

MATHEMATIC FOR AI
PERHITUNGAN RUMUS LVQ (Learning Vector Quantization) DATASET BEASISWA

Dosen Pengampuh : Desi Anggraeni, S.kom.,MT



DISUSUN OLEH :

5A Mathematic For AI
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PROGRAM STUDI INFORMATIKA
FAKULTAS TEKNIK
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A. Dataset

	A	B	C	D	E	F	G	H
	ID	GPA	Income	Dependent	Achievement	Class		
1	1	3.85	2	4	9	1		
2	2	3.7	1.8	5	8	1		
3	3	3.9	2.5	3	10	1		
4	4	3.65	2.2	4	8	1		
5	5	3.8	1.5	6	9	1		
6	6	3.75	2.8	3	8	1		
7	7	3.95	2.1	2	10	1		
8	8	3.6	1.9	5	7	1		
9	9	3.88	2.6	4	9	1		
10	10	3.72	1.6	5	8	1		
11	11	3.66	2.4	4	7	1		
12	12	3.92	2	3	10	1		
13	13	3.78	2.7	4	8	1		
14	14	3.84	1.7	6	9	1		
15	15	3.69	2.3	5	7	1		
16	16	3.9	1.9	4	10	1		
17	17	3.74	2.5	3	8	1		
18	18	3.2	4	3	6	2		
19	19	3.1	4.5	2	5	2		
20	20	3.35	3.8	4	6	2		
21	21	3.05	5.2	3	4	2		
22	22	3.25	4.8	2	6	2		
23	23	3.3	3.5	5	5	2		
24	24	3.15	4.2	4	5	2		
25	25	3.4	4	3	6	2		
26	26	3.28	5.5	2	4	2		
27	27	3.12	4.7	3	5	2		
28	28	3.36	3.9	4	6	2		
29	29	3.08	5	3	4	2		
30	30	3.22	4.3	2	5	2		
31	31	3.18	4.9	3	5	2		
32	32	3.33	3.6	4	6	2		
33	33	3.27	4.1	3	5	2		
34	34	3.14	5.3	2	4	2		
35	35	2.4	8	1	2	3		
36	36	2.55	7.5	0	3	3		
37	37	2.3	9	1	1	3		
38	38	2.65	6.8	2	2	3		
39	39	2.5	8.5	0	2	3		
40	40	2.2	9.5	1	0	3		
41	41	2.75	7	2	3	3		
42	42	2.35	8.8	1	1	3		
43	43	2.6	6.5	0	2	3		
44	44	2.45	8.2	1	2	3		
45	45	2.7	7.8	2	3	3		
46	46	2.25	9.2	0	1	3		
47	47	2.58	6.9	1	2	3		
48	48	2.33	8.6	0	1	3		
49	49	2.68	7.2	2	2	3		
50	50	2.15	9.8	1	0	3		
51								
52								

