

**MATHEMATIC FOR AI**

**PERHITUNGAN RUMUS LVQ (Learning Vector Quantization) DATASET BEASISWA**

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

### **B. Perbaikan tambahan program**

```

❷ LVQ.py > ...
40 # 6. TRAINING + SIMPAN PERHITTINGAN
41 epoch_weights = []
42 epoch_calculations = []
43
44 for epoch in range(epochs):
45     print(f"\n{'='*65}")
46     print(f"EPOCH {epoch+1} | Learning Rate (β) = {alpha:.4f}")
47     print(f"\n{'='*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         ])

```

```

❷ LVQ.py > ...
79 # ===== UPDATE BOBOT LVQ1 =====
80 if W_class[winner-1] == target:
81     W[winner-1] = W[winner-1] + alpha * (x - W[winner-1])
82 else:
83     W[winner-1] = W[winner-1] - alpha * (x - W[winner-1])
84
85 epoch_calculations.append(epoch_rows)
86 epoch_weights.append(W.copy())
87 alpha *= decay
88
89
90 # 7. PREDIKSI & AKURASI
91 def lvq_predict(X, W, W_class):
92     return np.array([
93         W_class[np.argmin([euclidean_distance(x, w) for w in W])]
94         for x in X
95     ])
96
97 y_pred = lvq_predict(X_norm, W, W_class)
98 accuracy = np.mean(y_pred == y) * 100
99
100
101 # 8. SIMPAN KE EXCEL (SEMUA EPOCH)
102 writer = pd.ExcelWriter("LVQ_Training_Result_50Data_5Epoch.xlsx", engine="openpyxl")
103
104 # Parameter LVQ
105 df_param = pd.DataFrame(W_initial, columns=['X1', 'X2', 'X3', 'X4'])
106 df_param['Class'] = W_class
107 df_param.to_excel(writer, sheet_name="Parameter_LVQ", index=False)
108
109 # Bobot per epoch
110 for i, w_epoch in enumerate(epoch_weights):
111     df_epoch = pd.DataFrame(w_epoch, columns=['X1', 'X2', 'X3', 'X4'])
112     df_epoch['Class'] = W_class
113     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/P
=====
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)) = 0.657308
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

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=====
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/P
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

=====
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

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

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

```

## D. Perhitungan Manual (Excel)

### • Iterasi Unit

A	B	C	D	E	F
Parameter LVQ (diambil dari kode Python, ditanam sebagai konstanta)					
Prototipe	X1_GPA	X2_Income	X3_Dependents	X4_Achievements	Class
W1	0.944444444	0.060240964	0.666666667		0.9 1
W2	0.583333333	0.301204819		0.5	0.6 2
W3	0.138888889	0.78313253	0.166666667		0.2 3

