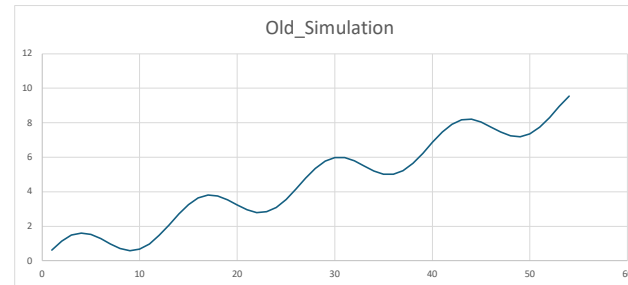


Index	Old_Simulation
1	0.625063471
2	1.148131491
3	1.489903076
4	1.611416557
5	1.52271873
6	1.2806294
7	0.976098704
8	0.713970279
9	0.589653056
10	0.667892323
11	0.968363593
12	1.461294712
13	2.074090298
14	2.707484564
15	3.257628415
16	3.63919637
17	3.804371877
18	3.753486727
19	3.534947226
20	3.234463562
21	2.955978889
22	2.79854148
23	2.83426602
24	3.092287752
25	3.552279741
26	4.148975719
27	4.786692137
28	5.360618201
29	5.780136775
30	5.988987117
31	5.977783159
32	5.786101954
33	5.493677596
34	5.202660002
35	5.01488608
36	5.009220534
37	5.224008093
38	5.648541632
39	6.225444555
40	6.863437361
41	7.457647758
42	7.912945251
43	8.165108118
44	8.195113052
45	8.033367932
46	7.752945931
47	7.453326387
48	7.238260181
49	7.192684679
50	7.363824764
51	7.75068508
52	8.3042715
53	8.938492555
54	9.549315187



## NOTE

In part (a) of Question 1.

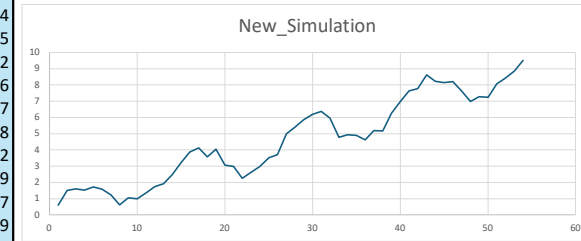
We use the function from the previous week,  $\sin(t/2.1) + t/6$  to simulate the original dataset.

Meanwhile expanding it to 54 data points.

Index	Old_Simulation	Error	New_Simulation	Index	Old_Simulation	Error	New_Simulation
1	0.625063471	0.652469202	1.277532674	1	0.625063471	-0.024975593	0.600087879
2	1.148131491	-0.201234257	0.946897234	2	1.148131491	0.356922342	1.505053833
3	1.489903076	0.119362548	1.609265625	3	1.489903076	0.121491861	1.611394937
4	1.611416557	-0.008504338	1.602912219	4	1.611416557	-0.082406732	1.529009825
5	1.52271873	0.102931114	1.625649844	5	1.52271873	0.194915771	1.717634501
6	1.2806294	0.150149363	1.430778763	6	1.2806294	0.288729004	1.569358404
7	0.976098704	0.367769803	1.343868507	7	0.976098704	0.23724216	1.213340864
8	0.713970279	-0.043110418	0.670859861	8	0.713970279	-0.086464094	0.627506185
9	0.589653056	0.093194566	0.682847622	9	0.589653056	0.456910347	1.046563402
10	0.667892323	-0.187502549	0.480389773	10	0.667892323	0.308810114	0.976702436
11	0.968363593	-0.226354438	0.742009155	11	0.968363593	0.365815344	1.334178937
12	1.461294712	0.080818707	1.542113419	12	1.461294712	0.271399016	1.732693728
13	2.074090298	0.676798175	2.750888473	13	2.074090298	-0.163811016	1.910279282
14	2.707484564	0.087415409	2.794899973	14	2.707484564	-0.232228855	2.475255709
15	3.257628415	0.01481063	3.272439045	15	3.257628415	-0.037888245	3.21974017
16	3.63919637	-0.0666857	3.57251067	16	3.63919637	0.219211008	3.858407379
17	3.804371877	-0.072248271	3.732123606	17	3.804371877	0.330006428	4.134378305
18	3.753486727	0.19052385	3.944010578	18	3.753486727	-0.189742538	3.56374419
19	3.534947226	-0.294438352	3.240508874	19	3.534947226	0.515806131	4.050753357
20	3.234463562	-0.130413221	3.104050341	20	3.234463562	-0.171095528	3.063368034
21	2.955978889	-0.195955138	2.760023751	21	2.955978889	0.036488013	2.992466902
22	2.79854148	-0.108362862	2.690178618	22	2.79854148	-0.540018844	2.258522636
23	2.83426602	0.176867224	3.011133245	23	2.83426602	-0.197420253	2.636845767
24	3.092287752	0.007112401	3.099400153	24	3.092287752	-0.113082418	2.979205334
25	3.552279741	-0.325155393	3.227124347	25	3.552279741	-0.046048656	3.506231084
26	4.148975719	0.043735335	4.192711054	26	4.148975719	-0.438268392	3.710707327
27	4.786692137	0.235181689	5.021873825	27	4.786692137	0.207700312	4.994392448
28	5.360618201	0.072859245	5.433477447	28	5.360618201	0.044154731	5.404772932
29	5.780136775	-0.145467617	5.634669159	29	5.780136775	0.077210306	5.857347081
30	5.988987117	0.207311445	6.196298562	30	5.988987117	0.202891579	6.191878695
31	5.977783159	-0.109132063	5.868651096	31	5.977783159	0.393638614	6.371421773
32	5.786101954	-0.203073476	5.583028479	32	5.786101954	0.166786372	5.952888327
33	5.493677596	-0.116111816	5.37756578	33	5.493677596	-0.722024679	4.771652916
34	5.202660002	0.085743015	5.288403017	34	5.202660002	-0.262956814	4.939703188
35	5.01488608	0.182553321	5.197439401	35	5.01488608	-0.118051144	4.896834936
36	5.009220534	0.288211037	5.297431571	36	5.009220534	-0.392675475	4.616545059
37	5.224008093	-0.124903925	5.099104168	37	5.224008093	-0.041177742	5.182830351
38	5.648541632	0.052881444	5.701423076	38	5.648541632	-0.47190634	5.176635291
39	6.225444555	0.258059013	6.483503568	39	6.225444555	0.025725964	6.251170519
40	6.863437361	0.465436586	7.328873947	40	6.863437361	0.130195916	6.993633277
41	7.457647758	-0.049725799	7.407921959	41	7.457647758	0.164658334	7.622306093
42	7.912945251	0.069781471	7.84316378	42	7.912945251	-0.133102219	7.779843031
43	8.165108118	0.509731523	8.67483964	43	8.165108118	0.463529664	8.628637782
44	8.195113052	0.193514205	8.388627256	44	8.195113052	0.030311752	8.225424803
45	8.033367932	-0.410680121	7.622687812	45	8.033367932	0.109494693	8.142862625
46	7.752945931	0.18767433	7.940620262	46	7.752945931	0.460453792	8.213399724
47	7.453326387	0.033579109	7.486905497	47	7.453326387	0.179406228	7.632732616
48	7.238260181	-0.100068992	7.13819119	48	7.238260181	-0.247563215	6.990696966
49	7.192684679	-0.117774859	7.07490982	49	7.192684679	0.090995989	7.283680668
50	7.363824764	0.352012498	7.715837262	50	7.363824764	-0.130341053	7.233483711
51	7.75068508	-0.083495953	7.667189127	51	7.75068508	0.322933144	8.073618224
52	8.3042715	0.105439575	8.409711076	52	8.3042715	0.121754925	8.426026426
53	8.938492555	0.013553796	8.952046351	53	8.938492555	-0.079034476	8.859458079
54	9.549315187	0.079746955	9.629062143	54	9.549315187	-0.039681534	9.509633654