B. Perbaikan tambahan program

```
LVQ.py > ...
1 import numpy as np
2 import pandas as pd
3
4 # 1. LOAD DATASET
5 file_path = "Dataset LVQ_50_Data.xlsx"
6 df = pd.read_excel(file_path)
7
8 x = df[['GPA', 'Income', 'Dependents', 'Achievements']].values
9 y = df['Class'].values
10
11
12 # 2. NORMALISASI MIN-MAX
13 def min_max_normalization(X):
14     return (X - X.min(axis=0)) / (X.max(axis=0) - X.min(axis=0))
15
16 X_norm = min_max_normalization(X)
17
18
19 # 3. INISIALISASI BOBOT
20 idx_c1 = np.where(y == 1)[0][0]
21 idx_c2 = np.where(y == 2)[0][0]
22 idx_c3 = np.where(y == 3)[0][0]
23
24 W = np.array([X_norm[idx_c1], X_norm[idx_c2], X_norm[idx_c3]])
25 W_class = np.array([1, 2, 3])
26 W_initial = W.copy()
27
28
29 # 4. PARAMETER LVQ
30 alpha = 0.3
31 decay = 0.9
32 epochs = 5
33
34
35 # 5. JARAK EUCLIDEAN
36 def euclidean_distance(x, w):
37     return np.sqrt(np.sum((x - w) ** 2))
```

```
LVQ.py > ...
38
39 # 6. TRAINING + SIMPAN PERHITUNGAN
40 epoch_weights = []
41 epoch_calculations = []
42
43
44 for epoch in range(epochs):
45     print(f"\n{'='*65}")
46     print(f"EPOCH {epoch+1} | Learning Rate ( $\alpha$ ) = {alpha:.4f}")
47     print(f"{'='*65}")
48
49     W_epoch_start = W.copy()
50     epoch_rows = []
51
52     for i in range(len(X_norm)):
53         x = X_norm[i]
54         target = y[i]
55
56         d1 = euclidean_distance(x, W_epoch_start[0])
57         d2 = euclidean_distance(x, W_epoch_start[1])
58         d3 = euclidean_distance(x, W_epoch_start[2])
59
60         distances = [d1, d2, d3]
61         winner = np.argmin(distances) + 1
62
63         # ===== OUTPUT RUMUS LVQ DI TERMINAL =====
64         if i < 3:
65             print(f"\nData ke-{i+1}")
66             print(f"d_w1 = sqrt(sum((X - W1)^2)) = {d1:.6f}")
67             print(f"d_w2 = sqrt(sum((X - W2)^2)) = {d2:.6f}")
68             print(f"d_w3 = sqrt(sum((X - W3)^2)) = {d3:.6f}")
69             print(f"winner = argmin(d_w1, d_w2, d_w3) = {winner}")
70
71         # simpan ke excel
72         epoch_rows.append([
73             df.loc[i, 'ID'],
74             x[0], x[1], x[2], x[3],
75             d1, d2, d3,
76             winner
77         ])
78
79     # Simpan ke excel
80     df_epoch = pd.DataFrame(epoch_rows, columns=['ID', 'GPA', 'Income', 'Dependents', 'Achievements', 'd1', 'd2', 'd3', 'winner'])
81     df_epoch.to_excel(f"epoch_{epoch+1}.xlsx", index=False)
```

```

LVQ.py > ...
# 6. TRAINING + SIMPAN PERHITUNGAN
epoch_weights = []
epoch_calculations = []

for epoch in range(epochs):
    print(f"\n{'='*65}")
    print(f"EPOCH {epoch+1} | Learning Rate (α) = {alpha:.4f}")
    print(f"{'='*65}")

    W_epoch_start = W.copy()
    epoch_rows = []

    for i in range(len(X_norm)):
        x = X_norm[i]
        target = y[i]

        d1 = euclidean_distance(x, W_epoch_start[0])
        d2 = euclidean_distance(x, W_epoch_start[1])
        d3 = euclidean_distance(x, W_epoch_start[2])

        distances = [d1, d2, d3]
        winner = np.argmin(distances) + 1

        # ===== OUTPUT RUMUS LVQ DI TERMINAL =====
        if i < 3:
            print(f"\nData ke-{i+1}")
            print(f"d_w1 = sqrt(sum((X - w1)^2)) = {d1:.6f}")
            print(f"d_w2 = sqrt(sum((X - w2)^2)) = {d2:.6f}")
            print(f"d_w3 = sqrt(sum((X - w3)^2)) = {d3:.6f}")
            print(f"winner = argmin(d_w1, d_w2, d_w3) = {winner}")

        # simpan ke excel
        epoch_rows.append([
            df.loc[i, 'ID'],
            x[0], x[1], x[2], x[3],
            d1, d2, d3,
            winner
        ])

```

```

LVQ.py > ...
# ===== UPDATE BOBOT LVQ1 =====
if W_class[winner-1] == target:
    W[winner-1] = W[winner-1] + alpha * (x - W[winner-1])
else:
    W[winner-1] = W[winner-1] - alpha * (x - W[winner-1])

epoch_calculations.append(epoch_rows)
epoch_weights.append(W.copy())
alpha *= decay

# 7. PREDIKSI & AKURASI
def lvq_predict(X, W, W_class):
    return np.array([
        W_class[np.argmin([euclidean_distance(x, w) for w in W])]
        for x in X
    ])

y_pred = lvq_predict(X_norm, W, W_class)
accuracy = np.mean(y_pred == y) * 100

# 8. SIMPAN KE EXCEL (SEMUA EPOCH)
writer = pd.ExcelWriter("LVQ_Training_Result_50Data_5Epoch.xlsx", engine="openpyxl")

# Parameter LVQ
df_param = pd.DataFrame(W_initial, columns=['X1', 'X2', 'X3', 'X4'])
df_param['Class'] = W_class
df_param.to_excel(writer, sheet_name="Parameter LVQ", index=False)

# Bobot per epoch
for i, w_epoch in enumerate(epoch_weights):
    df_epoch = pd.DataFrame(w_epoch, columns=['X1', 'X2', 'X3', 'X4'])
    df_epoch['Class'] = W_class
    df_epoch.to_excel(writer, sheet_name=f"Epoch {i+1}", index=False)