### • Iterasi 1

Perhitungan Manual LVQ - Epoch 1 (50 data)																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																										
Alpha	\*\*\*\*\*	ID	X1	X2	X3	X4	Target	V1\_x1	V1\_x2	V1\_x3	V1\_x4	V2\_x1	V2\_x2	V2\_x3	V2\_x4	V3\_x1	V3\_x2	V3\_x3	V3\_x4	d\_V1	d\_V2	d\_V3	Vbias	Sig(x1)	V1\_x1	V1\_x2	V1\_x3	V1\_x4	V2\_x1	V2\_x2	V2\_x3	V2\_x4	V3\_x1	V3\_x2	V3\_x3	V3\_x4	V4\_x1	V4\_x2	V4\_x3	V4\_x4	V5\_x1	V5\_x2	V5\_x3	V5\_x4	V6\_x1	V6\_x2	V6\_x3	V6\_x4	V7\_x1	V7\_x2	V7\_x3	V7\_x4	V8\_x1	V8\_x2	V8\_x3	V8\_x4	V9\_x1	V9\_x2	V9\_x3	V9\_x4	V10\_x1	V10\_x2	V10\_x3	V10\_x4	V11\_x1	V11\_x2	V11\_x3	V11\_x4	V12\_x1	V12\_x2	V12\_x3	V12\_x4	V13\_x1	V13\_x2	V13\_x3	V13\_x4	V14\_x1	V14\_x2	V14\_x3	V14\_x4	V15\_x1	V15\_x2	V15\_x3	V15\_x4	V16\_x1	V16\_x2	V16\_x3	V16\_x4	V17\_x1	V17\_x2	V17\_x3	V17\_x4	V18\_x1	V18\_x2	V18\_x3	V18\_x4	V19\_x1	V19\_x2	V19\_x3	V19\_x4	V20\_x1	V20\_x2	V20\_x3	V20\_x4	V21\_x1	V21\_x2	V21\_x3	V21\_x4	V22\_x1	V22\_x2	V22\_x3	V22\_x4	V23\_x1	V23\_x2	V23\_x3	V23\_x4	V24\_x1	V24\_x2	V24\_x3	V24\_x4	V25\_x1	V25\_x2	V25\_x3	V25\_x4	V26\_x1	V26\_x2	V26\_x3	V26\_x4	V27\_x1	V27\_x2	V27\_x3	V27\_x4	V28\_x1	V28\_x2	V28\_x3	V28\_x4	V29\_x1	V29\_x2	V29\_x3	V29\_x4	V30\_x1	V30\_x2	V30\_x3	V30\_x4	V31\_x1	V31\_x2	V31\_x3	V31\_x4	V32\_x1	V32\_x2	V32\_x3	V32\_x4	V33\_x1	V33\_x2	V33\_x3	V33\_x4	V34\_x1	V34\_x2	V34\_x3	V34\_x4	V35\_x1	V35\_x2	V35\_x3	V35\_x4	V36\_x1	V36\_x2	V36\_x3	V36\_x4	V37\_x1	V37\_x2	V37\_x3	V37\_x4	V38\_x1	V38\_x2	V38\_x3	V38\_x4	V39\_x1	V39\_x2	V39\_x3	V39\_x4	V40\_x1	V40\_x2	V40\_x3	V40\_x4	V41\_x1	V41\_x2	V41\_x3	V41\_x4	V42\_x1	V42\_x2	V42\_x3	V42\_x4	V43\_x1	V43\_x2	V43\_x3	V43\_x4	V44\_x1	V44\_x2	V44\_x3	V44\_x4	V45\_x1	V45\_x2	V45\_x3	V45\_x4	V46\_x1	V46\_x2	V46\_x3	V46\_x4	V47\_x1	V47\_x2	V47\_x3	V47\_x4	V48\_x1	V48\_x2	V48\_x3	V48\_x4	V49\_x1	V49\_x2	V49\_x3	V49\_x4	V50\_x1	V50\_x2	V50\_x3	V50\_x4	V51\_x1	V51\_x2	V51\_x3	V51\_x4	V52\_x1	V52\_x2	V52\_x3	V52\_x4	V53\_x1	V53\_x2	V53\_x3	V53\_x4	V54\_x1	V54\_x2	V54\_x3	V54\_x4	V55\_x1	V55\_x2	V55\_x3	V55\_x4	V56\_x1	V56\_x2	V56\_x3	V56\_x4	V57\_x1	V57\_x2	V57\_x3	V57\_x4	V58\_x1	V58\_x2	V58\_x3	V58\_x4	V59\_x1	V59\_x2	V59\_x3	V59\_x4	V60\_x1	V60\_x2	V60\_x3	V60\_x4	V61\_x1	V61\_x2	V61\_x3	V61\_x4	V62\_x1	V62\_x2	V62\_x3	V62\_x4	V63\_x1	V63\_x2	V63\_x3	V63\_x4	V64\_x1	V64\_x2	V64\_x3	V64\_x4	V65\_x1	V65\_x2	V65\_x3	V65\_x4	V66\_x1	V66\_x2	V66\_x3	V66\_x4	