#### NOTE

In order to prevent the values of the data from changing with each refresh, this table serves as an example.



#### NOTE

In part (b) of Question 1.

We use  $\text{NORM.INV}(\text{RAND}(), 0, 0.25)$  to generate disturbances that follow Gaussian Process with a expected value of 0 and a variance of 0.25.

The white block will refresh data automatically.

Thus, we have blue block to restore data.

Index

D_bar_6.5_to_42.5	D_bar_7_to_42	L+T
1		0.300048375
2		0.460868253
3		0.621688131
4		0.782508009
5		0.943327887
6		1.104147766
7	1.288627078	1.343218386
8	1.397809695	1.438234773
9	1.478659851	1.545674236
10	1.61268862	1.709746852
11	1.806805083	1.907502742
12	2.0082004	2.091299808
13	2.174399216	2.292624736
14	2.410850257	2.5123445
15	2.613838744	2.694918057
16	2.775997369	2.829406544
17	2.882815719	2.937093504
18	2.991371288	3.043309272
19	3.095247255	3.161745247
20	3.228243239	3.27972039
21	3.33119754	3.405141385
22	3.47908523	3.543517128
23	3.607949026	3.679739392
24	3.751529758	3.861035362
25	3.970540966	4.067235484
26	4.163930001	4.28432668
27	4.404723359	4.47885611
28	4.55298886	4.664704716
29	4.776420573	4.870586788
30	4.964753003	5.032975492
31	5.101197981	5.171056283
32	5.240914586	5.301994918
33	5.36307525	5.415441003
34	5.467806756	5.53400927
35	5.600211784	5.673751743
36	5.747291702	5.813456883
37	5.879622063	5.97367273
38	6.067723397	6.162412417
39	6.257101437	6.397568508
40	6.53803558	6.674439602
41	6.810843624	6.924839361
42	7.038835098	7.137758094
43	7.23668109	7.054483256
44		7.215303135
45		7.376123013
46		7.536942891
47		7.697762769
48		7.858582647

New\_Simulation

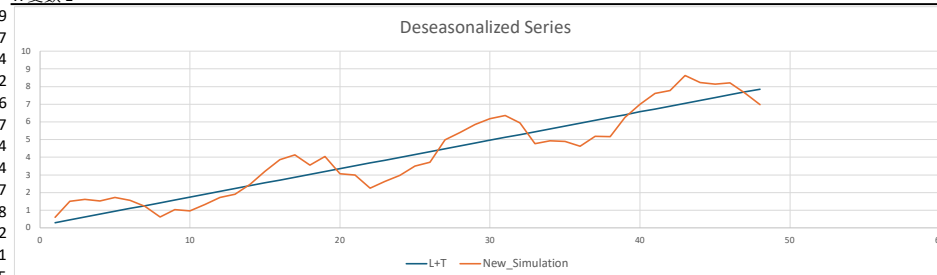
摘要輸出

0.600087879
1.505053833
1.611394937
1.529009825
1.717634501
1.569358404
1.213340864
0.627506185
1.046563402
0.976702436
1.334178937
1.732693728
1.910279282
2.475255709
3.21974017
3.858407379
4.134378305
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7.622306093
7.779843031
8.628637782
8.225424803
8.142862625
8.213399724
7.632732616
6.990696966

L =	0.139228497
T =	0.160819878

	自由度	SS	MS	F	顯著值
迴歸統計					
R 的倍數	1	100.477884	100.477884	10442.42524	6.68641E-44
R 平方	34	0.327150827	0.009622083		
調整的 R 平方	35	100.8050348			
標準誤					
觀察值個數					

	係數	標準誤	t 統計	P-值	下限 95%	上限 95%	下限 95.0%	上限 95.0%
截距	0.139228497	0.04188001	3.324461842	0.002130939	0.054118075	0.224338918	0.054118075	0.224338918
X 變數 1	0.160819878	0.001573762	102.1881854	6.68641E-44	0.157621609	0.164018147	0.157621609	0.164018147



## NOTE

In part (c) of Question 1.

D\_bar\_6.5\_to\_42.5 stands for the moving average of data.

For instance, D\_6.5 is the average of y\_1 to y\_12

and D\_7.5 is the average of y\_1 to y\_13, and so on.

To make analysis more precise, D\_bar\_7\_to\_42 is the moving average of D\_bar\_6.5\_to\_42.5, D\_7 means the average of D\_bar\_6.5 and D\_bar\_7.5.

After calculating D\_bar we need, we can get L and T by regression analysis.

In excel, we set X be D\_bar\_7\_to\_42(from 1.343218386 to 7.137758094) and Y be Index(from 7 to 42). We can estimate L and T.