```

```

LVQ.py > ...
113 df_epoch.to_excel(writer, sheet_name=f"Epoch_{i+1}", index=False)
114
115 # ===== PERHITUNGAN EPOCH 1-5 =====
116 for ep in range(5):
117     df_calc = pd.DataFrame(
118         epoch_calculations[ep],
119         columns=['ID', 'X1', 'X2', 'X3', 'X4', 'd_W1', 'd_W2', 'd_W3', 'Winner']
120     )
121     df_calc.to_excel(writer, sheet_name=f"Perhitungan_LVQ_Epoch{ep+1}", index=False)
122
123 # Prediksi
124 df_pred = df.copy()
125 df_pred['Predicted_Class'] = y_pred
126 df_pred.to_excel(writer, sheet_name="Prediksi", index=False)
127
128 # Akurasi
129 pd.DataFrame({
130     "Epoch": [epochs],
131     "Akurasi (%)": [accuracy]
132 }).to_excel(writer, sheet_name="Akurasi", index=False)
133
134 writer.close()
135
136 print("\nFile 'LVQ_Training_Result_50Data_5Epoch.xlsx' BERHASIL dibuat lengkap.")
137 print(f"Akurasi Training LVQ = {accuracy:.2f}%")

```

C. Perhitungan Python output

```

PS D:\TUGAS BESAR NANANG> & C:/Users/LENOVO/AppData/Local/Programs/Python/Python38-64/Python.exe LVQ.py
=====
EPOCH 1 | Learning Rate (α) = 0.3000
=====

Data ke-1
d_W1 = sqrt(sum((X - W1)^2)) = 0.000000
d_W2 = sqrt(sum((X - W2)^2)) = 0.553392
d_W3 = sqrt(sum((X - W3)^2)) = 1.382567
Winner = argmin(d_W1, d_W2, d_W3) = 1

Data ke-2
d_W1 = sqrt(sum((X - W1)^2)) = 0.212845
d_W2 = sqrt(sum((X - W2)^2)) = 0.546378
d_W3 = sqrt(sum((X - W3)^2)) = 1.372604
Winner = argmin(d_W1, d_W2, d_W3) = 1

Data ke-3
d_W1 = sqrt(sum((X - W1)^2)) = 0.205374
d_W2 = sqrt(sum((X - W2)^2)) = 0.586426
d_W3 = sqrt(sum((X - W3)^2)) = 1.372830
Winner = argmin(d_W1, d_W2, d_W3) = 1

=====
EPOCH 2 | Learning Rate (α) = 0.2700
=====

Data ke-1
d_W1 = sqrt(sum((X - W1)^2)) = 0.067741
d_W2 = sqrt(sum((X - W2)^2)) = 0.657388
d_W3 = sqrt(sum((X - W3)^2)) = 1.446432
Winner = argmin(d_W1, d_W2, d_W3) = 1

Data ke-2
d_W1 = sqrt(sum((X - W1)^2)) = 0.183244
d_W2 = sqrt(sum((X - W2)^2)) = 0.645281
d_W3 = sqrt(sum((X - W3)^2)) = 1.431940
Winner = argmin(d_W1, d_W2, d_W3) = 1

Data ke-3
d_W1 = sqrt(sum((X - W1)^2)) = 0.239665
d_W2 = sqrt(sum((X - W2)^2)) = 0.688326
d_W3 = sqrt(sum((X - W3)^2)) = 1.440985
Winner = argmin(d_W1, d_W2, d_W3) = 1