V67\_x1	V67\_x2	V67\_x3	V67\_x4	V68\_x1	V68\_x2	V68\_x3	V68\_x4	V69\_x1	V69\_x2	V69\_x3	V69\_x4	V70\_x1	V70\_x2	V70\_x3	V70\_x4	V71\_x1	V71\_x2	V71\_x3	V71\_x4	V72\_x1	V72\_x2	V72\_x3	V72\_x4	V73\_x1	V73\_x2	V73\_x3	V73\_x4	V74\_x1	V74\_x2	V74\_x3	V74\_x4	V75\_x1	V75\_x2	V75\_x3	V75\_x4	V76\_x1	V76\_x2	V76\_x3	V76\_x4	V77\_x1	V77\_x2	V77\_x3	V77\_x4	V78\_x1	V78\_x2	V78\_x3	V78\_x4	V79\_x1	V79\_x2	V79\_x3	V79\_x4	V80\_x1	V80\_x2	V80\_x3	V80\_x4	V81\_x1	V81\_x2	V81\_x3	V81\_x4	V82\_x1	V82\_x2	V82\_x3	V82\_x4	V83\_x1	V83\_x2	V83\_x3	V83\_x4	V84\_x1	V84\_x2	V84\_x3	V84\_x4	V85\_x1	V85\_x2	V85\_x3	V85\_x4	V86\_x1	V86\_x2	V86\_x3	V86\_x4	V87\_x1	V87\_x2	V87\_x3	V87\_x4	V88\_x1	V88\_x2	V88\_x3	V88\_x4	V89\_x1	V89\_x2	V89\_x3	V89\_x4	V90\_x1	V90\_x2	V90\_x3	V90\_x4	V91\_x1	V91\_x2	V91\_x3	V91\_x4	V92\_x1	V92\_x2	V92\_x3	V92\_x4	V93\_x1	V93\_x2	V93\_x3	V93\_x4	V94\_x1	V94\_x2	V94\_x3	V94\_x4	V95\_x1	V95\_x2	V95\_x3	V95\_x4	V96\_x1	V96\_x2	V96\_x3	V96\_x4	V97\_x1	V97\_x2	V97\_x3	V97\_x4	V98\_x1	V98\_x2	V98\_x3	V98\_x4	V99\_x1	V99\_x2	V99\_x3	V99\_x4	V100\_x1	V100\_x2	V100\_x3	V100\_x4	V101\_x1	V101\_x2	V101\_x3	V101\_x4	V102\_x1	V102\_x2	V102\_x3	V102\_x4	V103\_x1	V103\_x2	V103\_x3	V103\_x4	V104\_x1	V104\_x2	V104\_x3	V104\_x4	V105\_x1	V105\_x2	V105\_x3	V105\_x4	V106\_x1	V106\_x2	V106\_x3	V106\_x4	V107\_x1	V107\_x2	V107\_x3	V107\_x4	V108\_x1	V108\_x2	V108\_x3	V108\_x4	V109\_x1	V109\_x2	V109\_x3	V109\_x4	V110\_x1	V110\_x2	V110\_x3	V110\_x4	V111\_x1	V111\_x2	V111\_x3	V111\_x4	V112\_x1	V112\_x2	V112\_x3	V112\_x4	V113\_x1	V113\_x2	V113\_x3	V113\_x4	V114\_x1	V114\_x2	V114\_x3	V114\_x4	V115\_x1	V115\_x2	V115\_x3	V115\_x4	V116\_x1	V116\_x2	V116\_x3	V116\_x4	V117\_x1	V117\_x2	V117\_x3	V117\_x4	V118\_x1	V118\_x2	V118\_x3	V118\_x4	V119\_x1	V119\_x2	V119\_x3	V119\_x4	V120\_x1	V120\_x2	V120\_x3	V120\_x4	V121\_x1	V121\_x2	V121\_x3	V121\_x4	V122\_x1	V122\_x2	V122\_x3	V122\_x4	V123\_x1	V123\_x2	V123\_x3	V123\_x4	V124\_x1	V124\_x2	V124\_x3	V124\_x4	V125\_x1	V125\_x2	V125\_x3	V125\_x4	V126\_x1	V126\_x2	V126\_x3	V126\_x4	V127\_x1	V127\_x2	V127\_x3	V127\_x4	V128\_x1	V128\_x2	V128\_x3	V128\_x4	V129\_x1	V129\_x2	V129\_x3	V129\_x4	V130\_x1	V130\_x2	V130\_x3	V130\_x4	V131\_x1	V131\_x2	V131\_x3	V131\_x4	V132\_x1	V132\_x2	V132\_x3	V132\_x4	V133\_x1	V133\_x2	V133\_x3	V133\_x4	V134\_x1	V134\_x2	V134\_x3	V134\_x4	V135\_x1	V135\_x2	V135\_x3	V135\_x4	V136\_x1	V136\_x2	V136\_x3	V136\_x4	