Index	New_Simulation	L+T	Seasonality Factor	Seasonality Factor_Average
1	0.600087879	0.300048375	1.999970434	1.13766035
2	1.505053833	0.460868253	3.265692142	1.497030048
3	1.611394937	0.621688131	2.591966706	1.485843264
4	1.529009825	0.782508009	1.953986167	1.401236657
5	1.717634501	0.943327887	1.82082447	1.402853004
6	1.569358404	1.104147766	1.421330054	1.242972134
7	1.213340864	1.264967644	0.959187272	1.17338462
8	0.627506185	1.425787522	0.440111991	0.904822375
9	1.046563402	1.5866074	0.659623422	0.872673525
10	0.976702436	1.747427278	0.558937387	0.785961767
11	1.334178937	1.908247156	0.699164641	0.806678694
12	1.732693728	2.069067034	0.837427545	0.812666611
13	1.910279282	2.229886912	0.856670924	1.13766035
14	2.475255709	2.390706791	1.035365658	1.497030048
15	3.21974017	2.551526669	1.26188772	1.485843264
16	3.858407379	2.712346547	1.42253481	1.401236657
17	4.134378305	2.873166425	1.438962348	1.402853004
18	3.56374419	3.033986303	1.17460787	1.242972134
19	4.050753357	3.194806181	1.267918342	1.17338462
20	3.063368034	3.355626059	0.912905067	0.904822375
21	2.992466902	3.516445938	0.850991869	0.872673525
22	2.258522636	3.677265816	0.614185308	0.785961767
23	2.636845767	3.838085694	0.687021077	0.806678694
24	2.979205334	3.998905572	0.745005172	0.812666611
25	3.506231084	4.15972545	0.84289964	1.13766035
26	3.710707327	4.320545328	0.858851614	1.497030048
27	4.994392448	4.481365206	1.114480123	1.485843264
28	5.404772932	4.642185084	1.164273469	1.401236657
29	5.857347081	4.803004963	1.219517183	1.402853004
30	6.191878695	4.963824841	1.247400723	1.242972134
31	6.371421773	5.124644719	1.243290437	1.17338462
32	5.952888327	5.285464597	1.126275319	0.904822375
33	4.771652916	5.446284475	0.876129945	0.872673525
34	4.939703188	5.607104353	0.880972223	0.785961767
35	4.896834936	5.767924231	0.848976987	0.806678694
36	4.616545059	5.928744109	0.778671667	0.812666611
37	5.182830351	6.089563988	0.851100401	1.13766035
38	5.176635291	6.250383866	0.828210779	1.497030048
39	6.251170519	6.411203744	0.975038506	1.485843264
40	6.993633277	6.572023622	1.064152182	1.401236657
41	7.622306093	6.7328435	1.132108015	1.402853004
42	7.779843031	6.893663378	1.128549888	1.242972134
43	8.628637782	7.054483256	1.223142428	1.17338462
44	8.225424803	7.215303135	1.139997121	0.904822375
45	8.142862625	7.376123013	1.103948865	0.872673525
46	8.213399724	7.536942891	1.089752151	0.785961767
47	7.632732616	7.697762769	0.991552071	0.806678694
48	6.990696966	7.858582647	0.88956206	0.812666611

## NOTE

In part (d) of Question 1.

We use index multiply T and plus L(the formula is  $t * T + L$ ) to get a regression line(L+T) of dataset.

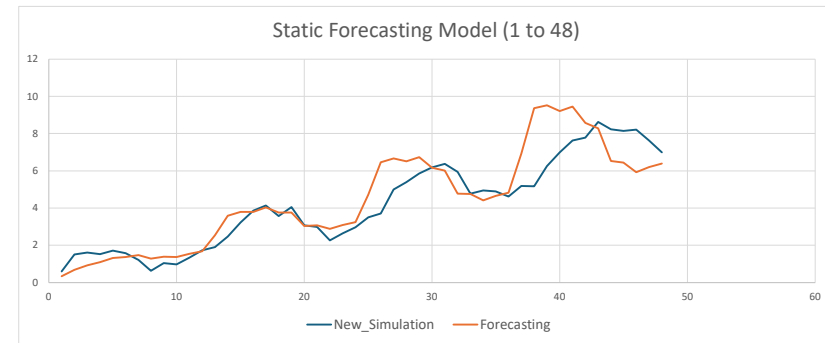
Furthermore, we still need to estimate  $S_t$  to make L+T fluctuate like dataset.

Hence, we use L+T divide simulation and get  $S_t$  of every data point.

We assume that there are 4 cycles in these 48 data points, so every cycle has 12 periods.

Taking the average of  $S_t$  in the same period can estimate the overall seasonality factor of that period.

Index	New_Simulation	Forecasting	E_t	E_t^2	E_t   /   D_t	MSE = 1.553701183 MAPE = 23.97%
1	0.600087879	0.341353139	-0.258734739	0.066943665	0.431161416	
2	1.505053833	0.689933623	-0.815120209	0.664420956	0.54158874	
3	1.611394937	0.923731122	-0.687663815	0.472881523	0.426750637	
4	1.529009825	1.096478907	-0.432530919	0.187082996	0.282883021	
5	1.717634501	1.323350361	-0.39428414	0.155459983	0.229550664	
6	1.569358404	1.372424904	-0.196933499	0.038782803	0.125486631	
7	1.213340864	1.484293578	0.270952714	0.073415373	0.223311291	
8	0.627506185	1.290084452	0.662578266	0.439009959	1.055891211	
9	1.046563402	1.384590273	0.33802687	0.114262165	0.322987475	
10	0.976702436	1.373411031	0.396708595	0.157377709	0.406171399	
11	1.334178937	1.539342324	0.205163386	0.042092015	0.153775015	
12	1.732693728	1.681461695	-0.051232033	0.002624721	0.029567853	
13	1.910279282	2.536853925	0.626574643	0.392595783	0.32800159	
14	2.475255709	3.578959902	1.103704193	1.218162945	0.44589502	
15	3.21974017	3.791168713	0.571428543	0.326530579	0.177476601	
16	3.858407379	3.800639407	-0.057767971	0.00337138	0.014971973	
17	4.134378305	4.030630151	-0.103748154	0.01076368	0.025094016	
18	3.56374419	3.771160429	0.20741624	0.043021496	0.058201776	
19	4.050753357	3.748736437	-0.30201692	0.09121422	0.07455821	
20	3.063368034	3.03624554	-0.027122494	0.00073563	0.008853815	
21	2.992466902	3.068709272	0.07624237	0.005812899	0.0254781	
22	2.258522636	2.890190339	0.631667702	0.399004086	0.279681812	
23	2.636845767	3.096101955	0.459256188	0.210916246	0.174168771	
24	2.979205334	3.249777039	0.270571705	0.073209048	0.090820093	
25	3.506231084	4.73235471	1.226123626	1.503379146	0.349698464	
26	3.710707327	6.467986181	2.757278854	7.602586679	0.74306018	
27	4.994392448	6.658606304	1.664213855	2.769607757	0.333216477	
28	5.404772932	6.504799908	1.100026975	1.210059347	0.203528805	
29	5.857347081	6.737909941	0.880562859	0.775390949	0.150334759	
30	6.191878695	6.169895954	-0.021982741	0.000483241	0.003550254	
31	6.371421773	6.013179296	-0.358242477	0.128337672	0.056226458	
32	5.952888327	4.782406629	-1.170481698	1.370027405	0.196624165	
33	4.771652916	4.752828272	-0.018824644	0.000354367	0.003945099	
34	4.939703188	4.406969646	-0.532733543	0.283805028	0.107847278	
35	4.896834936	4.652861586	-0.24397335	0.059522995	0.049822662	
36	4.616545059	4.818092383	0.201547324	0.040621324	0.04365761	
37	5.182830351	6.927855496	1.745025145	3.045112757	0.336693472	
38	5.176635291	9.357012459	4.180377168	17.47555327	0.807547168	
39	6.251170519	9.526043895	3.274873376	10.72479563	0.523881626	
40	6.993633277	9.208960408	2.215327132	4.907674301	0.316763411	
41	7.622306093	9.445189731	1.822883638	3.322904757	0.239151199	
42	7.779843031	8.568631479	0.788788448	0.622187215	0.101388736	
43	8.628637782	8.277622155	-0.351015627	0.12321197	0.040680306	
44	8.225424803	6.528567717	-1.696857086	2.87932397	0.206294158	
45	8.142862625	6.436947272	-1.705915353	2.910147193	0.209498236	
46	8.213399724	5.923748953	-2.289650771	5.242500653	0.278770162	
47	7.632732616	6.209621217	-1.423111399	2.025246053	0.186448481	
48	6.990696966	6.386407727	-0.604289239	0.365165485	0.086441916	