```

```

=====
EPOCH 3 | Learning Rate ( $\alpha$ ) = 0.2430
=====

Data ke-1
d_W1 = sqrt(sum((X - W1)^2)) = 0.068166
d_W2 = sqrt(sum((X - W2)^2)) = 0.652945
d_W3 = sqrt(sum((X - W3)^2)) = 1.439082
Winner = argmin(d_W1, d_W2, d_W3) = 1

Data ke-2
d_W1 = sqrt(sum((X - W1)^2)) = 0.178576
d_W2 = sqrt(sum((X - W2)^2)) = 0.640033
d_W3 = sqrt(sum((X - W3)^2)) = 1.425282
Winner = argmin(d_W1, d_W2, d_W3) = 1

Data ke-3
d_W1 = sqrt(sum((X - W1)^2)) = 0.243505
d_W2 = sqrt(sum((X - W2)^2)) = 0.685079
d_W3 = sqrt(sum((X - W3)^2)) = 1.433150
Winner = argmin(d_W1, d_W2, d_W3) = 1

=====
EPOCH 4 | Learning Rate ( $\alpha$ ) = 0.2187
=====

Data ke-1
d_W1 = sqrt(sum((X - W1)^2)) = 0.068632
d_W2 = sqrt(sum((X - W2)^2)) = 0.648934
d_W3 = sqrt(sum((X - W3)^2)) = 1.432397
Winner = argmin(d_W1, d_W2, d_W3) = 1

Data ke-2
d_W1 = sqrt(sum((X - W1)^2)) = 0.175010
d_W2 = sqrt(sum((X - W2)^2)) = 0.635219
d_W3 = sqrt(sum((X - W3)^2)) = 1.419227
Winner = argmin(d_W1, d_W2, d_W3) = 1

Data ke-3
d_W1 = sqrt(sum((X - W1)^2)) = 0.246435
d_W2 = sqrt(sum((X - W2)^2)) = 0.682072
d_W3 = sqrt(sum((X - W3)^2)) = 1.426017
Winner = argmin(d_W1, d_W2, d_W3) = 1

```

```

PS D:\TUGAS BESAR NANANG> & C:/Users/LENOVO/AppData/Local/Programs/Python/Python38-32/Python.exe
EPOCH 5 | Learning Rate ( $\alpha$ ) = 0.1968

Data ke-2
d_W1 = sqrt(sum((X - W1)^2)) = 0.175010
d_W2 = sqrt(sum((X - W2)^2)) = 0.635219
d_W3 = sqrt(sum((X - W3)^2)) = 1.419227
Winner = argmin(d_W1, d_W2, d_W3) = 1

Data ke-3
d_W1 = sqrt(sum((X - W1)^2)) = 0.246435
d_W2 = sqrt(sum((X - W2)^2)) = 0.682072
d_W3 = sqrt(sum((X - W3)^2)) = 1.426017
Winner = argmin(d_W1, d_W2, d_W3) = 1

=====
EPOCH 5 | Learning Rate ( $\alpha$ ) = 0.1968
Winner = argmin(d_W1, d_W2, d_W3) = 1

Data ke-3
d_W1 = sqrt(sum((X - W1)^2)) = 0.246435
d_W2 = sqrt(sum((X - W2)^2)) = 0.682072
d_W3 = sqrt(sum((X - W3)^2)) = 1.426017
Winner = argmin(d_W1, d_W2, d_W3) = 1
o

=====
EPOCH 5 | Learning Rate ( $\alpha$ ) = 0.1968
d_W1 = sqrt(sum((X - W1)^2)) = 0.246435
d_W2 = sqrt(sum((X - W2)^2)) = 0.682072
d_W3 = sqrt(sum((X - W3)^2)) = 1.426017
Winner = argmin(d_W1, d_W2, d_W3) = 1

=====
EPOCH 5 | Learning Rate ( $\alpha$ ) = 0.1968
d_W3 = sqrt(sum((X - W3)^2)) = 1.426017
Winner = argmin(d_W1, d_W2, d_W3) = 1

=====
EPOCH 5 | Learning Rate ( $\alpha$ ) = 0.1968

=====
EPOCH 5 | Learning Rate ( $\alpha$ ) = 0.1968
EPOCH 5 | Learning Rate ( $\alpha$ ) = 0.1968

