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## LAMPIRAN

**Lampiran 1** Data *return* setiap aset

No.	Tanggal	NG	COP	MFST	MAPI	IPY	CNY
1	2/1/2013	0.044025	-0.05134	0.01275	0.21875	-0.02656	-0.0375
2	3/1/2013	0.154332	-0.03805	0.028777	0.166667	-0.00198	-0.02068
3	4/1/2013	0.079274	-0.06593	0.157343	-0.09341	0.062288	0.019559
4	5/1/2013	-0.08266	0.035294	0.054381	0.078788	0.017934	-0.01782
5	6/1/2013	-0.10517	-0.07348	-0.01032	-0.21348	-0.01233	0.001468
6	7/1/2013	-0.03338	0.022077	-0.07817	-0.17143	0.009384	0.020956
7	8/1/2013	0.039176	0.03376	0.048995	-0.18103	-0.00292	-0.00735
8	9/1/2013	-0.00586	0.028633	-0.00359	0.305263	0.023503	0.022964
9	10/1/2013	0.005899	-0.00557	0.063702	-0.12903	0.005722	7.25E-05
10	11/1/2013	0.104161	-0.02693	0.077119	-0.0963	0.041773	0.000169
11	12/1/2013	0.069803	0.05629	-0.01888	0.127049	0.040026	0.005085
12	1/1/2014	0.168558	-0.05785	0.011494	-0.00909	-0.04933	-0.01789
13	2/1/2014	-0.06757	-0.00391	0.012421	0.266055	0.02115	0.037837
14	3/1/2014	-0.05164	-0.05114	0.069956	-0.0942	0.011317	0.009303
15	4/1/2014	0.101579	0.000992	-0.01439	0.0208	-0.00225	0.01389
16	5/1/2014	-0.0567	0.030718	0.013366	-0.18809	-2.14E-02	-0.01891
17	6/1/2014	-0.01783	0.026598	0.018564	-0.07722	-0.00029	-0.00253
18	7/1/2014	-0.13898	0.010145	0.035012	0.087866	-0.00772	-0.02684
19	8/1/2014	0.058318	-0.02565	0.052595	0.159615	-0.00712	-0.02399
20	9/1/2014	0.013776	-0.04615	0.020471	-0.08789	0.013539	-0.03894
21	10/1/2014	-0.06018	0.013134	0.012726	-0.03636	0.015525	-0.0126
22	11/1/2014	0.055513	-0.06268	0.018317	0.056604	0.049986	-0.00095
23	12/1/2014	-0.2933	-0.01068	-0.02845	-0.09286	-0.01957	-0.01859
24	1/1/2015	-0.06854	-0.11679	-0.13025	0.15748	-0.08454	-0.06029
25	2/1/2015	0.015979	0.078541	0.085396	-0.06803	0.00928	-0.00506
26	3/1/2015	-0.03438	0.018391	-0.07298	-0.04197	-0.03648	-0.05216
27	4/1/2015	0.042045	0.053083	0.196556	0.062857	0.039026	0.046335
28	5/1/2015	-0.03962	-0.05318	-0.0366	0.039427	0.01822	-0.02155
29	6/1/2015	0.071915	-0.04318	-0.05783	-0.04655	0.00022	0.013804
30	7/1/2015	-0.04096	-0.09407	0.057758	-0.15913	-0.00176	-0.01175

31	8/1/2015	-0.00994	-0.01287	-0.06809	-0.13978	-0.00191	0.047709
32	9/1/2015	-0.06136	0.001069	0.017004	-0.25	-0.01442	-0.00631
33	10/1/2015	-0.08043	-0.00748	0.189336	0.15	-0.00896	-0.02112
34	11/1/2015	-0.03705	-0.11943	0.032485	0.188406	-0.02049	-0.028
35	12/1/2015	0.045638	0.0435	0.020791	-0.07317	0.004923	0.043221
36	1/1/2016	-0.01669	-0.03021	-0.00703	-0.01316	0.003597	0.010382
37	2/1/2016	-0.25544	0.029944	-0.07642	0.133333	-0.06597	-5.62E-05
38	3/1/2016	0.144944	0.026026	0.085495	0.129412	0.045725	0.029944
39	4/1/2016	0.111792	0.043647	-0.09705	-0.125	-0.04888	0.010714
40	5/1/2016	0.050505	-0.08036	0.062763	-0.1	0.011247	-0.01246
41	6/1/2016	0.277972	0.045476	-0.03453	0.097884	-0.06925	0.00777
42	7/1/2016	-0.01642	0.013437	0.10768	0.144578	-0.00541	0.004878
43	8/1/2016	0.003825	-0.06742	0.013761	-0.03158	0.011839	0.004571
44	9/1/2016	0.006581	0.065301	0.002436	0	-1.31E-02	0.005987
45	10/1/2016	0.041294	-0.00113	0.040278	0.195652	0.010538	-0.00767
46	11/1/2016	0.107733	0.190897	0.005674	-0.04	0.052837	-0.0201
47	12/1/2016	0.110979	-0.04716	0.031198	0.022727	0.014197	0.00166
48	1/1/2017	-0.163	0.091	0.040393	0.024074	-0.00895	0.017125
49	2/1/2017	-0.11004	-0.00713	-0.01036	-0.02712	-0.02086	-0.02194
50	3/1/2017	0.149964	-0.0199	0.029384	0.111524	-0.00503	0.00964
51	4/1/2017	0.026959	-0.01974	0.039478	0.058528	0.024277	0.023925
52	5/1/2017	-0.06258	-0.00805	0.020158	-0.0079	0.024689	0.019742
53	6/1/2017	-0.01172	0.048135	-0.01303	0.082803	0.030761	0.01164
54	7/1/2017	-0.07941	0.069716	0.054693	-0.07353	0.016986	0.028048
55	8/1/2017	0.088046	0.066897	0.028473	0.111111	0.003218	-0.01448
56	9/1/2017	-0.01086	-0.04493	-0.00374	0	0.014587	0.001657
57	10/1/2017	-0.03691	0.051777	0.11666	-0.06429	-0.00414	-0.01727
58	11/1/2017	0.044544	-0.01689	0.011902	0.058015	0.012245	0.01849
59	12/1/2017	-0.0238	0.080183	0.016277	-0.10534	0.009259	-0.00775
60	1/1/2018	0.014223	-0.02848	0.110708	0.177419	0.003329	-1.28E-05
61	2/1/2018	-0.10952	-0.02308	-0.01305	0.041096	-0.041	-0.01151
62	3/1/2018	0.024747	-0.03049	-0.02666	0.023684	0.00669	0.001737
63	4/1/2018	0.010977	0.012183	0.024652	0.060411	0.008555	-0.01044
64	5/1/2018	0.068404	0.000325	0.056886	0.021818	-0.03666	-0.02027
65	6/1/2018	-0.00949	-0.03545	-0.00233	0.067616	0.016352	0.031764
66	7/1/2018	-0.04856	-0.04231	0.075753	-0.01667	0.011603	0.02891
67	8/1/2018	0.048167	-0.06196	0.058918	-0.0565	-0.01499	-0.00387
68	9/1/2018	0.03155	0.05273	0.018161	-0.01198	0.024375	0.006172

69	10/1/2018	0.084109	-0.0492	-0.0661	-0.0303	-0.03213	-0.01067
70	11/1/2018	0.41429	0.044807	0.038199	0.09375	0.005168	-0.00213
71	12/1/2018	-0.36253	-0.0558	-0.08405	-0.08	-0.02126	0.001906
72	1/1/2019	-0.04286	0.060053	0.028158	0.248447	-0.00828	-0.02809
73	2/1/2019	-0.00071	0.05701	0.072776	0.054726	0.016211	-0.00732
74	3/1/2019	-0.05334	-0.00288	0.052754	-0.08491	-0.01816	-0.0105
75	4/1/2019	-0.03268	-0.01208	0.107343	0.025773	0.005067	0.003228
76	5/1/2019	-0.04699	-0.09074	-0.05299	-0.11558	-0.03129	0.021927
77	6/1/2019	-0.05949	0.027646	0.083118	0.022727	0.013135	0.011169
78	7/1/2019	-0.0325	-0.01548	0.017244	0.05	-0.0181	-0.02326
79	8/1/2019	0.023287	-0.0466	0.011668	0.047619	-0.03006	0.031393
80	9/1/2019	0.019694	0.012367	0.008487	0.040404	0.008476	-0.0092
81	10/1/2019	0.130043	0.024821	0.031216	-0.02913	0.022498	0.007407
82	11/1/2019	-0.13369	0.005109	0.055869	0.015	0.001578	-0.01311
83	12/1/2019	-0.04033	0.053087	0.041749	0.039409	0.009285	0.007722
84	1/1/2020	-0.15898	-0.09903	0.079455	-0.09005	-0.01248	-0.01406
85	2/1/2020	-0.08528	0.007937	-0.04829	-0.16667	-0.00898	0.001638
86	3/1/2020	-0.02613	-0.12126	-0.02654	-0.41	-0.0047	0.013273
87	4/1/2020	0.188415	0.050179	0.136326	0.387712	-0.01003	-0.00963
88	5/1/2020	-0.05131	0.038609	0.022543	0.10687	0.01908	0.02379
89	6/1/2020	-0.053	0.120764	0.110559	0.075862	0.013039	0.002096
90	7/1/2020	0.027413	0.054426	0.007371	-0.12179	0.028548	0.035162
91	8/1/2020	0.461923	0.062044	0.100093	-0.0073	0.013878	-0.00496
92	9/1/2020	-0.03916	-0.00753	-0.0674	-0.16176	-0.02271	-0.02655
93	10/1/2020	0.327266	0.006265	-0.03737	0.149123	-0.01336	-0.02036
94	11/1/2020	-0.14073	0.125184	0.057292	0.229008	0.020514	0.006389
95	12/1/2020	-0.11901	0.024902	0.039006	-0.01863	0.01383	0.015949
96	1/1/2021	0.009846	0.011083	0.042892	-0.02532	0.007534	-0.02153
97	2/1/2021	0.080733	0.150225	0.001811	0.045455	0.012909	0.002321
98	3/1/2021	-0.05882	-0.0226	0.014588	-0.08075	0.008937	-0.01683
99	4/1/2021	0.12385	0.117	0.069602	0.081081	0.011477	0.012402
100	5/1/2021	0.018765	0.047672	-0.00991	-0.125	0.020027	0.000848
101	6/1/2021	0.222371	-0.08374	0.084989	-0.10714	-0.01672	-0.0169
102	7/1/2021	0.072329	0.041152	0.051717	-0.008	-0.01139	0.001934
103	8/1/2021	0.118293	-0.02139	0.059563	0.193548	-0.00238	-0.00539
104	9/1/2021	0.340416	-0.0643	-0.06612	0.054054	-0.00801	-0.02145
105	10/1/2021	-0.07517	0.066031	0.176291	0.128205	0.022815	-0.00796
106	11/1/2021	-0.15831	-0.01801	-0.00311	-0.15909	-0.02701	-0.02574