## NOTE

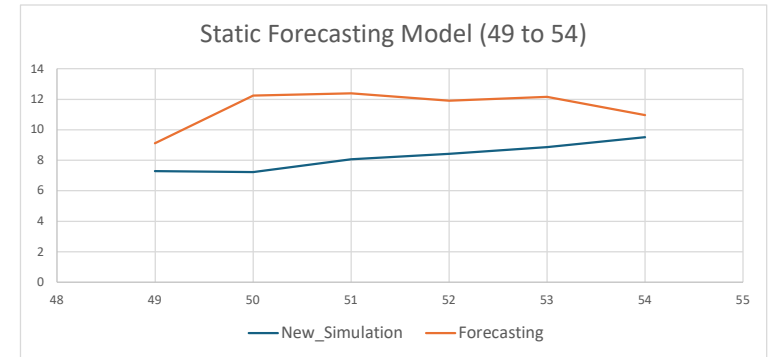
In part (e) of Question 1.

We use index multiply T and plus L, then multiply  $S_t$  to estimate the data.

The formula is  $(t * T + L) * S_t$ .

$E_t$  is the gap between data and forecasting which is used to estimate MSE and MAPE.

Index	New_Simulation	Forecasting	E_t	E_t^2	E_t   /   D_t	MSE = 12.05001195 MAPE = 40.32%
49	7.283680668	9.123356281	1.839675614	3.384406363	0.252574996	
50	7.233483711	12.24603874	5.012555027	25.1257079	0.692965551	
51	8.073618224	12.39348149	4.319863262	18.6612186	0.535059145	
52	8.426026426	11.91312091	3.487094483	12.15982793	0.413848035	
53	8.859458079	12.15246952	3.293011441	10.84392435	0.371694455	
54	9.509633654	10.967367	1.457733351	2.124986522	0.153290169	



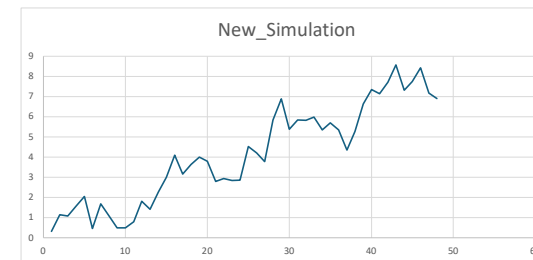
## NOTE

In part (f) of Question 1.

We use  $y_{49}$  to  $y_{54}$  which is generated before.

By the same parameters and formula to forecast and make comparison through MSE and MAPE like 1.(e).