```

```
Data ke-1
d_W1 = sqrt(sum((X - W1)^2)) = 0.069101
d_W2 = sqrt(sum((X - W2)^2)) = 0.645435
d_W3 = sqrt(sum((X - W3)^2)) = 1.426541
Winner = argmin(d_W1, d_W2, d_W3) = 1
```

```
Data ke-2
d_W1 = sqrt(sum((X - W1)^2)) = 0.172321
d_W2 = sqrt(sum((X - W2)^2)) = 0.630991
d_W3 = sqrt(sum((X - W3)^2)) = 1.413926
Winner = argmin(d_W1, d_W2, d_W3) = 1
```

```
Data ke-1
d_W1 = sqrt(sum((X - W1)^2)) = 0.069101
d_W2 = sqrt(sum((X - W2)^2)) = 0.645435
d_W3 = sqrt(sum((X - W3)^2)) = 1.426541
Winner = argmin(d_W1, d_W2, d_W3) = 1
```

```
Data ke-2
d_W1 = sqrt(sum((X - W1)^2)) = 0.172321
d_W2 = sqrt(sum((X - W2)^2)) = 0.630991
d_W3 = sqrt(sum((X - W3)^2)) = 1.413926
Winner = argmin(d_W1, d_W2, d_W3) = 1
d_W2 = sqrt(sum((X - W2)^2)) = 0.645435
d_W3 = sqrt(sum((X - W3)^2)) = 1.426541
Winner = argmin(d_W1, d_W2, d_W3) = 1
```

```
Data ke-2
d_W1 = sqrt(sum((X - W1)^2)) = 0.172321
d_W2 = sqrt(sum((X - W2)^2)) = 0.630991
d_W3 = sqrt(sum((X - W3)^2)) = 1.413926
Winner = argmin(d_W1, d_W2, d_W3) = 1
```

```
Data ke-2
d_W1 = sqrt(sum((X - W1)^2)) = 0.172321
d_W2 = sqrt(sum((X - W2)^2)) = 0.630991
d_W3 = sqrt(sum((X - W3)^2)) = 1.413926
Winner = argmin(d_W1, d_W2, d_W3) = 1
d_W1 = sqrt(sum((X - W1)^2)) = 0.172321
d_W2 = sqrt(sum((X - W2)^2)) = 0.630991
d_W3 = sqrt(sum((X - W3)^2)) = 1.413926
Winner = argmin(d_W1, d_W2, d_W3) = 1
d_W3 = sqrt(sum((X - W3)^2)) = 1.413926
Winner = argmin(d_W1, d_W2, d_W3) = 1
Winner = argmin(d_W1, d_W2, d_W3) = 1
```

```
Data ke-3
d_W1 = sqrt(sum((X - W1)^2)) = 0.248649
d_W2 = sqrt(sum((X - W2)^2)) = 0.679471
d_W3 = sqrt(sum((X - W3)^2)) = 1.419763
Winner = argmin(d_W1, d_W2, d_W3) = 1
```

File 'LVQ Training Result 50Data 5Epoch.xlsx' BERHASIL dibuat lengkap.

- **Iterasi Unit**

- **Iterasi 1**

- **Iterasi 2**

- **Iterasi 3**

Perhitungan Manual LVQ - Epoch 3 (50 data)