V137\_x1	V137\_x2	V137\_x3	V137\_x4	V138\_x1	V138\_x2	V138\_x3	V138\_x4	V139\_x1	V139\_x2	V139\_x3	V139\_x4	V140\_x1	V140\_x2	V140\_x3	V140\_x4	V141\_x1	V141\_x2	V141\_x3	V141\_x4	V142\_x1	V142\_x2	V142\_x3	V142\_x4	V143\_x1	V143\_x2	V143\_x3	V143\_x4	V144\_x1	V144\_x2	V144\_x3	V144\_x4	V145\_x1	V145\_x2	V145\_x3	V145\_x4	V146\_x1	V146\_x2	V146\_x3	V146\_x4	V147\_x1	V147\_x2	V147\_x3	V147\_x4	V148\_x1	V148\_x2	V148\_x3	V148\_x4	V149\_x1	V149\_x2	V149\_x3	V149\_x4	V150\_x1	V150\_x2	V150\_x3	V150\_x4	V151\_x1	V151\_x2	V151\_x3	V151\_x4	V152\_x1	V152\_x2	V152\_x3	V152\_x4	V153\_x1	V153\_x2	V153\_x3	V153\_x4	V154\_x1	V154\_x2	V154\_x3	V154\_x4	V155\_x1	V155\_x2	V155\_x3	V155\_x4	V156\_x1	V156\_x2	V156\_x3	V156\_x4	V157\_x1	V157\_x2	V157\_x3	V157\_x4	V158\_x1	V158\_x2	V158\_x3	V158\_x4	V159\_x1	V159\_x2	V159\_x3	V159\_x4	V160\_x1	V160\_x2	V160\_x3	V160\_x4	V161\_x1	V161\_x2	V161\_x3	V161\_x4	V162\_x1	V162\_x2	V162\_x3	V162\_x4	V163\_x1	V163\_x2	V163\_x3	V163\_x4	V164\_x1	V164\_x2	V164\_x3	V164\_x4	V165\_x1	V165\_x2	V165\_x3	V165\_x4	V166\_x1	V166\_x2	V166\_x3	V166\_x4	V167\_x1	V167\_x2	V167\_x3	V167\_x4	V168\_x1	V168\_x2	V168\_x3	V168\_x4	V169\_x1	V169\_x2	V169\_x3	V169\_x4	V170\_x1	V170\_x2	V170\_x3	V170\_x4	V171\_x1	V171\_x2	V171\_x3	V171\_x4	V172\_x1	V172\_x2	V172\_x3	V172\_x4	V173\_x1	V173\_x2	V173\_x3	V173\_x4	V174\_x1	V174\_x2	V174\_x3	V174\_x4	V175\_x1	V175\_x2	V175\_x3	V175\_x4	V176\_x1	V176\_x2	V176\_x3	V176\_x4	V177\_x1	V177\_x2	V177\_x3	V177\_x4	V178\_x1	V178\_x2	V178\_x3	V178\_x4	V179\_x1	V179\_x2	V179\_x3	V179\_x4	V180\_x1	V180\_x2	V180\_x3	V180\_x4	V181\_x1	V181\_x2	V181\_x3	V181\_x4	V182\_x1	V182\_x2	V182\_x3	V182\_x4	V183\_x1	V183\_x2	V183\_x3	V183\_x4	V184\_x1	V184\_x2	V184\_x3	V184\_x4	V185\_x1	V185\_x2	V185\_x3	V185\_x4	V186\_x1	V186\_x2	V186\_x3	V186\_x4	V187\_x1	V187\_x2	V187\_x3	V187\_x4	V188\_x1	V188\_x2	V188\_x3	V188\_x4	V189\_x1	V189\_x2	V189\_x3	V189\_x4	V190\_x1	V190\_x2	V190\_x3	V190\_x4	V191\_x1	V191\_x2	V191\_x3	V191\_x4	V192\_x1	V192\_x2	V192\_x3	V192\_x4	V193\_x1	V193\_x2	V193\_x3	V193\_x4	V194\_x1	V194\_x2	V194\_x3	V194\_x4	V195\_x1	V195\_x2	V195\_x3	V195\_x4	V196\_x1	V196\_x2	V196\_x3	V196\_x4	V197\_x1	V197\_x2	V197\_x3	V197\_x4	V198\_x1	V198\_x2	V198\_x3	V198\_x4	V199\_x1	V199\_x2	V199\_x3	V199\_x4	V200\_x1	V200\_x2	V200\_x3	V200\_x4	V201\_x1	V201\_x2	V201\_x3	V201\_x4	V202\_x1	V202\_x2	V202\_x3	V202\_x4	V203\_x1	