Index	Old_Simulation	Error	New_Simulation	Index	Old_Simulation	Error	New_Simulation
1	0.625063471	-0.868799916	-0.243736445	1	0.625063471	-0.305198719	0.319864752
2	1.148131491	-0.061424298	1.086707193	2	1.148131491	-0.00367526	1.144456231
3	1.489903076	-0.387468866	1.10243421	3	1.489903076	-0.401236675	1.088666401
4	1.611416557	-0.711312404	0.900104153	4	1.611416557	-0.044286505	1.567130052
5	1.52271873	0.345721248	1.868439978	5	1.52271873	0.526316379	2.049035109
6	1.2806294	-0.305009062	0.975620338	6	1.2806294	-0.821129346	0.459500054
7	0.976098704	-0.268690383	0.707408321	7	0.976098704	0.712768912	1.688867616
8	0.713970279	-0.444204409	0.26976587	8	0.713970279	0.369987099	1.083957378
9	0.589653056	0.49023771	1.079890766	9	0.589653056	-0.096422523	0.493230532
10	0.667892323	-0.42028248	0.247609843	10	0.667892323	-0.17727189	0.490620433
11	0.968363593	-0.003168024	0.965195569	11	0.968363593	-0.185307164	0.783056429
12	1.461294712	-0.01637724	1.444917472	12	1.461294712	0.352926935	1.814221646
13	2.074090298	0.007031627	2.081121925	13	2.074090298	-0.667750728	1.406339571
14	2.707484564	0.454088343	3.161572907	14	2.707484564	-0.436830581	2.270653983
15	3.257628415	0.132335657	3.389964073	15	3.257628415	-0.251284636	3.006343779
16	3.63919637	-0.134787975	3.504408395	16	3.63919637	0.454302583	4.093498953
17	3.804371877	0.161151534	3.965523412	17	3.804371877	-0.648391121	3.155980757
18	3.753486727	0.095265717	3.848752445	18	3.753486727	-0.123523427	3.629963301
19	3.534947226	-0.241937074	3.293010152	19	3.534947226	0.454808819	3.989756044
20	3.234463562	0.110765109	3.345228672	20	3.234463562	0.562235841	3.796699403
21	2.955978889	0.544877628	3.500856518	21	2.955978889	-0.163608588	2.792370301
22	2.79854148	0.047741306	2.846282786	22	2.79854148	0.131562855	2.930104335
23	2.83426602	-0.488325425	2.345940596	23	2.83426602	0.010103546	2.844369566
24	3.092287752	0.209662178	3.30194993	24	3.092287752	-0.234393057	2.857894694
25	3.552279741	-0.12339204	3.4288877	25	3.552279741	0.964317942	4.516597683
26	4.148975719	-0.20837687	3.940598849	26	4.148975719	0.057314639	4.206290358
27	4.786692137	0.037066937	4.823759073	27	4.786692137	-1.003106021	3.783586116
28	5.360618201	-1.219627094	4.140991107	28	5.360618201	0.47310205	5.833720251
29	5.780136775	0.20760852	5.987745295	29	5.780136775	1.114970298	6.895107073
30	5.988987117	0.391285541	6.380272658	30	5.988987117	-0.613301012	5.375686104
31	5.977783159	-0.586448298	5.391334861	31	5.977783159	-0.143701009	5.834082149
32	5.786101954	-0.975618632	4.810483322	32	5.786101954	0.040573247	5.826675202
33	5.493677596	-0.421282099	5.072395497	33	5.493677596	0.488930669	5.982608265
34	5.202660002	0.343059426	5.545719428	34	5.202660002	0.145739652	5.348399654
35	5.01488608	-0.544052756	4.470833324	35	5.01488608	0.681833385	5.696719465
36	5.009220534	-0.644071067	4.365149467	36	5.009220534	0.336879993	5.346100526
37	5.224008093	0.146587837	5.370595929	37	5.224008093	-0.880167045	4.343841048
38	5.648541632	-0.006063012	5.64247862	38	5.648541632	-0.377336096	5.271205536
39	6.225444555	0.412921276	6.638365831	39	6.225444555	0.406852212	6.632296767
40	6.863437361	-0.495869275	6.367568086	40	6.863437361	0.48205554	7.345492901
41	7.457647758	0.523426841	7.981074599	41	7.457647758	-0.315199523	7.142448235
42	7.912945251	0.609357626	8.522302877	42	7.912945251	-0.21027772	7.70266753
43	8.165108118	0.154135114	8.319243232	43	8.165108118	0.394798755	8.559906873
44	8.195113052	-0.052132826	8.142980226	44	8.195113052	-0.890056272	7.30505678
45	8.033367932	-1.186987621	6.846380311	45	8.033367932	-0.294935058	7.738432874
46	7.752945931	-0.087052682	7.66589325	46	7.752945931	0.665569956	8.418515888
47	7.453326387	1.064736832	8.518063219	47	7.453326387	-0.292052374	7.161274013
48	7.238260181	-0.294937588	6.943322594	48	7.238260181	-0.335015953	6.903244228
49	7.192684679	-1.303899893	5.888784786	49	7.192684679	-0.048476375	7.144208304
50	7.363824764	0.601831507	7.965656271	50	7.363824764	0.336297903	7.700122667
51	7.75068508	-0.035241106	7.715443974	51	7.75068508	0.036887734	7.787572814
52	8.3042715	-0.354789262	7.949482238	52	8.3042715	0.197123916	8.501395416
53	8.938492555	-0.083178362	8.855314194	53	8.938492555	-0.124679259	8.813813297
54	9.549315187	-0.28803368	9.261281507	54	9.549315187	0.176477378	9.725792565



## NOTE

In part (g) of Question 1.

This page repeat 1.(b).

We use  $\text{NORM.INV}(\text{RAND}(), 0, 0.5)$  to generate disturbances that follow Gaussian Process with a expected value of 0 and a variance of 0.5.

We would like to observe if there will be effects when the variance increments.

Index

D_bar_6.5_to_42.5	D_bar_7_to_42	L+T
1		0.170832289
2		0.341664578
3		0.512496868
4		0.683329157
5		0.854161446
6		1.024993735
7	1.081883886	1.12715367
8	1.172423454	1.219348361
9	1.266273267	1.346176491
10	1.426079715	1.531345086
11	1.636610457	1.682733192
12	1.728855928	1.860958563
13	1.993061198	2.088931549
14	2.184801901	2.297832818
15	2.410863736	2.506661226
16	2.602458717	2.704103879
17	2.805749042	2.891637089
18	2.977525137	3.021011514
19	3.064497891	3.194091979
20	3.323686067	3.404337582
21	3.484989098	3.517374195
22	3.549759293	3.622268513
23	3.694777734	3.850574664
24	4.006371594	4.079110044
25	4.151848494	4.228695415
26	4.305542336	4.390124661
27	4.474706986	4.607633568
28	4.74056015	4.841322455
29	4.94208476	5.060932672
30	5.179780585	5.283455828
31	5.387131071	5.379932877
32	5.372734684	5.41710615
33	5.461477616	5.580173893
34	5.69887017	5.761860697
35	5.824851224	5.835157106
36	5.845462988	5.942420547
37	6.039378106	6.152954137
38	6.266530167	6.328129399
39	6.389728632	6.46288799
40	6.536047349	6.663968859
41	6.791890369	6.852913475
42	6.913936581	6.978817568
43	7.043698556	7.345788436
44		7.516620725
45		7.687453015
46		7.858285304
47		8.029117593
48		8.199949882

New\_Simulation

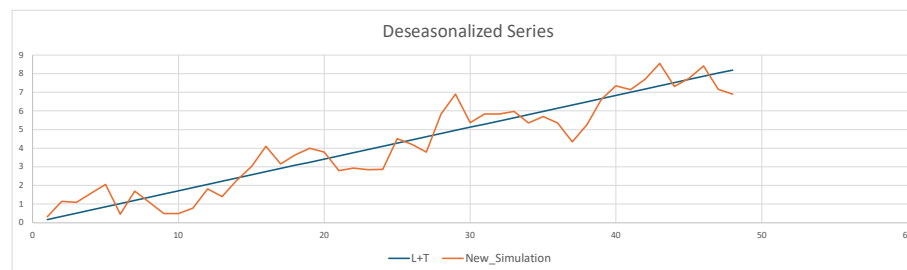
摘要輸出

0.319864752
1.144456231
0.998683028
0.99736779
0.997290373
0.093812039
36

L =	-0.082220594
T =	0.170832289

ANOVA					
	自由度	SS	MS	F	顯著值
迴歸	1	113.378562	113.378562	12882.90463	1.90106E-45
殘差	34	0.299223756	0.008800699		
總和	35	113.6777857			

	係數	標準誤	t 統計	P-值	下限 95%	上限 95%	下限 95.0%	上限 95.0%
截距	-0.082220594	0.040052608	-2.052814956	0.04785033	-0.163617287	-0.0008239	-0.163617287	-0.0008239
X 變數 1	0.170832289	0.001505092	113.5028838	1.90106E-45	0.167773574	0.173891004	0.167773574	0.173891004



NOTE

In part (g) of Question 1.

This page repeat 1.(c) to calculate L and T.