Alpha | 0.24300

ID	X1	X2	X3	X4	Target	W1_x1	W1_x2	W1_x3	W1_x4	W2_x1	W2_x2	W2_x3	W2_x4	W3_x1	W3_x2	W3_x3
1	0.944444	0.060241	0.666667	0.900000	1	0.9101584	0.08487	0.6756027	0.8472309	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
2	0.861111	0.036145	0.833333	0.800000	1	0.9184899	0.0788851	0.6734312	0.8600538	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
3	0.972222	0.120482	0.500000	1.000000	1	0.9045469	0.0684992	0.7122874	0.8454607	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
4	0.833333	0.084337	0.666667	0.800000	1	0.9209922	0.081131	0.6607016	0.8830138	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
5	0.916667	0.000000	1.000000	0.900000	1	0.8996909	0.0819101	0.6621511	0.8628414	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
6	0.888889	0.156627	0.500000	0.800000	1	0.903816	0.062006	0.7442484	0.871871	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
7	1.000000	0.072289	0.333333	1.000000	1	0.9001887	0.0849988	0.684896	0.8544063	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
8	0.805556	0.048193	0.833333	0.700000	1	0.9244429	0.0819103	0.5994663	0.8897856	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
9	0.961111	0.132530	0.666667	0.900000	1	0.8955533	0.073717	0.6562696	0.8436677	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
10	0.872222	0.012048	0.833333	0.800000	1	0.9114838	0.0880086	0.6588161	0.8573564	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
11	0.838889	0.108434	0.666667	0.700000	1	0.9019432	0.0695502	0.7012238	0.8434188	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
12	0.983333	0.060241	0.500000	1.000000	1	0.886621	0.0789989	0.6928264	0.808568	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
13	0.905556	0.144578	0.666667	0.800000	1	0.9101221	0.0744407	0.6459696	0.855086	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
14	0.938889	0.024096	1.000000	0.900000	1	0.9090124	0.0914842	0.650999	0.8417001	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
15	0.855556	0.096386	0.833333	0.700000	1	0.9162724	0.0751089	0.7358062	0.855867	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
16	0.972222	0.048193	0.666667	1.000000	1	0.9015182	0.0802791	0.7595053	0.8179913	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604
17	0.883333	0.120482	0.500000	0.800000	1	0.9186993	0.0724822	0.7369455	0.8622194	0.5935124	0.3672026	0.4731069	0.4858821	0.1469183	0.8270627	0.1688604

• Iterasi 4

Perhitungan Manual LVQ - Epoch 4 (50 data)