V203\_x2	V203\_x3	V203\_x4	V204\_x1	V204\_x2	V204\_x3	V204\_x4	V205\_x1	V205\_x2	V205\_x3	V205\_x4	V206\_x1	V206\_x2	V206\_x3	V206\_x4	V207\_x1	V207\_x2	V207\_x3	V207\_x4	V208\_x1	V208\_x2	V208\_x3	V208\_x4	V209\_x1	V209\_x2	V209\_x3	V209\_x4	V210\_x1	V210\_x2	V210\_x3	V210\_x4	V211\_x1	V211\_x2	V211\_x3	V211\_x4	V212\_x1	V212\_x2	V212\_x3	V212\_x4	V213\_x1	V213\_x2	V213\_x3	V213\_x4	V214\_x1	V214\_x2	V214\_x3	V214\_x4	V215\_x1	V215\_x2	V215\_x3	V215\_x4	V216\_x1	V216\_x2	V216\_x3	V216\_x4	V217\_x1	V217\_x2	V217\_x3	V217\_x4	V218\_x1	V218\_x2	V218\_x3	V218\_x4	V219\_x1	V219\_x2	V219\_x3	V219\_x4	V220\_x1	V220\_x2	V220\_x3	V220\_x4	V221\_x1	V221\_x2	V221\_x3	V221\_x4	V222\_x1	V222\_x2	V222\_x3	V222\_x4	V223\_x1	V223\_x2	V223\_x3	V223\_x4	V224\_x1	V224\_x2	V224\_x3	V224\_x4	V225\_x1	V225\_x2	V225\_x3	V225\_x4	V226\_x1	V226\_x2	V226\_x3	V226\_x4	V227\_x1	V227\_x2	V227\_x3	V227\_x4	V228\_x1	V228\_x2	V228\_x3	V228\_x4	V229\_x1	V229\_x2	V229\_x3	V229\_x4	V230\_x1	V230\_x2	V230\_x3	V230\_x4	V231\_x1	V231\_x2	V231\_x3	V231\_x4	V232\_x1	V232\_x2	V232\_x3	V232\_x4	V233\_x1	V233\_x2	V233\_x3	V233\_x4	V234\_x1	V234\_x2	V234\_x3	V234\_x4	V235\_x1	V235\_x2	V235\_x3	V235\_x4	V236\_x1	V236\_x2	V236\_x3	V236\_x4	V237\_x1	V237\_x2	V237\_x3	V237\_x4	V238\_x1	V238\_x2	V238\_x3	V238\_x4	V239\_x1	V239\_x2	V239\_x3	V239\_x4	V240\_x1	V240\_x2	V240\_x3	V240\_x4	V241\_x1	V241\_x2	V241\_x3	V241\_x4	V242\_x1	V242\_x2	V242\_x3	V242\_x4	V243\_x1	V243\_x2	V243\_x3	V243\_x4	V244\_x1	V244\_x2	V244\_x3	V244\_x4	V245\_x1	V245\_x2	V245\_x3	V245\_x4	V246\_x1	V246\_x2	V246\_x3	V246\_x4	V247\_x1	V247\_x2	V247\_x3	V247\_x4	V248\_x1	V248\_x2	V248\_x3	V248\_x4	V249\_x1	V249\_x2	V249\_x3	V249\_x4	V250\_x1	V250\_x2	V250\_x3	V250\_x4	V251\_x1	V251\_x2	V251\_x3	V251\_x4	V252\_x1	V252\_x2	V252\_x3	V252\_x4	V253\_x1	V253\_x2	V253\_x3	V253\_x4	V254\_x1	V254\_x2	V254\_x3	V254\_x4	V255\_x1	V255\_x2	V255\_x3	V255\_x4	V256\_x1	V256\_x2	V256\_x3	V256\_x4	V257\_x1	V257\_x2	V257\_x3	V257\_x4	V258\_x1	V258\_x2	V258\_x3	V258\_x4	V259\_x1	V259\_x2	V259\_x3	V259\_x4	V260\_x1	V260\_x2	V260\_x3	V260\_x4	V261\_x1	V261\_x2	V261\_x3	V261\_x4	V262\_x1	V262\_x2	V262\_x3	V262\_x4	V263\_x1	V263\_x2	V263\_x3	V263\_x4	V264\_x1	V264\_x2	V264\_x3	V264\_x4	V265\_x1	V265\_x2	V265\_x3	V265\_x4	V266\_x1	V266\_x2	V266\_x3	V266\_x4	V267\_x1	V267\_x2	V267\_x3	V267\_x4	V268\_x1	V268\_x2	V268\_x3	V268\_x4	V269\_x1	V269\_x