107	12/1/2021	-0.18327	0.042752	0.017333	-0.04054	0.020119	0.000956
108	1/1/2022	0.306702	-0.03069	-0.07534	0.098592	-0.0117	-0.01058
109	2/1/2022	-0.09684	0.029585	-0.0392	0.00641	-0.00217	-0.00942
110	3/1/2022	0.28169	0.066786	0.031862	0.121019	0.043718	-0.0089
111	4/1/2022	0.283942	-0.07229	-0.09987	0.022727	0.016339	-0.00697
112	5/1/2022	0.124379	-0.02484	-0.02036	0	0.009207	0.027995
113	6/1/2022	-0.33407	-0.13701	-0.05532	0.088889	0.030121	-0.0195
114	7/1/2022	0.517146	-0.03639	0.093097	-0.09184	-0.04344	-0.0186
115	8/1/2022	0.109126	-0.01566	-0.06864	0.134831	0.026894	0.005515
116	9/1/2022	-0.25868	-0.03026	-0.10927	0.049505	0.014955	0.006091
117	10/1/2022	-0.06074	-0.01099	-0.00331	0.136792	0.036168	0.035234
118	11/1/2022	0.09048	0.107556	0.099125	0.20332	-0.02279	0.02202
119	12/1/2022	-0.35426	0.019395	-0.06005	-0.00345	-0.02305	0.000868
120	1/1/2023	-0.40022	0.109041	0.033317	-0.10035	0.007056	-0.00612
121	2/1/2023	0.023472	-0.03597	0.006497	0.161538	0.019462	-0.0006
122	3/1/2023	-0.1933	0.005032	0.155882	0	-0.00083	0.015276
123	4/1/2023	0.087545	-0.05209	0.065765	-0.09272	0.043424	0.02312
124	5/1/2023	-0.05975	-0.06292	0.068769	0.29562	-0.00832	-0.00207
125	6/1/2023	0.234775	0.033682	0.036999	-0.04789	0.057208	0.040934
126	7/1/2023	-0.05861	0.066099	-0.01357	0.171598	-0.00635	-0.00723
127	8/1/2023	0.050873	-0.04641	-0.02429	-0.02273	0.008437	0.001859
128	9/1/2023	0.058165	-0.02211	-0.03664	-0.05943	0.00057	-0.01924
129	10/1/2023	0.149198	-0.02001	0.044561	0.076923	0.016154	0.002592

## Lampiran 2 Program Software Python

```
#memanggil paket
import math as mt
import numpy as np
import pandas as pd

#load the data
from google.colab import files
files.upload()

#memanggil data
data=pd.read_csv("SAHAM1.csv")
print(data)
      NG      COP      MFST      MAPI      JPY      CNY
```

```

0    0.044025 -0.051344  0.012750  0.218750 -0.026555 -0.037499
1    0.154332 -0.038055  0.028777  0.166667 -0.001984 -0.020680
2    0.079274 -0.065934  0.157343 -0.093407  0.062288  0.019559
3   -0.082662  0.035294  0.054381  0.078788  0.017934 -0.017816
4   -0.105171 -0.073485 -0.010315 -0.213483 -0.012332  0.001468
..
124  0.234775  0.033682  0.036999 -0.047887  0.057208  0.040934
125 -0.058613  0.066099 -0.013567  0.171598 -0.006351 -0.007230
126  0.050873 -0.046407 -0.024291 -0.022727  0.008437  0.001859
127  0.058165 -0.022109 -0.036643 -0.059432  0.000570 -0.019241
128  0.149198 -0.020013  0.044561  0.076923  0.016154  0.002592

```

```
[129 rows x 6 columns]
```

```

#pendefinisian data
x1=data["NG"]
x2=data["COP"]
x3=data["MFST"]
x4=data["MAPI"]
x5=data["JPY"]
x6=data["CNY"]
i=1j
N=len(data)

#menghitung transformasi hilbert
exp = [np.cos(2*np.pi*(k/N)) + 1j*np.sin(2*np.pi*(k/N)) for k in
range(N)]

sgn1=-i*np.sign(1-N/2)
rk1=x1*exp
sig1=sum(rk1)
hdr1=sgn1*sig1
z1t=x1+i*hdr1

sgn2=-i*np.sign(2-N/2)
rk2=x2*exp
sig2=sum(rk2)
hdr2=sgn2*sig2
z2t=x2+i*hdr2

sgn3=-i*np.sign(3-N/2)
rk3=x3*exp
sig3=sum(rk3)

```



```

hdr3=sgn3*sig3
z3t=x3+i*hdr3

sgn4=-i*np.sign(4-N/2)
rk4=x4*exp
sig4=sum(rk4)
hdr4=sgn4*sig4
z4t=x4+i*hdr4

sgn5=-i*np.sign(5-N/2)
rk5=x5*exp
sig5=sum(rk5)
hdr5=sgn5*sig5
z5t=x5+i*hdr5

sgn6=-i*np.sign(6-N/2)
rk6=x6*exp
sig6=sum(rk6)
hdr6=sgn6*sig6
z6t=x6+i*hdr6

zt=(z1t,z2t,z3t,z4t,z5t,z6t)
print(zt)

(0      -0.096206+1.231498j
1       0.014100+1.231498j
2      -0.060957+1.231498j
3      -0.222893+1.231498j
4      -0.245402+1.231498j
      ...
124     0.094544+1.231498j
125    -0.198845+1.231498j
126    -0.089358+1.231498j
127    -0.082066+1.231498j
128     0.008966+1.231498j
Name: NG, Length: 129, dtype: complex128, 0      0.515205+0.509814j
1       0.528494+0.509814j
2       0.500615+0.509814j
3       0.601843+0.509814j
4       0.493064+0.509814j
      ...
124     0.600230+0.509814j
125     0.632648+0.509814j
126     0.520141+0.509814j
127     0.544440+0.509814j
128     0.546535+0.509814j

```

```

Name: COP, Length: 129, dtype: complex128, 0      0.415097+0.102520j
1      0.431124+0.102520j
2      0.559690+0.102520j
3      0.456728+0.102520j
4      0.392032+0.102520j
...
124    0.439346+0.102520j
125    0.388780+0.102520j
126    0.378055+0.102520j
127    0.365704+0.102520j
128    0.446907+0.102520j
Name: MFST, Length: 129, dtype: complex128, 0      -
0.018135+0.412183j
1      -0.070218+0.412183j
2      -0.330291+0.412183j
3      -0.158097+0.412183j
4      -0.450368+0.412183j
...
124    -0.284772+0.412183j
125    -0.065287+0.412183j
126    -0.259612+0.412183j
127    -0.296316+0.412183j
128    -0.159961+0.412183j
Name: MAPI, Length: 129, dtype: complex128, 0      -
0.309197+0.198615j
1      -0.284625+0.198615j
2      -0.220354+0.198615j
3      -0.264708+0.198615j
4      -0.294974+0.198615j
...
124    -0.225433+0.198615j
125    -0.288993+0.198615j
126    -0.274204+0.198615j
127    -0.282071+0.198615j
128    -0.266488+0.198615j
Name: JPY, Length: 129, dtype: complex128, 0      0.077965-0.003427j
1      0.094785-0.003427j
2      0.135023-0.003427j
3      0.097648-0.003427j
4      0.116932-0.003427j
...
124    0.156398-0.003427j
125    0.108234-0.003427j
126    0.117323-0.003427j
127    0.096223-0.003427j
128    0.118056-0.003427j
Name: CNY, Length: 129, dtype: complex128)

```

```

# membentuk matriks diagonal E[z1t-E(z1t)]*[(z1t-E(z1t))*]
e1=sum(z1t)/N
z1=z1t-e1