Alpha (0.21870

ID	X1	X2	X3	X4	Target	W1_x1	W1_x2	W1_x3	W1_x4	W2_x1	W2_x2	W2_x3	W2_x4	W3_x1	W3_x2	W3_x3	W3_x4	d_W1	d_W2	d_W3	Winner
1	0.944444	0.060241	0.666667	0.900000	1	0.91010537	0.0841461	0.67936776	0.8471001	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.068632	0.648934	1.432397	1
2	0.861111	0.036145	0.833333	0.800000	1	0.91761532	0.07891805	0.67659003	0.85866931	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.181749	0.635219	1.419227	1
3	0.972222	0.120482	0.500000	1.000000	1	0.90525785	0.06956349	0.71086979	0.84583833	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.274424	0.682072	1.426017	1
4	0.833333	0.084337	0.666667	0.800000	1	0.91990296	0.08069935	0.66475257	0.87955349	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.117643	0.518822	1.306819	1
5	0.916667	0.000000	1.000000	0.900000	1	0.90097018	0.08149498	0.66517118	0.86215514	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.347031	0.824672	1.596250	1
6	0.888889	0.156627	0.500000	0.800000	1	0.904403	0.06367203	0.73839824	0.87043181	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.265849	0.477606	1.243579	1
7	1.000000	0.072289	0.333333	1.000000	1	0.90101007	0.08400117	0.68626055	0.85502837	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.394348	0.730215	1.437616	1
8	0.805556	0.048193	0.833333	0.700000	1	0.92265917	0.08143976	0.60907537	0.88673367	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.316196	0.562867	1.340055	1
9	0.961111	0.132530	0.666667	0.900000	1	0.89704861	0.07416864	0.65812058	0.84589502	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.102520	0.627809	1.404822	1
10	0.872222	0.012048	0.833333	0.800000	1	0.91105908	0.0869323	0.65998961	0.85772778	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.201237	0.652698	1.438257	1
11	0.838889	0.108434	0.666667	0.700000	1	0.90255646	0.07055514	0.69789888	0.84510271	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.165891	0.454496	1.247628	1
12	0.983333	0.060241	0.500000	1.000000	1	0.88863939	0.07883919	0.69106918	0.81336875	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.283992	0.711974	1.462908	1
13	0.905556	0.144578	0.666667	0.800000	1	0.90934896	0.07477176	0.64928235	0.854185	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.090142	0.527893	1.313879	1
14	0.938889	0.024096	1.000000	0.900000	1	0.90851934	0.09003845	0.6530843	0.84233474	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.359091	0.823328	1.594832	1
15	0.855556	0.096386	0.833333	0.700000	1	0.91516116	0.07561692	0.72895476	0.85494613	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.197199	0.558731	1.338438	1
16	0.972222	0.048193	0.666667	1.000000	1	0.90212541	0.08015902	0.75178236	0.82105941	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.212602	0.735643	1.590621	1
17	0.883333	0.120482	0.500000	0.800000	1	0.91745559	0.073168	0.73316756	0.86019372	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.247777	0.491200	1.260061	1
18	0.583333	0.301205	0.500000	0.600000	2	0.90999305	0.08351556	0.68217381	0.84702935	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.498303	0.131269	0.888573	2
19	0.527778	0.361446	0.333333	0.500000	2	0.90999305	0.08351556	0.68217381	0.84702935	0.59184763	0.35151868	0.48240783	0.51279204	0.15084375	0.82241329	0.16716493	0.13087692	0.682241	0.163065	0.720024	2
20	0.666667	0.277108	0.666667	0.600000	2	0.90999305	0.08351556	0.68217381	0.84702935	0.57935554	0.50999442	0.15084375	0.82241329	0.16716493	0.13087692	0.13087692	0.13087692	0.397429	0.262461	0.101671	2
21	0.500000	0.445783	0.500000	0.400000	2	0.90999305	0.08351556	0.68217381	0.84702935	0.59726292	0.33694141	0.49723281	0.52967864	0.15084375	0.82241329	0.16716493	0.13087692	0.729626	0.195271	0.668556	2
22	0.611111	0.397590	0.333333	0.600000	2	0.90999305	0.08351556	0.68217381	0.84702935	0.36704509	0.50131792	0.15084375	0.82241329	0.16716493	0.13087692	0.13087692	0.13087692	0.608840	0.198471	0.800006	2
23	0.638889	0.240964	0.833333	0.500000	2	0.90999305	0.08351556	0.68217381	0.84702935	0.58367217	0.36880315	0.46186084	0.52289969	0.15084375	0.82241329	0.16716493	0.13087692	0.491494	0.393736	1.075316	2
24	0.555556	0.325301	0.666667	0.500000	2	0.90999305	0.08351556	0.68217381	0.84702935	0.59574807	0.3408447	0.54310187	0.51789153	0.15084375	0.82241329	0.16716493	0.13087692	0.552047	0.132081	0.892561	2
25	0.694444	0.301205	0.500000	0.600000	2	0.90999305	0.08351556	0.68217381	0.84702935	0.58095797	0.33744534	0.57012549	0.51397865	0.15084375	0.82241329	0.16716493	0.13087692	0.433660	0.158695	0.947637	2
26	0.627778	0.481928	0.333333	0.400000	2	0.90999305	0.08351556	0.68217381	0.84702935	0.61046526	0.32951954	0.55478905	0.53279152	0.15084375	0.82241329	0.16716493	0.13087692	0.748266	0.300340	0.665910	2
27	0.538889	0.385542	0.500000	0.500000	2	0.90999305	0.08351556	0.68217381	0.84702935	0.61425151	0.36288512	0.50635868	0.50375902	0.15084375	0.82241329	0.16716493	0.13087692	0.618510	0.079050	0.767116	2

• Iterasi 5

Perhitungan Manual LVQ - Epoch 5 (50 data)