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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	0.078851	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	0.920992	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	0.6848496	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	0.656296	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	0.650099	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	0.71068797	0.8458333	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.274424	0.686207	1.426017	1		
4	0.833333	0.084337	0.666667	0.800000	1	0.9190296	0.08069935	0.66457257	0.8795533	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	0.86215314	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.347031	0.842467	1.596250	1		
6	0.888889	0.156627	0.500000	0.800000	1	0.904403	0.06367203	0.73839924	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	0.9010007	0.08400117	0.68626053	0.88502837	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.394340	0.730215	1.437616	1		
8	0.805556	0.048193	0.833333	0.700000	1	0.92265917	0.08143976	0.60907537	0.86773367	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	0.102250	0.627809	1.404822	1		
10	0.872222	0.012048	0.833333	0.800000	1	0.91150908	0.0889323	0.6598961	0.85772778	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.201237	0.655298	1.432357	1		
11	0.888889	0.108434	0.666667	0.700000	1	0.90256546	0.07055514	0.6978988	0.84510271	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.165891	0.454946	1.247628	1		
12	0.983333	0.060241	0.500000	1.000000	1	0.88863399	0.0788399	0.69110691	0.81338679	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	0.14578	0.666667	0.800000	1	0.90934896	0.07477176	0.64928238	0.85184285	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	0.90899305	0.08351566	0.68217381	0.84702935	0.59423093	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.355091	0.823328	1.594832	1		
15	0.855556	0.096386	0.833333	0.700000	1	0.91516116	0.07516192	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	0.90299305	0.08351566	0.68217381	0.84702935	0.59579152	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.608844	0.198471	0.800006	2		
23	0.638889	0.204096	0.833333	0.500000	2	0.90999305	0.08351566	0.68217381	0.84702935	0.59579152	0.36560244	0.47748346	0.48838096	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	0.08351566	0.68217381	0.84702935	0.59689579	0.36560244	0.47748346	0.48838096	0.15084375	0.82241329	0.16716493	0.13087692	0.433660	0.158695	0.947637	2		
26	0.627778	0.481928	0.833333	0.400000	2	0.90999305	0.08351566	0.68217381	0.84702935	0.61042656	0.36295194	0.55478905	0.329571954	0.4813916	0.4813916	0.4813916	0.4813916	0.16716493	0.13087692	0.748266	0.300340	0.665910	2
27	0.538889	0.385542	0.500000	0.500000	2	0.90999305	0.08351566	0.68217381	0.84702935	0.61425151	0.36285152												