```

```

z1tk=np.conj(z1t)
ze1k=np.conj(ze1)
z1tot=ze1*ze1k
ez1t=sum(z1tot)/N
z1=complex(round(ez1t.real,18),round(ez1t.imag,18))

e2=sum(z2t)/N
ze2=z2t-e2
z2tk=np.conj(z2t)
ze2k=np.conj(ze2)
z2tot=ze2*ze2k
ez2t=sum(z2tot)/N
z2=complex(round(ez2t.real,18),round(ez2t.imag,18))

e3=sum(z3t)/N
ze3=z3t-e3
z3tk=np.conj(z3t)
ze3k=np.conj(ze3)
z3tot=ze3*ze3k
ez3t=sum(z3tot)/N
z3=complex(round(ez3t.real,18),round(ez3t.imag,18))

e4=sum(z4t)/N
ze4=z4t-e4
z4tk=np.conj(z4t)
ze4k=np.conj(ze4)
z4tot=ze4*ze4k
ez4t=sum(z4tot)/N
z4=complex(round(ez4t.real,18),round(ez4t.imag,18))

e5=sum(z5t)/N
ze5=z5t-e5
z5tk=np.conj(z5t)
ze5k=np.conj(ze5)
z5tot=ze5*ze5k
ez5t=sum(z5tot)/N
z5=complex(round(ez5t.real,18),round(ez5t.imag,18))

e6=sum(z6t)/N
ze6=z6t-e6
z6tk=np.conj(z6t)

```

```

ze6k=np.conj(ze6)
z6tot=ze6*ze6k
ez6t=sum(z6tot)/N
z6=complex(round(ez6t.real,18),round(ez6t.imag,18))

#membentuk matriks non diagonal
z7tot=ze1*ze2
ez7t=sum(z7tot)/N
z7=complex(round(ez7t.real,18),round(ez7t.imag,18))

z12tot=ze2*ze1
ez12t=sum(z12tot)/N
z12=complex(round(ez12t.real,18),round(ez12t.imag,18))

z8tot=ze1*ze3
ez8t=sum(z8tot)/N
z8=complex(round(ez8t.real,18),round(ez8t.imag,18))

z17tot=ze3*ze1
ez17t=sum(z17tot)/N
z17=complex(round(ez17t.real,18),round(ez17t.imag,18))

z9tot=ze1*ze4
ez9t=sum(z9tot)/N
z9=complex(round(ez9t.real,18),round(ez9t.imag,18))

z22tot=ze4*ze1
ez22t=sum(z22tot)/N
z22=complex(round(ez22t.real,18),round(ez22t.imag,18))

z10tot=ze1*ze5
ez10t=sum(z10tot)/N
z10=complex(round(ez10t.real,18),round(ez10t.imag,18))

z27tot=ze5*ze1
ez27t=sum(z27tot)/N
z27=complex(round(ez27t.real,18),round(ez27t.imag,18))

z11tot=ze1*ze6
ez11t=sum(z11tot)/N
z11=complex(round(ez11t.real,18),round(ez11t.imag,18))

```

```

z32tot=ze6*ze1
ez32t=sum(z32tot)/N
z32=complex(round(ez32t.real,18),round(ez32t.imag,18))

z13tot=ze2*ze3
ez13t=sum(z13tot)/N
z13=complex(round(ez13t.real,18),round(ez13t.imag,18))

z18tot=ze3*ze2
ez18t=sum(z18tot)/N
z18=complex(round(ez18t.real,18),round(ez18t.imag,18))

z14tot=ze2*ze4
ez14t=sum(z14tot)/N
z14=complex(round(ez14t.real,18),round(ez14t.imag,18))

z23tot=ze4*ze2
ez23t=sum(z23tot)/N
z23=complex(round(ez23t.real,18),round(ez23t.imag,18))

z15tot=ze2*ze5
ez15t=sum(z15tot)/N
z15=complex(round(ez15t.real,18),round(ez15t.imag,18))

z28tot=ze5*ze2
ez28t=sum(z28tot)/N
z28=complex(round(ez28t.real,18),round(ez28t.imag,18))

z16tot=ze2*ze6
ez16t=sum(z16tot)/N
z16=complex(round(ez16t.real,18),round(ez16t.imag,18))

z33tot=ze6*ze2
ez33t=sum(z33tot)/N
z33=complex(round(ez33t.real,18),round(ez33t.imag,18))

z19tot=ze3*ze4
ez19t=sum(z19tot)/N
z19=complex(round(ez19t.real,18),round(ez19t.imag,18))

```

```

z24tot=ze4*ze3
ez24t=sum(z24tot)/N
z24=complex(round(ez24t.real,18),round(ez24t.imag,18))

z20tot=ze3*ze5
ez20t=sum(z20tot)/N
z20=complex(round(ez20t.real,18),round(ez20t.imag,18))

z29tot=ze5*ze3
ez29t=sum(z29tot)/N
z29=complex(round(ez29t.real,18),round(ez29t.imag,18))

z21tot=ze3*ze6
ez21t=sum(z21tot)/N
z21=complex(round(ez21t.real,18),round(ez21t.imag,18))

z34tot=ze6*ze3
ez34t=sum(z34tot)/N
z34=complex(round(ez34t.real,18),round(ez34t.imag,18))

z25tot=ze4*ze5
ez25t=sum(z25tot)/N
z25=complex(round(ez25t.real,18),round(ez25t.imag,18))

z30tot=ze5*ze4
ez30t=sum(z30tot)/N
z30=complex(round(ez30t.real,18),round(ez30t.imag,18))

z26tot=ze4*ze6
ez26t=sum(z26tot)/N
z26=complex(round(ez26t.real,18),round(ez26t.imag,18))

z35tot=ze6*ze4
ez35t=sum(z35tot)/N
z35=complex(round(ez35t.real,18),round(ez35t.imag,18))

z31tot=ze5*ze6
ez31t=sum(z31tot)/N
z31=complex(round(ez31t.real,18),round(ez31t.imag,18))

z36tot=ze6*ze5

```

```

ez36t=sum(z36tot)/N
z36=complex(round(ez36t.real,18),round(ez36t.imag,18))

#matriks varians-kovarians dan menghitung invers
matrix=np.matrix([[z1,z7,z8,z9,z10,z11],[z12,z2,z13,z14,z15,z16],
[z17,z18,z3,z19,z20,z21],[z22,z23,z24,z4,z25,z26],[z27,z28,z29,z3
0,z5,z31],[z32,z33,z34,z35,z36,z6]])
invers=np.linalg.inv(matrix)
print(invers)

[[ 4.74422579e+01-0.j  3.33809421e+00-0.j -9.03247736e+00+0.j
 -6.93644599e+00+0.j -1.41956806e+01+0.j  3.97167973e-01-0.j]
 [ 3.33809421e+00-0.j  3.37762332e+02-0.j -5.04452146e+01+0.j
 -2.08398347e+01+0.j -8.60120276e+01+0.j -1.08539552e+02+0.j]
 [-9.03247736e+00+0.j -5.04452146e+01+0.j  3.12743854e+02-0.j
 -1.78269019e+01+0.j -2.05722082e+02+0.j  1.36975132e+01-0.j]
 [-6.93644599e+00+0.j -2.08398347e+01+0.j -1.78269019e+01+0.j
  7.06718214e+01-0.j -2.57891406e+00+0.j  4.12641916e+01-0.j]
 [-1.41956806e+01+0.j -8.60120276e+01+0.j -2.05722082e+02+0.j
 -2.57891406e+00+0.j  2.22899887e+03+0.j -1.07783036e+03+0.j]
 [ 3.97167973e-01+0.j -1.08539552e+02+0.j  1.36975132e+01+0.j
  4.12641916e+01+0.j -1.07783036e+03+0.j  3.30819923e+03+0.j]]

#membuat vector
vector = np.array([1, 1, 1, 1, 1, 1]).reshape(-1, 1)
vt=np.transpose(vector)

#membuat vektor Sigma^2
sig=np.array([z1,z2,z3,z4,z5,z6]).reshape(-1, 1)

# menghitung bobot mula mula lamda=0
iv = invers @ vector
vi = vt @ invers
viv = vi @ vector
imu = invers @ sig
vim = vi @ sig
ipv = vim / viv
bot = imu - (iv * ipv)
bobot1 = iv / viv
print("bobot minimum", bobot1)
bobot minimum [[0.00651908+0.j]
 [0.02334997+0.j]
 [0.01346905+0.j]
 [0.01977912+0.j]
 [0.26142821+0.j]]

```

```

[0.67545457+0.j]]

# Menghitung Bobot Complex
# Inisialisasi bobot awal
bobot = bobot1

# Menghitung dan mencetak total sum untuk bobot awal
total_sum = np.sum(bobot)
print(f"Inisial bobot: {bobot}")
print(f"Total sum: {total_sum}")
print("\n")

# Inisialisasi variabel untuk menyimpan bobot terakhir yang valid
bobot_optimal = None

# Iterasi untuk nilai t dari 0.01 hingga 0.09
for t in np.arange(0.01, 0.1, 0.01):
    # Menghitung bobot untuk nilai t saat ini
    bobot = bobot + (t * bobot)

    # Pengecekan apakah ada elemen negatif dalam bobot
    if any((i < 0).any() for i in bobot):
        print("Ditemukan bobot minus. Iterasi dihentikan.")
        break

    # Menyimpan bobot saat ini sebagai bobot terakhir yang valid
    bobot_optimal = bobot

    # Menghitung dan mencetak total sum untuk bobot saat ini
    total_sum = np.sum(bobot)
    print(f"Bobot untuk t={t}: {bobot}")
    print(f"Total sum untuk t={t}: {total_sum}")
    print("\n")

# Menggunakan bobot terakhir yang valid sebagai hasil
hasil_akhir = bobot_optimal
print(f"Bobot akhir yang tidak negatif: {hasil_akhir}")

Inisial bobot: [[0.00651908+0.j]
[0.02334997+0.j]
[0.01346905+0.j]]

```



[0.01977912+0.j]  
[0.26142821+0.j]  
[0.67545457+0.j]]  
Total sum: (1+0j)

Bobot untuk t=0.01: [[0.01524315+0.j]  
[0.02846571+0.j]  
[0.01715644+0.j]  
[0.02710394+0.j]  
[0.24863747+0.j]  
[0.66339329+0.j]]  
Total sum untuk t=0.01: (1+0j)