- Hasil iterasi

Hasil Akhir (Setelah Iterasi / Epoch 5)									
No	ID	X1	X2	X3	X4	U_C1	U_C2	U_C3	Clester
1	1	0.3444	0.0602	0.6667	0.9	0.0631	0.6454	1.4265	1
2	2	0.8611	0.0361	0.8333	0.8	0.1783	0.631	1.4139	1
3	3	0.3722	0.1205	0.5	1	0.2729	0.6795	1.4198	1
4	4	0.8333	0.0843	0.6667	0.8	0.1148	0.5151	1.3011	1
5	5	0.3167	0	1	0.9	0.3443	0.8203	1.5912	1
6	6	0.8889	0.1566	0.5	0.8	0.2604	0.475	1.2375	1
7	7	1	0.0723	0.3333	1	0.395	0.7286	1.431	1
8	8	0.8056	0.0482	0.8333	0.7	0.3078	0.5586	1.3349	1
9	9	0.3611	0.1325	0.6667	0.9	0.1004	0.6243	1.339	1
10	10	0.8722	0.012	0.8333	0.8	0.1997	0.6485	1.4329	1
11	11	0.8389	0.1084	0.6667	0.7	0.1667	0.4508	1.242	1
12	12	0.3833	0.0602	0.5	1	0.2795	0.7094	1.4566	1
13	13	0.3056	0.1446	0.6667	0.8	0.089	0.5243	1.3082	1
14	14	0.3389	0.0241	1	0.9	0.3568	0.8189	1.5898	1
15	15	0.8556	0.0964	0.8333	0.7	0.1995	0.5545	1.3333	1
16	16	0.3722	0.0482	0.6667	1	0.2073	0.7322	1.5037	1
17	17	0.8833	0.1205	0.5	0.8	0.2435	0.4886	1.2539	1
18	18	0.5833	0.3012	0.5	0.6	0.4992	0.1281	0.8828	2
19	19	0.5278	0.3614	0.3333	0.5	0.6834	0.1657	0.7138	2
20	20	0.6667	0.2771	0.6667	0.6	0.3977	0.2575	1.0112	2
21	21	0.5	0.4458	0.5	0.4	0.7303	0.1927	0.6635	2
22	22	0.6111	0.3376	0.3333	0.6	0.6102	0.1915	0.7937	2
23	23	0.6389	0.241	0.8333	0.5	0.491	0.3936	1.071	2
24	24	0.5556	0.3253	0.6667	0.5	0.5523	0.1372	0.8878	2
25	25	0.6944	0.3012	0.5	0.6	0.4347	0.1558	0.3418	2
26	26	0.6278	0.4819	0.3333	0.4	0.7495	0.2955	0.66	2
27	27	0.5389	0.3855	0.5	0.5	0.6193	0.0778	0.7616	2
28	28	0.6722	0.2892	0.6667	0.6	0.4004	0.2153	1.0076	2
29	29	0.5167	0.4217	0.5	0.4	0.7093	0.1754	0.686	2
30	30	0.5944	0.3373	0.3333	0.5	0.6385	0.199	0.7657	2
31	31	0.5722	0.4096	0.5	0.5	0.6124	0.0553	0.7659	2
32	32	0.6556	0.253	0.6667	0.6	0.3936	0.241	1.0188	2
33	33	0.6222	0.3133	0.5	0.5	0.5386	0.051	0.8479	2
34	34	0.55	0.4578	0.3333	0.4	0.77	0.2569	0.6203	2
35	35	0.1389	0.7831	0.1667	0.2	1.3308	0.7561	0.0753	3
36	36	0.2222	0.7229	0	0.3	1.2844	0.7354	0.252	3
37	37	0.0833	0.9036	0.1667	0.1	1.4773	0.8998	0.1609	3
38	38	0.2778	0.6386	0.3333	0.2	1.118	0.5341	0.2945	3
39	39	0.1944	0.8434	0	0.2	1.406	0.8438	0.192	3
40	40	0.0278	0.9633	0.1667	0	1.5935	1.0134	0.2887	3
41	41	0.3333	0.6627	0.3333	0.3	1.0444	0.4674	0.3474	3
42	42	0.1111	0.8795	0.1667	0.1	1.4486	0.8638	0.1352	3
43	43	0.25	0.6024	0	0.2	1.2618	0.7051	0.2903	3
44	44	0.1667	0.8072	0.1667	0.2	1.3279	0.7536	0.0579	3
45	45	0.3056	0.759	0.3333	0.3	1.1156	0.5484	0.2559	3
46	46	0.0556	0.9277	0	0.1	1.5715	0.9997	0.3009	3
47	47	0.2389	0.6506	0.1667	0.2	1.2079	0.6297	0.1678	3
48	48	0.1	0.8554	0	0.1	1.5096	0.9365	0.2115	3
49	49	0.2944	0.6867	0.3333	0.2	1.1336	0.5512	0.2671	3
50	50	0	1	0.1667	0	1.6291	1.0507	0.3484	3