- Hasil iterasi

Hasil Akhir (Setelah Iterasi / Epoch 5)										
No	ID	X1	X2	X3	X4	U_C1	U_C2	U_C3	Cluster	
1	1	0.3444	0.0602	0.6667	0.3	0.0631	0.6454	1.4265	1	
2	2	0.8611	0.0361	0.8333	0.8	0.1789	0.631	1.4133	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	0.3	0.3443	0.8203	1.5812	1	
6	6	0.8883	0.1566	0.5	0.8	0.2604	0.475	1.2375	1	
7	7	1	0.0723	0.3333	1	0.335	0.7286	1.431	1	
8	8	0.8056	0.0482	0.8333	0.7	0.3078	0.5586	1.3343	1	
9	9	0.9611	0.1325	0.6667	0.9	0.1004	0.6243	1.399	1	
10	10	0.8722	0.012	0.8333	0.8	0.1937	0.6485	1.4323	1	
11	11	0.6383	0.1084	0.6667	0.7	0.1667	0.4508	1.242	1	
12	12	0.8883	0.0602	0.5	1	0.2735	0.7094	1.4566	1	
13	13	0.3056	0.1446	0.6667	0.8	0.089	0.5243	1.3082	1	
14	14	0.3365	0.0241	1	0.3	0.3568	0.8163	1.5836	1	
15	15	0.8556	0.0364	0.8333	0.7	0.1935	0.5545	1.3333	1	
16	16	0.3722	0.0482	0.6667	1	0.2073	0.7322	1.5037	1	
17	17	0.8883	0.1205	0.5	0.8	0.2435	0.4886	1.2533	1	
18	18	0.5833	0.3012	0.5	0.6	0.4392	0.1281	0.8828	2	
19	19	0.5278	0.3614	0.3333	0.5	0.6834	0.1657	0.7198	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	0.3976	0.3333	0.6	0.6102	0.1975	0.7937	2	
23	23	0.6383	0.241	0.8333	0.5	0.431	0.3356	1.071	2	
24	24	0.5556	0.3253	0.6667	0.5	0.5523	0.1372	0.8878	2	
25	25	0.6344	0.3012	0.5	0.6	0.4347	0.1558	0.3418	2	
26	26	0.6278	0.4818	0.3333	0.4	0.7435	0.2955	0.66	2	
27	27	0.5383	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	0.5161	0.4217	0.5	0.4	0.7033	0.1754	0.686	2	
30	30	0.5944	0.3373	0.3333	0.5	0.6385	0.193	0.7657	2	
31	31	0.5722	0.4036	0.5	0.5	0.6124	0.0553	0.7653	2	
32	32	0.6556	0.253	0.6667	0.6	0.3395	0.241	1.0188	2	
33	33	0.6222	0.3133	0.5	0.5	0.5386	0.051	0.8473	2	
34	34	0.55	0.4578	0.3333	0.4	0.77	0.2563	0.6203	2	
35	35	0.1383	0.7831	0.1667	0.2	1.3308	0.7561	0.0753	3	
36	36	0.2222	0.7223	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	1.18	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	0.3633	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	1.3219	0.7536	0.0579	3	
45	45	0.3056	0.753	0.3333	0.3	1.156	0.5484	0.2553	3	
46	46	0.0556	0.9277	0	0.1	1.5715	0.9397	0.3009	3	
47	47	0.2383	0.6506	0.1667	0.2	1.2073	0.6237	0.1678	3	
48	48	0.1	0.8554	0	0.1	1.5096	0.3365	0.2115	3	
49	49	0.2944	0.6867	0.3333	0.2	1.1356	0.5512	0.2671	3	
50	50	0	0	1	0.1667	0	1.6231	1.0507	0.3484	3