Bobot untuk t=0.02: [[0.03269129+0.j]  
[0.03869719+0.j]  
[0.02453122+0.j]  
[0.04175357+0.j]  
[0.22305598+0.j]  
[0.63927074+0.j]]  
Total sum untuk t=0.02: (1+0j)

Bobot untuk t=0.03: [[0.0588635 +0.j]  
[0.05404441+0.j]  
[0.0355934 +0.j]  
[0.06372802+0.j]  
[0.18468376+0.j]  
[0.60308691+0.j]]  
Total sum untuk t=0.03: (1+0j)

Bobot untuk t=0.04: [[0.09375978+0.j]  
[0.07450737+0.j]  
[0.05034297+0.j]  
[0.09302728+0.j]  
[0.13352079+0.j]  
[0.5548418 +0.j]]  
Total sum untuk t=0.04: (1+0j)

Bobot untuk t=0.05: [[0.13738014+0.j]  
[0.10008607+0.j]  
[0.06877993+0.j]  
[0.12965136+0.j]  
[0.06956708+0.j]  
[0.49453542+0.j]]  
Total sum untuk t=0.05: (1+0j)

Ditemukan bobot minus. Iterasi dihentikan.

```

Bobot akhir yang tidak negatif: [[0.13738014+0.j]
 [0.10008607+0.j]
 [0.06877993+0.j]
 [0.12965136+0.j]
 [0.06956708+0.j]
 [0.49453542+0.j]]

# menghitung portofolio
z1w = x1 * hasil_akhir[0, 0]
z2w = x2 * hasil_akhir[1, 0]
z3w = x3 * hasil_akhir[2, 0]
z4w = x4 * hasil_akhir[3, 0]
z5w = x5 * hasil_akhir[4, 0]
z6w = x6 * hasil_akhir[5, 0]
zw=z1w+z2w+z3w+z4w+z5w+z6w
print(zw)

0      0.009756+0.000000j
1      0.030616+0.000000j
2      0.017009+0.000000j
3      -0.001431+0.000000j
4      -0.050323+0.000000j
...
124    0.056184+0.000000j
125    0.015861+0.000000j
126    -0.000767+0.000000j
127    -0.013923+0.000000j
128    0.033937+0.000000j
Length: 129, dtype: complex128

# menghitung standar deviasi
std_dev_reall = np.std(np.real(zw))
std_dev_imagl = np.std(np.imag(zw))

print("Standar Deviasi pada Bagian Real:", std_dev_reall)
print("Standar Deviasi pada Bagian Imajiner:", std_dev_imagl)

# menghitung Value at Risk
VaR95=std_dev_reall*(1.65)*(1000000)
print("Nilai VaR 95%:", VaR95)

Standar Deviasi pada Bagian Real: 0.03201907146862666
Standar Deviasi pada Bagian Imajiner: 0.0
Nilai VaR 95%: 52831.46792323399

```

```

#Transformasi Fourier
#memanggil paket
import math as mt
import numpy as np
import pandas as pd

#load the data
from google.colab import files
files.upload()

pip install numpy matplotlib

import numpy as np
import matplotlib.pyplot as plt

#memanggil data
data=pd.read_csv("SAHAM1.csv")
print(data)
#pendefinisian data
x1=data["NG"]
x2=data["COP"]
x3=data["MFST"]
x4=data["MAPI"]
x5=data["JPY"]
x6=data["CNY"]
i=1j
N=len(data)

      NG      COP      MFST      MAPI      JPY      CNY
0   0.044025 -0.051344  0.012750  0.218750 -0.026555 -0.037499
1   0.154332 -0.038055  0.028777  0.166667 -0.001984 -0.020680
2   0.079274 -0.065934  0.157343 -0.093407  0.062288  0.019559
3  -0.082662  0.035294  0.054381  0.078788  0.017934 -0.017816
4  -0.105171 -0.073485 -0.010315 -0.213483 -0.012332  0.001468
..      ...      ...      ...      ...      ...      ...
124  0.234775  0.033682  0.036999 -0.047887  0.057208  0.040934
125 -0.058613  0.066099 -0.013567  0.171598 -0.006351 -0.007230
126  0.050873 -0.046407 -0.024291 -0.022727  0.008437  0.001859
127  0.058165 -0.022109 -0.036643 -0.059432  0.000570 -0.019241
128  0.149198 -0.020013  0.044561  0.076923  0.016154  0.002592

[129 rows x 6 columns]

# Hitung Discrete Fourier Transform (DFT)

```

```

X1 = np.fft.fft(x1)
print(X1)
[ 1.40296433+0.j          0.14023122+1.23149843j -
 0.94385694+0.8582028j    0.40489706-1.71998427j  1.15810892-1.84642802j -0.25617617-
 0.4842747j              1.91095456-0.37632453j  1.11100606+0.77398161j
 0.27016211+1.59707996j  0.75844296+2.07833038j  1.1635019 +1.28250061j -
 2.17036579+1.64639071j -1.42631123-1.17017345j  1.70941734-0.07629278j  0.86115197-
 0.09012476j            0.15475523-0.24571678j -0.29323304-0.87283336j  0.41260675-
 1.34032943j            0.11152125-0.51846305j  1.25531554+0.20163574j
 1.51854822+0.32020469j  0.8040502 +0.30886791j -2.59656156+0.6925899j  1.2713339 -
 1.75330969j            0.58378065+0.21935569j -0.37707725+0.67449046j
 1.27959225+0.86824619j  0.5395101 -0.94751431j -0.98095983+0.43089821j -0.73518395-
 2.12708869j            -0.16006463-0.25064823j  0.02904336+0.1496485j  0.11746793-
 1.02659199j            -0.277192 +0.51237363j  1.66402311+0.85479032j -
 0.40205243+1.65078683j -0.85508867+1.8916725j -2.71222935+1.60349679j -1.87519474-
 2.95038732j            0.24877976-2.14239756j -0.23338124-1.79405488j
 0.89614744+0.69238234j  1.60153379+0.50866998j  0.25745398+1.53001618j -
 2.19038496+1.00425247j  1.50446701-0.42740037j -1.41973639-0.52502011j -
 0.59070131+0.38430275j -0.32659045+0.03741105j  1.53687338-1.71856514j  0.2980138
+0.93867268j            -1.23964902-1.08806694j -0.30812174-0.37070213j -0.40224056-
 0.96218594j            -0.42967 +2.23491703j  0.96025909-0.11841377j  1.49145354-
 0.09749745j            0.00765064-0.30703409j  1.09584525+0.45209511j
 0.55618285+1.38622575j  0.05016702+1.84247678j -3.0507722 +2.00491335j -1.39570867-
 2.50386355j            0.47152546+0.02496281j -0.41913105+0.22215973j -0.41913105-
 0.22215973j            0.47152546-0.02496281j -1.39570867+2.50386355j -3.0507722 -
 2.00491335j            0.05016702-1.84247678j  0.55618285-1.38622575j  1.09584525-
 0.45209511j

```

```

0.00765064+0.30703409j 1.49145354+0.09749745j
0.96025909+0.11841377j
-0.42967 -2.23491703j -0.40224056+0.96218594j -
0.30812174+0.37070213j
-1.23964902+1.08806694j 0.2980138 -0.93867268j
1.53687338+1.71856514j
-0.32659045-0.03741105j -0.59070131-0.38430275j -
1.41973639+0.52502011j
1.50446701+0.42740037j -2.19038496-1.00425247j 0.25745398-
1.53001618j
1.60153379-0.50866998j 0.89614744-0.69238234j -
0.23338124+1.79405488j
0.24877976+2.14239756j -1.87519474+2.95038732j -2.71222935-
1.60349679j
-0.85508867-1.8916725j -0.40205243-1.65078683j 1.66402311-
0.85479032j
-0.277192 -0.51237363j 0.11746793+1.02659199j 0.02904336-
0.1496485j
-0.16006463+0.25064823j -0.73518395+2.12708869j -0.98095983-
0.43089821j
0.5395101 +0.94751431j 1.27959225-0.86824619j -0.37707725-
0.67449046j
0.58378065-0.21935569j 1.2713339 +1.75330969j -2.59656156-
0.6925899j
0.8040502 -0.30886791j 1.51854822-0.32020469j 1.25531554-
0.20163574j
0.11152125+0.51846305j 0.41260675+1.34032943j -
0.29323304+0.87283336j
0.15475523+0.24571678j 0.86115197+0.09012476j
1.70941734+0.07629278j
-1.42631123+1.17017345j -2.17036579-1.64639071j 1.1635019 -
1.28250061j
0.75844296-2.07833038j 0.27016211-1.59707996j 1.11100606-
0.77398161j
1.91095456+0.37632453j -0.25617617+0.4842747j
1.15810892+1.84642802j
0.40489706+1.71998427j -0.94385694-0.8582028j 0.14023122-
1.23149843j]

```

```
# Hitung Discrete Fourier Transform (DFT)
```

```
X2 = np.fft.fft(x2)
```

```
print(X2)
```

```

[ 0.18885736+0.j -0.56654857+0.50981355j -
0.53631672+0.34240652j
0.69227595-0.94275746j -0.06336401+0.24425756j -0.4546964
+0.71629675j
-0.5106128 -0.28376107j 0.00997073-0.04838238j -0.73646669-
0.00571736j
-0.14772307-0.17642103j -0.12881946-0.51224326j 0.51027763-
0.07863593j]