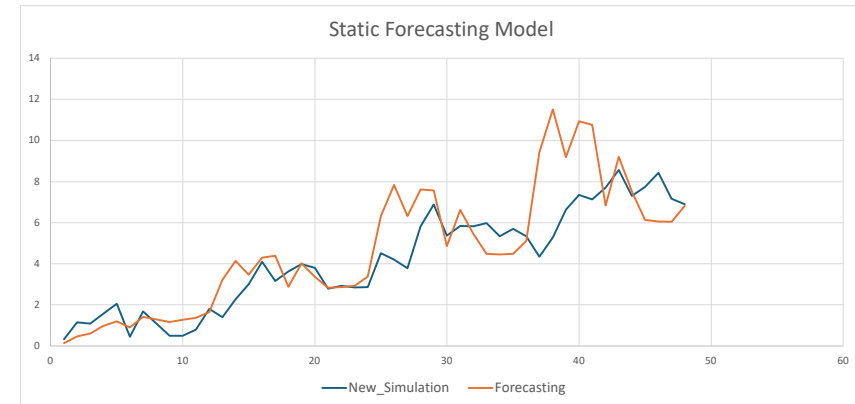
Index	New_Simulation	L+T	Seasonality Factor	Seasonality Factor_Average
1	0.319864752	0.088611696	3.60973515	1.510482947
2	1.144456231	0.259443985	4.41118815	1.79542211
3	1.088666401	0.430276274	2.530156709	1.39633922
4	1.567130052	0.601108563	2.60706659	1.620031021
5	2.049035109	0.771940852	2.654393925	1.554977377
6	0.459500054	0.942773142	0.487391965	0.96308092
7	1.688867616	1.113605431	1.51657631	1.268802812
8	1.083957378	1.28443772	0.843915872	1.011823084
9	0.493230532	1.455270009	0.338927161	0.807498656
10	0.490620433	1.626102298	0.301715601	0.778862033
11	0.783056429	1.796934588	0.435773475	0.760588623
12	1.814221646	1.967766877	0.921969806	0.841186623
13	1.406339571	2.138599166	0.657598485	1.510482947
14	2.270653983	2.309431455	0.983209083	1.79542211
15	3.006343779	2.480263745	1.212106489	1.39633922
16	4.093498953	2.651096034	1.544077959	1.620031021
17	3.155980757	2.821928323	1.118377363	1.554977377
18	3.629963301	2.992760612	1.212914687	0.96308092
19	3.989756044	3.163592901	1.26114711	1.268802812
20	3.796699403	3.334425191	1.138636852	1.011823084
21	2.792370301	3.50525748	0.796623448	0.807498656
22	2.930104335	3.676089769	0.797070942	0.778862033
23	2.844369566	3.846922058	0.739388405	0.760588623
24	2.857894694	4.017754347	0.711316434	0.841186623
25	4.516597683	4.188586637	1.078310675	1.510482947
26	4.206290358	4.359418926	0.964874088	1.79542211
27	3.783586116	4.530251215	0.835182407	1.39633922
28	5.833720251	4.701083504	1.240930999	1.620031021
29	6.895107073	4.871915793	1.415276324	1.554977377
30	5.375686104	5.042748083	1.066023132	0.96308092
31	5.834082149	5.213580372	1.11901644	1.268802812
32	5.826675202	5.384412661	1.082137564	1.011823084
33	5.982608265	5.55524495	1.076929698	0.807498656
34	5.348399654	5.72607724	0.934042527	0.778862033
35	5.696719465	5.896909529	0.966051698	0.760588623
36	5.346100526	6.067741818	0.881069216	0.841186623
37	4.343841048	6.238574107	0.696287481	1.510482947
38	5.271205536	6.409406396	0.822417118	1.79542211
39	6.632296767	6.580238686	1.007911276	1.39633922
40	7.345492901	6.751070975	1.088048537	1.620031021
41	7.142448235	6.921903264	1.031861897	1.554977377
42	7.70266753	7.092735553	1.085993898	0.96308092
43	8.559906873	7.263567842	1.178471387	1.268802812
44	7.30505678	7.434400132	0.982602046	1.011823084
45	7.738432874	7.605232421	1.017514317	0.807498656
46	8.418515888	7.77606471	1.082619063	0.778862033
47	7.161274013	7.946896999	0.901140912	0.760588623
48	6.903244228	8.117729288	0.850391037	0.841186623

## NOTE

In part (g) of Question 1.

This page repeat 1.(d) to calculate  $S_t$ .

Index	New_Simulation	Forecasting	$E_t$	$E_t^2$	$ E_t  /  D_t $	MSE = 3.190049713 MAPE = 39.06%
1	0.319864752	0.133846455	-0.186018297	0.034602807	0.581552971	
2	1.144456231	0.465811466	-0.678644765	0.460558717	0.592984464	
3	1.088666401	0.600811637	-0.487854764	0.238002271	0.448121448	
4	1.567130052	0.97381452	-0.593315532	0.352023231	0.37860006	
5	2.049035109	1.200350562	-0.848684547	0.72026546	0.414187411	
6	0.459500054	0.907966825	0.448466771	0.201122445	0.975988507	
7	1.688867616	1.412945702	-0.275921913	0.076132902	0.163376875	
8	1.083957378	1.299623735	0.215666356	0.046511977	0.198962026	
9	0.493230532	1.175128577	0.681898044	0.464984943	1.382513854	
10	0.490620433	1.266509342	0.77588891	0.6020036	1.581444347	
11	0.783056429	1.366728003	0.583671574	0.340672506	0.745376134	
12	1.814221646	1.655259175	-0.158962472	0.025269067	0.087620205	
13	1.406339571	3.230317572	1.823978001	3.326895748	1.296968413	
14	2.270653983	4.146404295	1.875750312	3.518439235	0.82608373	
15	3.006343779	3.463289543	0.456945764	0.208799431	0.151993849	
16	4.093498953	4.294857815	0.201358862	0.040545391	0.049189914	
17	3.155980757	4.388034702	1.232053945	1.517956924	0.390387027	
18	3.629963301	2.882270644	-0.747692656	0.559044308	0.205978021	
19	3.989756044	4.013975569	0.024219525	0.000586585	0.006070428	
20	3.796699403	3.373848378	-0.422851025	0.178802989	0.111373322	
21	2.792370301	2.830490704	0.038120403	0.001453165	0.013651629	
22	2.930104335	2.863166752	-0.066937583	0.00448064	0.022844778	
23	2.844369566	2.92592515	0.081555584	0.006651313	0.02867264	
24	2.857894694	3.379681213	0.521786518	0.272261171	0.182577237	
25	4.516597683	6.326788689	1.810191006	3.276791478	0.400786418	
26	4.206290358	7.826997124	3.620706766	13.10951749	0.86078384	
27	3.783586116	6.325767449	2.542181333	6.462685928	0.67189731	
28	5.833720251	7.615901111	1.78218086	3.176168619	0.305496456	
29	6.895107073	7.575718842	0.680611769	0.46323238	0.098709384	
30	5.375686104	4.856574464	-0.51911164	0.269476895	0.096566583	
31	5.834082149	6.615005437	0.780923287	0.609841181	0.133855381	
32	5.826675202	5.448073022	-0.37860218	0.143339611	0.064977396	
33	5.982608265	4.485852831	-1.496755433	2.240276827	0.250184429	
34	5.348399654	4.459824161	-0.888575493	0.789566407	0.166138574	
35	5.696719465	4.485122297	-1.211597168	1.467967698	0.212683313	
36	5.346100526	5.104103251	-0.241997275	0.058562681	0.045266129	
37	4.343841048	9.423259805	5.079418757	25.80049491	1.169338081	
38	5.271205536	11.50758995	6.236384417	38.8924906	1.183104012	
39	6.632296767	9.188245354	2.555948587	6.532873181	0.385379104	
40	7.345492901	10.93694441	3.591451506	12.89852392	0.488932677	
41	7.142448235	10.76340298	3.620954748	13.11131328	0.506962687	
42	7.70266753	6.830878284	-0.871789246	0.76001649	0.113180173	
43	8.559906873	9.216035304	0.656128431	0.430504518	0.076651352	
44	7.30505678	7.522297666	0.217240886	0.047193602	0.029738425	
45	7.738432874	6.141214959	-1.597217916	2.55110507	0.206400694	
46	8.418515888	6.05648157	-2.362034317	5.579206116	0.280576095	
47	7.161274013	6.044319444	-1.116954569	1.24758751	0.155971489	
48	6.903244228	6.828525289	-0.074718939	0.00558292	0.010823743	



