Create a chart view
    QtCharts::QChartView *chartView = new QtCharts::QChartView(chart);
    chartView->setRenderHint(QPainter::Antialiasing);

    // Show the chart
    chartView->resize(640, 480);
    chartView->show();

    return app.exec();
}
```