```

0.10599457-0.24451072j -0.16709523+0.91636414j -  
 0.41155864+1.47878148j  
 -0.17450333-0.44615481j 0.05839672+0.44471207j -  
 0.26824408+0.03584954j  
 -0.41168497-0.82994773j -0.20541505+0.26118813j 0.0623805  
 +0.06355168j  
 0.00373918-0.45789963j 0.12124554-0.08932527j -0.53124935-  
 0.14610984j  
 -0.09713164-0.0590984j 1.15377342-0.07957106j 0.16670329-  
 0.02658339j  
 0.26996948-0.54258431j -0.60593119+0.27233102j  
 0.29516679+0.04422866j  
 -0.45380798+0.0267032j -0.01870096-0.14752537j  
 0.48888632+0.002475j  
 -0.03388177+0.19657844j 0.06087613+0.48000273j -  
 0.09652916+0.28682952j  
 -0.18203181+0.18961092j 0.08718481+0.65782953j  
 0.32812451+0.07707155j  
 -0.04036846-0.06291835j -0.49119151+0.55173579j  
 0.08459373+0.01657096j  
 0.41531981-0.40044269j 0.14360329+0.18088625j -  
 0.29318681+0.45939656j  
 -0.49971793-0.17141339j 0.31155221-0.52042314j -  
 0.12791421+0.22353625j  
 -0.6248341 +0.39111207j 0.06093419-1.09030997j 0.00749117-  
 0.48234133j  
 0.91284358+0.03952101j 0.53586129+0.67810218j -0.5520866 -  
 0.1897934j  
 0.75496435-0.4145361j -0.07311268-0.53569705j -0.6068565 -  
 0.45846387j  
 0.05800284+0.49008711j 0.23070855-0.471375j -  
 1.40479137+0.40689419j  
 0.64114198-0.31701617j -0.79049948-0.63416465j -0.71321629-  
 0.56891585j  
 0.60655547+0.46748089j 0.43544938+0.56599255j 0.43544938-  
 0.56599255j  
 0.60655547-0.46748089j -0.71321629+0.56891585j -  
 0.79049948+0.63416465j  
 0.64114198+0.31701617j -1.40479137-0.40689419j  
 0.23070855+0.471375j  
 0.05800284-0.49008711j -0.6068565 +0.45846387j -  
 0.07311268+0.53569705j  
 0.75496435+0.4145361j -0.5520866 +0.1897934j 0.53586129-  
 0.67810218j  
 0.91284358-0.03952101j 0.00749117+0.48234133j  
 0.06093419+1.09030997j  
 -0.6248341 -0.39111207j -0.12791421-0.22353625j  
 0.31155221+0.52042314j  
 -0.49971793+0.17141339j -0.29318681-0.45939656j 0.14360329-  
 0.18088625j  
 0.41531981+0.40044269j 0.08459373-0.01657096j -0.49119151-  
 0.55173579j

```

-0.04036846+0.06291835j  0.32812451-0.07707155j  0.08718481-
0.65782953j
-0.18203181-0.18961092j -0.09652916-0.28682952j  0.06087613-
0.48000273j
-0.03388177-0.19657844j  0.48888632-0.002475j  -
0.01870096+0.14752537j
-0.45380798-0.0267032j  0.29516679-0.04422866j -0.60593119-
0.27233102j
  0.26996948+0.54258431j  0.16670329+0.02658339j
1.15377342+0.07957106j
-0.09713164+0.0590984j -0.53124935+0.14610984j
0.12124554+0.08932527j
  0.00373918+0.45789963j  0.0623805 -0.06355168j -0.20541505-
0.26118813j
-0.41168497+0.82994773j -0.26824408-0.03584954j  0.05839672-
0.44471207j
-0.17450333+0.44615481j -0.41155864-1.47878148j -0.16709523-
0.91636414j
  0.10599457+0.24451072j  0.51027763+0.07863593j  -
0.12881946+0.51224326j
-0.14772307+0.17642103j -0.73646669+0.00571736j
0.00997073+0.04838238j
-0.5106128 +0.28376107j -0.4546964 -0.71629675j -0.06336401-
0.24425756j
  0.69227595+0.94275746j -0.53631672-0.34240652j -0.56654857-
0.50981355j]

```

```
# Hitung Discrete Fourier Transform (DFT)
```

```
X3 = np.fft.fft(x3)
```

```
print(X3)
```

```

[ 2.74350998+0.j          -0.40234691+0.10251993j  0.22442563-
0.37292854j
  0.64474035-0.37084578j  0.41812605+0.23468222j
0.27048524+0.21869354j
-0.45063199+0.6453621j -0.56211698+0.23378354j -0.29636679-
0.10558235j
  0.36890854-0.69071684j -0.12679964-0.21325669j -0.43505679-
0.16119828j
  0.38746463-0.28029435j -0.13567047-0.44669198j  0.39274447-
0.40784161j
  0.41361353-0.26197772j  1.33565984-0.61877959j -0.60338256-
0.33490204j
-0.10993834-0.4747736j  0.35636489+0.50022012j  0.43773278-
0.37692783j
  0.13287684-0.90813988j  0.061132 -0.37765355j  -
0.47692637+0.52942611j
-0.31509526+0.03378537j  0.11429021-0.09916212j -0.51461367-
0.00332457j
-0.55088959+0.26594257j -0.49828755+0.48229839j -0.47456363-
0.59382704j

```

0.07072209-0.10329931j 0.48290372-0.09713752j 0.66395267-  
 0.40891592j  
 0.23681294+0.47512135j -0.71485016+0.92471399j -  
 0.50470144+0.39692716j  
 -0.1822969 -0.22066812j 0.01515471+0.410188j -  
 0.96973612+0.07575831j  
 0.35968294+0.24706741j -0.15491924+0.02483643j  
 0.47027342+0.18037459j  
 0.1064606 -0.21247382j -0.18786645+1.24534932j 0.37199916-  
 0.14287349j  
 0.13017056-0.25721586j -0.26850432-0.4464609j -  
 0.40938846+0.25717305j  
 -0.07847799+0.79497103j 0.66692064-0.53187394j 0.40599536-  
 0.73745271j  
 -0.29908479-0.34741819j -0.47744649+1.3295624j  
 0.54567071+0.0786155j  
 -0.13779104+0.55207729j -1.17925277+0.09347281j 0.41904078-  
 0.5981466j  
 -0.07372395+0.68853932j -0.24155126+0.91862642j  
 0.17335328+1.0355726j  
 0.17811539-0.90291906j -0.66611154-0.74669091j -0.09055579-  
 0.4691533j  
 1.0541381 +0.39443993j 0.12966252+0.12285074j 0.12966252-  
 0.12285074j  
 1.0541381 -0.39443993j -0.09055579+0.4691533j -  
 0.66611154+0.74669091j  
 0.17811539+0.90291906j 0.17335328-1.0355726j -0.24155126-  
 0.91862642j  
 -0.07372395-0.68853932j 0.41904078+0.5981466j -1.17925277-  
 0.09347281j  
 -0.13779104-0.55207729j 0.54567071-0.0786155j -0.47744649-  
 1.3295624j  
 -0.29908479+0.34741819j 0.40599536+0.73745271j  
 0.66692064+0.53187394j  
 -0.07847799-0.79497103j -0.40938846-0.25717305j -  
 0.26850432+0.4464609j  
 0.13017056+0.25721586j 0.37199916+0.14287349j -0.18786645-  
 1.24534932j  
 0.1064606 +0.21247382j 0.47027342-0.18037459j -0.15491924-  
 0.02483643j  
 0.35968294-0.24706741j -0.96973612-0.07575831j 0.01515471-  
 0.410188j  
 -0.1822969 +0.22066812j -0.50470144-0.39692716j -0.71485016-  
 0.92471399j  
 0.23681294-0.47512135j 0.66395267+0.40891592j  
 0.48290372+0.09713752j  
 0.07072209+0.10329931j -0.47456363+0.59382704j -0.49828755-  
 0.48229839j  
 -0.55088959-0.26594257j -0.51461367+0.00332457j  
 0.11429021+0.09916212j  
 -0.31509526-0.03378537j -0.47692637-0.52942611j 0.061132  
 +0.37765355j



```

0.13287684+0.90813988j 0.43773278+0.37692783j 0.35636489-
0.50022012j
-0.10993834+0.4747736j -0.60338256+0.33490204j
1.33565984+0.61877959j
0.41361353+0.26197772j 0.39274447+0.40784161j -
0.13567047+0.44669198j
0.38746463+0.28029435j -0.43505679+0.16119828j -
0.12679964+0.21325669j
0.36890854+0.69071684j -0.29636679+0.10558235j -0.56211698-
0.23378354j
-0.45063199-0.6453621j 0.27048524-0.21869354j 0.41812605-
0.23468222j
0.64474035+0.37084578j 0.22442563+0.37292854j -0.40234691-
0.10251993j]