## NOTE

In part (g) of Question 1.

This page repeat 1.(e) to calculate MSE and MAPE.

We can see the performance of this model is worse than the first model.

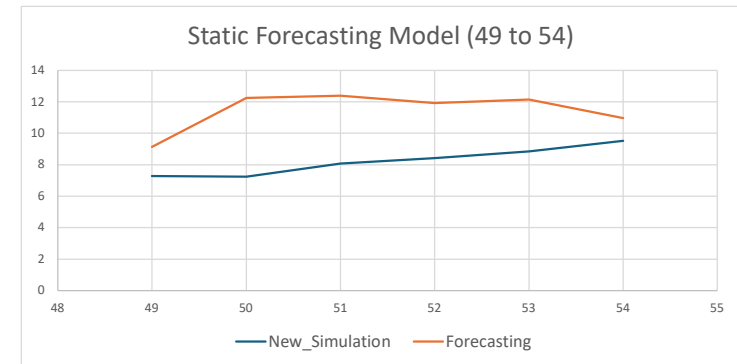
My thought is that the reason is the larger variance.

Even though the expected value is zero, it still affects  $S_t$  more than before.

Once there is an outlier, it will boost or sharply decrease  $S_t$ .

It will severely affect the performance of the model.

Index	New_Simulation	Forecasting	$E_t$	$E_t^2$	$ E_t  /  D_t $	MSE = 28.2024839 MAPE = 61.89%
49	7.192684679	12.51973092	5.327046243	28.37742167	0.740620016	
50	7.363824764	15.18818278	7.824358018	61.22057839	1.062539953	
51	7.75068508	12.05072326	4.300038181	18.49032835	0.554794594	
52	8.3042715	14.2579877	5.953716202	35.44673662	0.716946237	
53	8.938492555	13.95108712	5.012594567	25.1261043	0.560787463	
54	9.549315187	8.805182103	-0.744133084	0.553734046	0.077925283	



## NOTE

In part (g) of Question 1.

This page repeat 1.(f) to calculate MSE and MAPE of future.

We can see the performance of this model is worse than the first model.

I think the reason is the same as part (e).

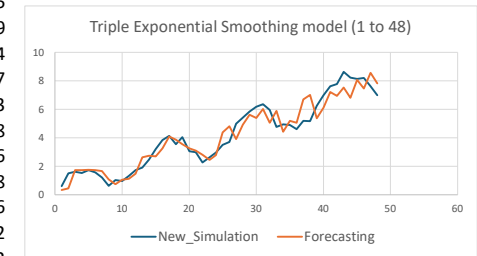
However, the prediction of future are both worth than MAPE of  $y_{49}$  to  $y_{54}$ .

My thought is the average of  $S_t$ .

When  $L+T$  is tiny, we need a larger  $S_t$  to make the regression line to reach the data, whereas  $L+T$  becomes larger and larger, we do not need the large  $S_t$  anymore.  $S_t$  should become smaller accordingly.

To conclude, we can observe that the forecasting is lower than the data in the first cycle, and have a great performance in the second cycle. Nonetheless, it becomes higher than the data in the third cycle, and the gap is even larger in the fourth cycle. The phenomenon fits my hypothesis. Thus, I think this is the reason why MSE and MAPE is larger than before.