```

```
# Hitung Discrete Fourier Transform (DFT)
```

```
X4 = np.fft.fft(x4)
```

```
print(X4)
```

```

[ 2.10141504+0.00000000e+00j 0.2368845 +4.12183351e-01j
 0.70723192+1.62133244e+00j 0.10536078+2.97629183e-01j
-0.28565927+1.41041215e-01j -0.71817049+9.64550149e-01j
 0.1651417 +2.82516236e-01j 0.83149602+1.55144426e-01j
-0.21134427+5.75248503e-01j -0.23111325-1.69789108e-03j
 0.93364554+1.14273125e+00j 1.4502856 +8.88133882e-01j
 0.37133839+8.76977245e-02j 0.54262 -4.36385849e-01j
-0.31864447+5.43694019e-01j 2.06612071-9.42402098e-01j
-0.02415628-1.15037785e-01j -1.88195028-1.76174914e+00j
 1.00956915-9.28305370e-01j 1.38185633+4.30468999e-01j
 0.84757574-9.04367064e-02j 1.15767931-4.19833026e-01j
 1.04457244+4.10140955e-01j -0.64002249-4.74617962e-01j
 1.69222275-2.79209942e+00j 0.73545411-9.97281197e-02j
-2.32710969+5.61789128e-03j -1.19910772-7.07531761e-01j
 1.15226075+1.13086510e+00j 0.60999721-8.37169823e-01j
-0.20357034+2.31369166e-01j 0.28605068+1.68986166e+00j
 0.4979209 +4.01580832e-01j 0.26680585-5.70247969e-01j
-0.58785581+2.21243712e+00j 0.71370891+5.58021972e-01j
 1.54666282-2.63997739e-01j -0.59713172+4.61647024e-01j
 0.42554365-1.69110122e-01j 1.73512489-1.22106461e+00j
 0.47864114-8.40591387e-01j -0.50853583-1.02549888e+00j
 1.36944062-1.12881323e+00j 1.09132549+1.55774476e+00j
 0.48025451-5.08204345e-01j 1.17738019-1.07934694e+00j
-0.21621176+1.12784683e-01j -0.28826782+1.82234629e+00j
-1.39909208+4.41859291e-01j -1.63920261+1.12045923e+00j
-0.28026611+5.48368737e-01j 0.02893588-7.36031164e-01j
 0.66959646+1.50366255e+00j -0.31378493-1.68108045e+00j
 0.63348607+1.30356912e+00j -0.4246067 -1.52642386e-01j
 0.44746198-1.98482596e+00j 0.69057362-2.57604973e+00j
-0.84215003-1.19442027e+00j -0.88737301-1.29464608e+00j
 0.01254817-9.10351558e-01j -0.77652979+5.85311364e-01j
 0.05904715-1.30851376e+00j 0.99271504+9.26114339e-01j
-0.78401273+1.00340150e+00j -0.78401273-1.00340150e+00j]

```

```

0.99271504-9.26114339e-01j 0.05904715+1.30851376e+00j
-0.77652979-5.85311364e-01j 0.01254817+9.10351558e-01j
-0.88737301+1.29464608e+00j -0.84215003+1.19442027e+00j
0.69057362+2.57604973e+00j 0.44746198+1.98482596e+00j
-0.4246067 +1.52642386e-01j 0.63348607-1.30356912e+00j
-0.31378493+1.68108045e+00j 0.66959646-1.50366255e+00j
0.02893588+7.36031164e-01j -0.28026611-5.48368737e-01j
-1.63920261-1.12045923e+00j -1.39909208-4.41859291e-01j
-0.28826782-1.82234629e+00j -0.21621176-1.12784683e-01j
1.17738019+1.07934694e+00j 0.48025451+5.08204345e-01j
1.09132549-1.55774476e+00j 1.36944062+1.12881323e+00j
-0.50853583+1.02549888e+00j 0.47864114+8.40591387e-01j
1.73512489+1.22106461e+00j 0.42554365+1.69110122e-01j
-0.59713172-4.61647024e-01j 1.54666282+2.63997739e-01j
0.71370891-5.58021972e-01j -0.58785581-2.21243712e+00j
0.26680585+5.70247969e-01j 0.4979209 -4.01580832e-01j
0.28605068-1.68986166e+00j -0.20357034-2.31369166e-01j
0.60999721+8.37169823e-01j 1.15226075-1.13086510e+00j
-1.19910772+7.07531761e-01j -2.32710969-5.61789128e-03j
0.73545411+9.97281197e-02j 1.69222275+2.79209942e+00j
-0.64002249+4.74617962e-01j 1.04457244-4.10140955e-01j
1.15767931+4.19833026e-01j 0.84757574+9.04367064e-02j
1.38185633-4.30468999e-01j 1.00956915+9.28305370e-01j
-1.88195028+1.76174914e+00j -0.02415628+1.15037785e-01j
2.06612071+9.42402098e-01j -0.31864447-5.43694019e-01j
0.54262 +4.36385849e-01j 0.37133839-8.76977245e-02j
1.4502856 -8.88133882e-01j 0.93364554-1.14273125e+00j
-0.23111325+1.69789108e-03j -0.21134427-5.75248503e-01j
0.83149602-1.55144426e-01j 0.1651417 -2.82516236e-01j
-0.71817049-9.64550149e-01j -0.28565927-1.41041215e-01j
0.10536078-2.97629183e-01j 0.70723192-1.62133244e+00j
0.2368845 -4.12183351e-01j]

```

```
# Hitung Discrete Fourier Transform (DFT)
```

```
X5 = np.fft.fft(x5)
```

```
print(X5)
```

```

[ 0.29457149+0.00000000e+00j 0.28264143+1.98614747e-01j
 0.20654234+1.87186916e-01j 0.17379952-3.10957676e-01j
-0.12966735+1.81103471e-01j 0.11325505-2.99068094e-02j
-0.04727311-2.61390180e-01j -0.06430084+1.00189299e-01j
0.064168 +1.01849833e-01j -0.10076521+1.10798648e-01j
-0.23147158-9.87353175e-02j -0.24020404+9.01863160e-02j
0.07624522-3.34850561e-02j 0.01001794+3.82452662e-02j
0.14946062+7.10194703e-02j 0.06629879-5.10215648e-01j
-0.11039441-2.50751971e-01j -0.25616817+2.14119802e-01j
0.14806122-2.90943240e-01j 0.16273735+1.83555892e-02j
0.26795915-9.58705259e-02j -0.1144562 -3.37718693e-01j
-0.31999757+1.39840565e-01j -0.30398676+1.54691795e-01j
-0.28284323-3.30594887e-02j 0.07869374-8.74272062e-02j
0.04239651-1.79988487e-01j -0.06678243+6.17458525e-02j
-0.08925574-5.48649599e-02j 0.12933952+9.96471810e-02j
0.00661354+3.52535604e-01j 0.11265969-1.80971957e-02j]

```

```

-0.1599287 -2.27736201e-01j -0.38769607+1.14124340e-01j
-0.10232764-4.30632098e-02j -0.39447537+1.46668909e-04j
-0.22221782+8.27768596e-02j -0.02411966+3.49827976e-01j
 0.08608773-9.16983855e-02j -0.02558706+1.25603819e-01j
 0.06214995+2.83303349e-01j -0.20566211+1.13207164e-01j
 0.07729781+1.23228488e-01j -0.06493018-1.40518053e-01j
 0.04511156-1.11400971e-01j -0.27875242-1.48323688e-01j
-0.01911974+4.44345826e-02j -0.26768588+1.70931990e-01j
-0.20050614+4.30486649e-01j  0.13266482-2.46063177e-01j
 0.07867587-2.46865103e-01j -0.17925441-1.16466129e-01j
 0.52568834+5.00562385e-01j  0.1636913 +1.32835152e-01j
 0.20724846+1.41446835e-01j -0.00134543+7.40055901e-03j
 0.31853584-1.15272368e-01j -0.20722022-1.54870321e-01j
-0.17757481+1.53398407e-01j  0.15150905+3.31682526e-01j
 0.08362204-2.74233263e-01j -0.14845667+9.28588574e-02j
-0.2120911 +4.37132571e-01j  0.05855441+2.97480673e-01j
-0.30529726-4.85617102e-02j -0.30529726+4.85617102e-02j
 0.05855441-2.97480673e-01j -0.2120911 -4.37132571e-01j
-0.14845667-9.28588574e-02j  0.08362204+2.74233263e-01j
 0.15150905-3.31682526e-01j -0.17757481-1.53398407e-01j
-0.20722022+1.54870321e-01j  0.31853584+1.15272368e-01j
-0.00134543-7.40055901e-03j  0.20724846-1.41446835e-01j
 0.1636913 -1.32835152e-01j  0.52568834-5.00562385e-01j
-0.17925441+1.16466129e-01j  0.07867587+2.46865103e-01j
 0.13266482+2.46063177e-01j -0.20050614-4.30486649e-01j
-0.26768588-1.70931990e-01j -0.01911974-4.44345826e-02j
-0.27875242+1.48323688e-01j  0.04511156+1.11400971e-01j
-0.06493018+1.40518053e-01j  0.07729781-1.23228488e-01j
-0.20566211-1.13207164e-01j  0.06214995-2.83303349e-01j
-0.02558706-1.25603819e-01j  0.08608773+9.16983855e-02j
-0.02411966-3.49827976e-01j -0.22221782-8.27768596e-02j
-0.39447537-1.46668909e-04j -0.10232764+4.30632098e-02j
-0.38769607-1.14124340e-01j -0.1599287 +2.27736201e-01j
 0.11265969+1.80971957e-02j  0.00661354-3.52535604e-01j
 0.12933952-9.96471810e-02j -0.08925574+5.48649599e-02j
-0.06678243-6.17458525e-02j  0.04239651+1.79988487e-01j
 0.07869374+8.74272062e-02j -0.28284323+3.30594887e-02j
-0.30398676-1.54691795e-01j -0.31999757-1.39840565e-01j
-0.1144562 +3.37718693e-01j  0.26795915+9.58705259e-02j
 0.16273735-1.83555892e-02j  0.14806122+2.90943240e-01j
-0.25616817-2.14119802e-01j -0.11039441+2.50751971e-01j
 0.06629879+5.10215648e-01j  0.14946062-7.10194703e-02j
 0.01001794-3.82452662e-02j  0.07624522+3.34850561e-02j
-0.24020404-9.01863160e-02j -0.23147158+9.87353175e-02j
-0.10076521-1.10798648e-01j  0.064168 -1.01849833e-01j
-0.06430084-1.00189299e-01j -0.04727311+2.61390180e-01j
 0.11325505+2.99068094e-02j -0.12966735-1.81103471e-01j
 0.17379952+3.10957676e-01j  0.20654234-1.87186916e-01j
 0.28264143-1.98614747e-01j]

```

# Hitung Discrete Fourier Transform (DFT)

```

X6 = np.fft.fft(x6)
print(X6)
[-6.33888440e-02+0.00000000e+00j -1.15464241e-01-3.42669365e-03j
 1.52279786e-01+2.02599327e-01j 3.32434753e-01+1.39925376e-01j
-1.01411595e-02+2.63307804e-04j -2.54975343e-01-2.04046756e-01j
-3.79567481e-01-4.82800404e-02j -2.34466466e-01+1.51384077e-01j
 1.21050991e-03-2.51912827e-03j -1.05982400e-01+8.30507444e-02j
 1.88368909e-01+6.03744028e-02j -3.06051292e-01-1.63823178e-01j
 9.57640493e-03-6.06728853e-02j -8.86537409e-02+1.44977558e-01j
-1.34357608e-01-9.64560214e-02j -1.53569933e-01-5.86267746e-02j
-2.39193493e-01-1.01443160e-01j -2.00807342e-01+1.80827680e-01j
 2.25377689e-01-1.07147741e-01j 1.58929173e-01-7.77478338e-02j
 9.33369606e-02-4.34122988e-02j -6.79798931e-05-8.73835521e-02j
 1.97255819e-01+6.01053961e-02j -1.57453829e-01+4.99440166e-02j
-1.44308074e-01+1.69178318e-01j -3.10742612e-02-7.96668248e-02j
 6.54375684e-02-1.17537799e-01j -2.88807326e-01-6.42264799e-02j
-3.32394502e-02-1.86648552e-01j 6.02563387e-02+2.70130919e-01j
 2.32540370e-01+1.99412604e-01j 2.47646326e-03-8.36031629e-02j
-1.05178031e-01-2.15571292e-01j -7.02510310e-02+1.62070884e-01j
 7.79206396e-02+3.99810364e-02j -1.48367541e-01-1.27070359e-01j
-7.33964811e-04+8.91720807e-02j -2.62227639e-01+1.55465602e-01j
 7.69287043e-02+1.54877574e-01j -8.03138402e-02-9.05141969e-02j
-2.80987261e-01-6.78780707e-02j -6.00047658e-02+1.42499679e-01j
 4.88122766e-02-6.14372798e-02j -6.18234101e-02-4.25399755e-02j
-2.82766207e-02+2.36660606e-01j -1.65338322e-01-1.84573279e-01j
-6.13838843e-02+4.39458401e-02j -2.48012637e-01+5.22865060e-02j
-2.19188974e-02+1.57500537e-01j -9.82446683e-03+2.29509732e-02j
 2.00275537e-01-1.02356409e-01j 4.82459475e-03+1.33700736e-01j
 1.69408717e-01+2.36527676e-01j -5.45992719e-02+2.96967963e-01j
 4.99342625e-02+9.30243585e-02j 2.49878625e-01-8.78128341e-02j
 1.49100327e-01+6.09480129e-02j -2.08917569e-01-1.62339606e-01j
 5.96682262e-02+1.02746513e-01j 2.10706707e-02+2.70874748e-01j
-1.17936587e-01-2.89156846e-02j 2.80345606e-02-1.60911861e-01j
-4.86420573e-01+5.64631262e-02j -1.86391165e-02+4.00061537e-01j
 1.26991113e-01+1.93702941e-01j 1.26991113e-01-1.93702941e-01j
-1.86391165e-02-4.00061537e-01j -4.86420573e-01-5.64631262e-02j
 2.80345606e-02+1.60911861e-01j -1.17936587e-01+2.89156846e-02j
 2.10706707e-02-2.70874748e-01j 5.96682262e-02-1.02746513e-01j
-2.08917569e-01+1.62339606e-01j 1.49100327e-01-6.09480129e-02j
 2.49878625e-01+8.78128341e-02j 4.99342625e-02-9.30243585e-02j
-5.45992719e-02-2.96967963e-01j 1.69408717e-01-2.36527676e-01j
 4.82459475e-03-1.33700736e-01j 2.00275537e-01+1.02356409e-01j
-9.82446683e-03-2.29509732e-02j -2.19188974e-02-1.57500537e-01j
-2.48012637e-01-5.22865060e-02j -6.13838843e-02-4.39458401e-02j
-1.65338322e-01+1.84573279e-01j -2.82766207e-02-2.36660606e-01j
-6.18234101e-02+6.25399755e-02j 4.88122766e-02+6.14372798e-02j
-6.00047658e-02-1.42499679e-01j -2.80987261e-01+6.78780707e-02j
-8.03138402e-02+9.05141969e-02j 7.69287043e-02-1.54877574e-01j
-2.62227639e-01-1.55465602e-01j -7.33964811e-04-8.91720807e-02j
-1.48367541e-01+1.27070359e-01j 7.79206396e-02-3.99810364e-02j
-7.02510310e-02-1.62070884e-01j -1.05178031e-01+2.15571292e-01j

```

```

2.47646326e-03+8.36031629e-02j 2.32540370e-01-1.99412604e-01j
6.02563387e-02-2.70130919e-01j -3.32394502e-02+1.86648552e-01j
-2.88807326e-01+6.42264799e-02j 6.54375684e-02+1.17537799e-01j
-3.10742612e-02+7.96668248e-02j -1.44308074e-01-1.69178318e-01j
-1.57453829e-01-4.99440166e-02j 1.97255819e-01-6.01053961e-02j
-6.79798931e-05+8.73835521e-02j 9.33369606e-02+4.34122988e-02j
1.58929173e-01+7.77478338e-02j 2.25377689e-01+1.07147741e-01j
-2.00807342e-01-1.80827680e-01j -2.39193493e-01+1.01443160e-01j
-1.53569933e-01+5.86267746e-02j -1.34357608e-01+9.64560214e-02j
-8.86537409e-02-1.44977558e-01j 9.57640493e-03+6.06728853e-02j
-3.06051292e-01+1.63823178e-01j 1.88368909e-01-6.03744028e-02j
-1.05982400e-01-8.30507444e-02j 1.21050991e-03+2.51912827e-03j
-2.34466466e-01-1.51384077e-01j -3.79567481e-01+4.82800404e-02j
-2.54975343e-01+2.04046756e-01j -1.01411595e-02-2.63307804e-04j
3.32434753e-01-1.39925376e-01j 1.52279786e-01-2.02599327e-01j
-1.15464241e-01+3.42669365e-03j]

```

```

e1=sum(X1)/N
ze1=X1-e1
X1k=np.conj(X1)
ze1k=np.conj(ze1)
X1ot=ze1*ze1k
eX1=sum(X1ot)/N
z1=complex(round(eX1.real,12),round(eX1.imag,12))

```

```

e2=sum(X2)/N
ze2=X2-e2
X2k=np.conj(X2)
ze2k=np.conj(ze2)
X2ot=ze2*ze2k
eX2=sum(X2ot)/N
z2=complex(round(eX2.real,12),round(eX2.imag,12))

```

```

e3=sum(X3)/N
ze3=X3-e3
X3k=np.conj(X3)
ze3k=np.conj(ze3)
X3ot=ze3*ze3k
eX3=sum(X3ot)/N
z3=complex(round(eX3.real,12),round(eX3.imag,12))

```

```

e4=sum(X4)/N
ze4=X4-e4
X4k=np.conj(X4)
ze4k=np.conj(ze4)

```

```

X4ot=ze4*ze4k
eX4=sum(X4ot)/N
z4=complex(round(eX4.real,12),round(eX4.imag,12))

e5=sum(X5)/N
ze5=X5-e5
X5k=np.conj(X5)
ze5k=np.conj(ze5)
X5ot=ze5*ze5k
eX5=sum(X5ot)/N
z5=complex(round(eX5.real,12),round(eX5.imag,12))

e6=sum(X6)/N
ze6=X6-e6
X6k=np.conj(X6)
ze6k=np.conj(ze6)
X6ot=ze6*ze6k
eX6=sum(X6ot)/N
z6=complex(round(eX6.real,12),round(eX6.imag,12))

#membentuk matriks non diagonal
z7tot=ze1*ze2
ez7t=sum(z7tot)/N
z7=complex(round(ez7t.real,12),round(ez7t.imag,12))

z12tot=ze2*ze1
ez12t=sum(z12tot)/N
z12=complex(round(ez12t.real,12),round(ez12t.imag,12))

z8tot=ze1*ze3
ez8t=sum(z8tot)/N
z8=complex(round(ez8t.real,12),round(ez8t.imag,12))

z17tot=ze3*ze1
ez17t=sum(z17tot)/N
z17=complex(round(ez17t.real,12),round(ez17t.imag,12))

z9tot=ze1*ze4
ez9t=sum(z9tot)/N
z9=complex(round(ez9t.real,12),round(ez9t.imag,12))

```

```

z22tot=ze4*ze1
ez22t=sum(z22tot)/N
z22=complex(round(ez22t.real,12),round(ez22t.imag,12))

z10tot=ze1*ze5
ez10t=sum(z10tot)/N
z10=complex(round(ez10t.real,12),round(ez10t.imag,12))

z27tot=ze5*ze1
ez27t=sum(z27tot)/N
z27=complex(round(ez27t.real,12),round(ez27t.imag,12))

z11tot=ze1*ze6
ez11t=sum(z11tot)/N
z11=complex(round(ez11t.real,12),round(ez11t.imag,12))

z32tot=ze6*ze1
ez32t=sum(z32tot)/N
z32=complex(round(ez32t.real,12),round(ez32t.imag,12))

z13tot=ze2*ze3
ez13t=sum(z13tot)/N
z13=complex(round(ez13t.real,12),round(ez13t.imag,12))

z18tot=ze3*ze2
ez18t=sum(z18tot)/N
z18=complex(round(ez18t.real,12),round(ez18t.imag,12))

z14tot=ze2*ze4
ez14t=sum(z14tot)/N
z14=complex(round(ez14t.real,12),round(ez14t.imag,12))

z23tot=ze4*ze2
ez23t=sum(z23tot)/N
z23=complex(round(ez23t.real,12),round(ez23t.imag,12))

z15tot=ze2*ze5
ez15t=sum(z15tot)/N
z15=complex(round(ez15t.real,12),round(ez15t.imag,12))

z28tot=ze5*ze2