Index	New_Simulation	L_t	T_t	S_t	Forecasting	E_t	E_t^2	E_t  /  D_t	Alpha = 1 Beta = 0 Gamma = 0.002575604 MSE = 0.500551001 MAPE = 12.24%
1	0.600087879	0.139228497	0.160819878	1.13766035	0.341353139	-0.258734739	0.066943665	0.111556437	
2	1.505053833	1.005359802	0.160819878	1.497030048	0.449181433	-1.055872399	1.114866524	0.740748603	
3	1.611394937	1.084498599	0.160819878	1.485843264	1.732760222	0.121365284	0.014729532	0.009140858	
4	1.529009825	1.091186002	0.160819878	1.401236657	1.744985899	0.215976074	0.046645664	0.030507106	
5	1.717634501	1.224386658	0.160819878	1.402853004	1.75638021	0.038745709	0.00150123	0.00087401	
6	1.569358404	1.262585348	0.160819878	1.242972134	1.721773124	0.15241472	0.023230247	0.014802385	
7	1.213340864	1.034052128	0.160819878	1.17338462	1.67020218	0.456860936	0.208721915	0.172022489	
8	0.627506185	0.693513117	0.160819878	0.904822375	1.081146927	0.453640741	0.205789922	0.327948834	
9	1.046563402	1.199261089	0.160819878	0.872673525	0.745553786	-0.301009616	0.090606789	0.086575537	
10	0.976702436	1.242684412	0.160819878	0.785961767	1.06897164	0.092269204	0.008513606	0.008716683	
11	1.334178937	1.653916172	0.160819878	0.806678694	1.132177008	-0.20200193	0.04080478	0.030584188	
12	1.732693728	2.132108917	0.160819878	0.812666611	1.474775396	-0.257918332	0.066521866	0.038392166	
13	1.910279282	1.667155813	0.160819878	1.145831282	2.62730954	0.717030258	0.514132391	0.269139908	
14	2.475255709	1.653444239	0.160819878	1.497030048	2.736534537	0.261278828	0.068266626	0.027579626	
15	3.21974017	2.16694469	0.160819878	1.485843264	2.695712117	-0.524028053	0.2746054	0.085288062	
16	3.858407379	2.753572967	0.160819878	1.401236657	3.261749041	-0.596658338	0.356001172	0.092266352	
17	4.134378305	2.947121539	0.160819878	1.402853004	4.088464758	-0.045913547	0.002108054	0.000509884	
18	3.56374419	2.867115113	0.160819878	1.242972134	3.863084575	0.299340386	0.089604667	0.025143406	
19	4.050753357	3.452195714	0.160819878	1.17338462	3.552932349	-0.497821008	0.247825756	0.061180164	
20	3.063368034	3.385601549	0.160819878	0.904822375	3.269137349	0.205769315	0.042341011	0.013821719	
21	2.992466902	3.429079508	0.160819878	0.872673525	3.094868089	0.102401187	0.010486003	0.003504133	
22	2.258522636	2.873578246	0.160819878	0.785961767	2.821523665	0.563001029	0.316970158	0.140344025	
23	2.636845767	3.26876833	0.160819878	0.806678694	2.447784316	-0.189061451	0.035744232	0.013555678	
24	2.979205334	3.665962516	0.160819878	0.812666611	2.787111826	-0.192093508	0.036899916	0.012385825	
25	3.506231084	3.05998897	0.160819878	1.145831282	4.384846975	0.878615891	0.771965884	0.220169711	
26	3.710707327	2.478712656	0.160819878	1.497030048	4.821647626	1.110940299	1.234188348	0.332601911	
27	4.994392448	3.361318499	0.160819878	1.485843264	3.921931635	-1.072460813	1.150172196	0.230292715	
28	5.404772932	3.857144977	0.160819878	1.401236657	4.935349404	-0.469423528	0.220358449	0.040771084	
29	5.857347081	4.175310645	0.160819878	1.402853004	5.636614068	-0.220733014	0.048723063	0.008318282	
30	6.191878695	4.981510467	0.160819878	1.242972134	5.389689409	-0.802189287	0.643507652	0.10392769	
31	6.371421773	5.429951667	0.160819878	1.17338462	6.033931338	-0.337490435	0.113899793	0.017876668	
32	5.952888327	6.579068436	0.160819878	0.904822375	5.058655186	-0.89423314	0.799652909	0.134330239	
33	4.771652916	5.467855709	0.160819878	0.872673525	5.881722095	1.110069178	1.23225358	0.258244596	
34	4.939703188	6.284915368	0.160819878	0.785961767	4.423923811	-0.515779378	0.266028367	0.053855132	
35	4.896834936	6.070366023	0.160819878	0.806678694	5.19963729	0.302802354	0.091689266	0.01872419	
36	4.616545059	5.680736721	0.160819878	0.812666611	5.063876729	0.44733167	0.200105623	0.04334532	
37	5.182830351	4.523205495	0.160819878	1.145831282	6.693438285	1.510607935	2.281936333	0.440287676	
38	5.176635291	3.457936798	0.160819878	1.497030048	7.012126731	1.835491439	3.369028823	0.650814406	
39	6.251170519	4.207153387	0.160819878	1.485843264	5.37690523	-0.874265289	0.764339795	0.122271468	
40	6.99363277	4.99104362	0.160819878	1.401236657	6.120564255	-0.873069022	0.762249517	0.10899192	
41	7.622306093	5.433431778	0.160819878	1.402853004	7.227307186	-0.394998907	0.156024136	0.020469414	
42	7.779843031	6.25906472	0.160819878	1.242972134	6.953498918	-0.826344113	0.682844593	0.087770999	
43	8.628637782	7.353631226	0.160819878	1.17338462	7.53299385	-1.095643931	1.200435625	0.139122264	
44	8.225424803	9.090651416	0.160819878	0.904822375	6.799243494	-1.42618131	2.033993128	0.247281226	
45	8.142862625	9.330938077	0.160819878	0.872673525	8.073514068	-0.069348557	0.004809222	0.000590606	
46	8.213399724	10.45012629	0.160819878	0.785961767	7.460158855	-0.753240869	0.567371806	0.069078801	
47	7.632732616	9.46192415	0.160819878	0.806678694	8.559624195	0.92689158	0.859128001	0.112558378	
48	6.990696966	8.602170769	0.160819878	0.812666611	7.820082778	0.829385811	0.687880824	0.098399462	



## NOTE

In part (a) of Question 2.

We compare the result of triple exponential smoothing model with static model like part (e) of Question 1.

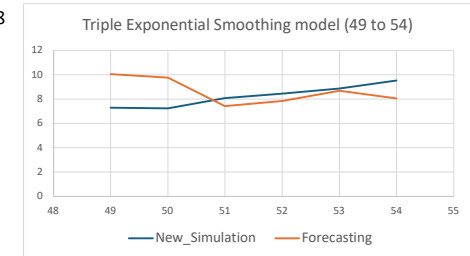
We use alpha to forget the previous data of L. Beta is for T, and Gamma is for S.

The core concept of estimation is take average of present data and previos data by the weight.

The performance is better because of the lager S in the previous cycle can be forgot.

It avoid the boost like static model, the model is robust even at the large<sup>42</sup>index(t).

Index	New_Simulation	L_t	T_t	S_t	Forecasting	E_t	E_t^2	E_t   /   D_t	Alpha =
49	7.283680668	6.356678146	0.160819878	1.145831282	10.04090881	2.757228138	7.602307006	0.37854874	1
50	7.233483711	4.831889460	0.160819878	1.497030048	9.756890381	2.523406669	6.367581219	0.348850812	0
51	8.073618224	5.433694402	0.160819878	1.485843264	7.418383538	-0.655234686	0.429332494	0.081157502	Gamma = 0.002575604
52	8.426026426	6.013278617	0.160819878	1.401236657	7.839238486	-0.58678794	0.344320086	0.069639936	MSE = 2.819054182
53	8.859458079	6.315314614	0.160819878	1.402853004	8.661352622	-0.198105457	0.039245772	0.0223609	MAPE = 17.57%
54	9.509633654	7.650721520	0.160819878	1.242972134	8.049654709	-1.459978944	2.131538518	0.153526308	



## NOTE

In part (b) of Question 2.

We compare the result of triple exponential smoothing model in the future with static model like part (f) of Question 1.

The performance is also better than the static model because of  $S_t$  is more proper to fit the status of data.