```

```

ez28t=sum(z28tot)/N
z28=complex(round(ez28t.real,12),round(ez28t.imag,12))

z16tot=ze2*ze6
ez16t=sum(z16tot)/N
z16=complex(round(ez16t.real,12),round(ez16t.imag,12))

z33tot=ze6*ze2
ez33t=sum(z33tot)/N
z33=complex(round(ez33t.real,12),round(ez33t.imag,12))

z19tot=ze3*ze4
ez19t=sum(z19tot)/N
z19=complex(round(ez19t.real,12),round(ez19t.imag,12))

z24tot=ze4*ze3
ez24t=sum(z24tot)/N
z24=complex(round(ez24t.real,12),round(ez24t.imag,12))

z20tot=ze3*ze5
ez20t=sum(z20tot)/N
z20=complex(round(ez20t.real,12),round(ez20t.imag,12))

z29tot=ze5*ze3
ez29t=sum(z29tot)/N
z29=complex(round(ez29t.real,12),round(ez29t.imag,12))

z21tot=ze3*ze6
ez21t=sum(z21tot)/N
z21=complex(round(ez21t.real,12),round(ez21t.imag,12))

z34tot=ze6*ze3
ez34t=sum(z34tot)/N
z34=complex(round(ez34t.real,12),round(ez34t.imag,12))

z25tot=ze4*ze5
ez25t=sum(z25tot)/N
z25=complex(round(ez25t.real,12),round(ez25t.imag,12))

z30tot=ze5*ze4
ez30t=sum(z30tot)/N

```



```

z30=complex(round(ez30t.real,12),round(ez30t.imag,12))

z26tot=ze4*ze6
ez26t=sum(z26tot)/N
z26=complex(round(ez26t.real,12),round(ez26t.imag,12))

z35tot=ze6*ze4
ez35t=sum(z35tot)/N
z35=complex(round(ez35t.real,12),round(ez35t.imag,12))

z31tot=ze5*ze6
ez31t=sum(z31tot)/N
z31=complex(round(ez31t.real,12),round(ez31t.imag,12))

z36tot=ze6*ze5
ez36t=sum(z36tot)/N
z36=complex(round(ez36t.real,12),round(ez36t.imag,12))

#matriks varians-kovarians dan menghitung invers
matrix=np.matrix([[z1,z7,z8,z9,z10,z11],[z12,z2,z13,z14,z15,z16],
[z17,z18,z3,z19,z20,z21],[z22,z23,z24,z4,z25,z26],[z27,z28,z29,z3
0,z5,z31],[z32,z33,z34,z35,z36,z6]])
invers=np.linalg.inv(matrix)
print(invers)
[[ 3.76296663e-01-0.j  8.65350638e-02+0.j -1.81472368e-02+0.j
  9.02864199e-02-0.j -9.80341992e-02+0.j -2.29069389e-01+0.j]
 [ 8.65350638e-02+0.j  2.54199284e+00-0.j  3.36309105e-01-0.j
  5.26005780e-02+0.j  1.11559249e+00+0.j  1.38273304e-02+0.j]
 [-1.81472368e-02+0.j  3.36309105e-01-0.j  1.93256242e+00-0.j
  4.62829824e-03+0.j  5.91267384e-01+0.j  4.39979489e-01+0.j]
 [ 9.02864199e-02-0.j  5.26005780e-02-0.j  4.62829824e-03+0.j
  5.41982830e-01-0.j -1.20579524e-02+0.j -2.65675411e-01+0.j]
 [-9.80341992e-02+0.j  1.11559249e+00+0.j  5.91267384e-01+0.j
 -1.20579524e-02+0.j  1.36013039e+01+0.j  1.19723609e+00+0.j]
 [-2.29069389e-01+0.j  1.38273304e-02+0.j  4.39979489e-01+0.j
 -2.65675411e-01+0.j  1.19723609e+00+0.j  2.16817676e+01+0.j]]

#membuat vector
vector = np.array([1, 1, 1, 1, 1, 1]).reshape(-1, 1)
vt=np.transpose(vector)
#membuat vektor Sigma^2
sig=np.array([z1,z2,z3,z4,z5,z6]).reshape(-1, 1)
# menghitung bobot mula mula lamda=0
iv = invers @ vector
vi = vt @ invers

```

```

viv = vi @ vector
imu = invers @ sig
vim = vi @ sig
ipv = vim / viv
bot = imu - (iv * ipv)
bobot1 = iv / viv
print("bobot minimum", bobot1)
bobot minimum [[0.00439592+0.j]
 [0.0876965 +0.j]
 [0.06950402+0.j]
 [0.00870788+0.j]
 [0.34672308+0.j]
 [0.4829726 +0.j]]
# Menghitung Bobot Complex
# Inisialisasi bobot awal
bobot = bobot1

# Menghitung dan mencetak total sum untuk bobot awal
total_sum = np.sum(bobot)
print(f"Inisial bobot: {bobot}")
print(f"Total sum: {total_sum}")
print("\n")

# Inisialisasi variabel untuk menyimpan bobot terakhir yang valid
bobot_optimal = None

# Iterasi untuk nilai t dari 0.01 hingga 0.09
for t in np.arange(0.01, 0.1, 0.01):
    # Menghitung bobot untuk nilai t saat ini
    bobot = bobot + (t * bot)

    # Pengecekan apakah ada elemen negatif dalam bobot
    if any((i < 0).any() for i in bobot):
        print("Ditemukan bobot minus. Iterasi dihentikan.")
        break

    # Menyimpan bobot saat ini sebagai bobot terakhir yang valid
    bobot_optimal = bobot

# Menghitung dan mencetak total sum untuk bobot saat ini
total_sum = np.sum(bobot)
print(f"Bobot untuk t={t}: {bobot}")

```

```

    print(f"Total sum untuk t={t}: {total_sum}")
    print("\n")

# Menggunakan bobot terakhir yang valid sebagai hasil
hasil_akhir = bobot_optimal
print(f"Bobot akhir yang tidak negatif: {hasil_akhir}")
Inisial bobot: [[0.00439592+0.j]
 [0.0876965 +0.j]
 [0.06950402+0.j]
 [0.00870788+0.j]
 [0.34672308+0.j]
 [0.4829726 +0.j]]
Total sum: (0.9999999999999999+0j)

Bobot untuk t=0.01: [[0.01647736+0.j]
 [0.09815978+0.j]
 [0.07650369+0.j]
 [0.02119408+0.j]
 [0.33763054+0.j]
 [0.45003455+0.j]]
Total sum untuk t=0.01: (1+0j)

Bobot untuk t=0.02: [[0.04064025+0.j]
 [0.11908633+0.j]
 [0.09050301+0.j]
 [0.0461665 +0.j]
 [0.31944547+0.j]
 [0.38415844+0.j]]
Total sum untuk t=0.02: (1+0j)

Bobot untuk t=0.03: [[0.07688458+0.j]
 [0.15047616+0.j]
 [0.11150199+0.j]
 [0.08362512+0.j]
 [0.29216787+0.j]
 [0.28534428+0.j]]
Total sum untuk t=0.03: (1+0j)

Bobot untuk t=0.04: [[0.12521036+0.j]
 [0.19232927+0.j]
 [0.13950063+0.j]
 [0.13356994+0.j]
 [0.25579773+0.j]
 [0.15359206+0.j]]
Total sum untuk t=0.04: (0.9999999999999999+0j)

```

```

Ditemukan bobot minus. Iterasi dihentikan.
Bobot akhir yang tidak negatif: [[0.12521036+0.j]
 [0.19232927+0.j]
 [0.13950063+0.j]
 [0.13356994+0.j]
 [0.25579773+0.j]
 [0.15359206+0.j]]
# menghitung portofolio
z1w = x1 * hasil_akhir[0, 0]
z2w = x2 * hasil_akhir[1, 0]
z3w = x3 * hasil_akhir[2, 0]
z4w = x4 * hasil_akhir[3, 0]
z5w = x5 * hasil_akhir[4, 0]
z6w = x6 * hasil_akhir[5, 0]
zw=z1w+z2w+z3w+z4w+z5w+z6w
print(zw)
0      0.014082+0.000000j
1      0.034597+0.000000j
2      0.025655+0.000000j
3      0.016399+0.000000j
4      -0.060185+0.000000j
...
124    0.055560+0.000000j
125    0.023666+0.000000j
126    -0.006536+0.000000j
127    -0.012829+0.000000j
128    0.035853+0.000000j
Length: 129, dtype: complex128
# menghitung standar deviasi
std_dev_reall = np.std(np.real(zw))
std_dev_imagl = np.std(np.imag(zw))

print("Standar Deviasi pada Bagian Real:", std_dev_reall)
print("Standar Deviasi pada Bagian Imajiner:", std_dev_imagl)

# menghitung Value at Risk
VaR95=std_dev_reall*(1.65)*(1000000)
print("Nilai VaR 95%:", VaR95)
Standar Deviasi pada Bagian Real: 0.03506185096282436
Standar Deviasi pada Bagian Imajiner: 0.0
Nilai VaR 95%: 57852.05408866019

```