

1. Suppose that a time series follows a following horizontal process:

$$X_t = c + a_t$$

where a_t is a white noise with mean 0 and variance σ_a^2 .

Show for a simple moving average with span N that

$$\text{a) } \text{Cov}[M_t, M_{t-k}] = \sigma_a^2 (N-k) / N^2, \quad k < N$$

$$\text{b) } \text{Corr}[M_t, M_{t-k}] = 1 - \frac{k}{N}, \quad k = 1, \dots, N-1$$

1-a)

$$\text{Cov}[M_t, M_{t-k}] = \frac{1}{N^2} \text{Cov}[(X_t + X_{t-1} + \dots + X_{t-N+1})(X_{t-k} + X_{t-k-1} + \dots + X_{t-k-N+1})]$$

$\text{Cov}[X_i, X_j]$ 에서 $i=j$ 일 때 σ_a^2 , $i \neq j$ 일 때 0.

$k < N$ 일 때 같은 X 항이 $(N-k)$ 개 있으므로

$$\text{Cov}[M_t, M_{t-k}] = \frac{N-k}{N^2} \cdot \sigma_a^2$$

1-b)

$$\rho[M_t, M_{t-k}] = \frac{\text{Cov}[M_t, M_{t-k}]}{\sqrt{\text{Var}[M_t]} \sqrt{\text{Var}[M_{t-k}]}} \quad \left(\text{Var}[M_t] = \text{Var}[M_{t-k}] = \frac{\sigma_a^2}{N} \right)$$

$$= \frac{\cancel{\sigma_a^2} (N-k) / N^2}{\sqrt{\frac{\cancel{\sigma_a^2}}{N}} \cdot \sqrt{\frac{\cancel{\sigma_a^2}}{N}}} = \frac{N-k}{N} = 1 - \frac{k}{N}$$

2. Suppose that a time series follows a following linear trend process:

$$X_t = c + bt + a_t$$

where b and c are constants. The term a_t is a white noise with mean 0 and variance σ_a^2 .

a) Prove for a simple exponential smoothing

$$E[S_T] = c + bT - \frac{1-\alpha}{\alpha}b = E[X_T] - \frac{1-\alpha}{\alpha}b$$

b) Prove for a double exponential smoothing

$$E[S_T^{(2)}] = E[S_T] - \frac{1-\alpha}{\alpha}b$$

2-(a)

$$S_T = \alpha \cdot \sum_{i=0}^{\infty} (1-\alpha)^i X_{T-i}$$

$$E[S_T] = \alpha \cdot \sum_{i=0}^{\infty} (1-\alpha)^i E[X_{T-i}] = \alpha \sum_{i=0}^{\infty} (1-\alpha)^i [c + b(T-i)]$$

$$= \alpha(c + bT) \sum_{i=0}^{\infty} (1-\alpha)^i - \alpha \cdot b \sum_{i=1}^{\infty} i(1-\alpha)^i$$

$$= \alpha(c + bT)(1/\alpha) + b\alpha \cdot (1-\alpha) \frac{d}{d\alpha} \sum_{i=1}^{\infty} i(1-\alpha)^{i-1}$$

$$= \alpha(c + bT)(1/\alpha) + b\alpha(1-\alpha)(-1/\alpha^2)$$

$$= c + bT - b(1-\alpha)/\alpha = E[X_T] - \frac{1-\alpha}{\alpha}b$$

2-(b)

$$E[S_T^{(2)}] = E[S_T] - \frac{1-\alpha}{\alpha}b$$

$$\Rightarrow S_T^{(2)} = \alpha \cdot S_T^{(1)} + (1-\alpha)S_{T-1}^{(2)} = \alpha \cdot \sum_{j=0}^{n-1} (1-\alpha)^j S_{T-j}^{(1)} + (1-\alpha)^T S_0^{(2)}$$

• (or) $T \rightarrow \infty$ or after

$$(i) \lim_{T \rightarrow \infty} \alpha \cdot \sum_{j=0}^{n-1} (1-\alpha)^j = \frac{\alpha}{1-(1-\alpha)} = 1$$

$$(ii) \lim_{T \rightarrow \infty} (1-\alpha)^T S_0^{(2)} = 0$$

$$\therefore \text{By (i), (ii), } S_T^{(2)} = \alpha \sum_{j=0}^{T-1} (1-\alpha)^j S_{n-T}^{(1)}$$

$$\begin{aligned}
 E[S_T^{(2)}] &= E\left[\alpha \sum_{j=0}^{T-1} (1-\alpha)^j S_{T-j}^{(1)}\right] = \alpha \sum_{j=0}^{T-1} (1-\alpha)^j E[S_{T-j}^{(1)}] \\
 &= \alpha \sum_{j=0}^{T-1} (1-\alpha)^j E\left[c + b(n-j+1) - \frac{1}{\alpha} b\right] \\
 &= \underbrace{\alpha \sum_{j=0}^{T-1} (1-\alpha)^j E\left[c + b(n+1) - \frac{1}{\alpha} b\right]}_{\textcircled{1}} - \underbrace{\alpha b \sum_{j=0}^{T-1} (1-\alpha)^j j}_{\textcircled{2}}
 \end{aligned}$$

①: By (i), ① = $E\left[c + b(n+1) - \frac{1}{\alpha} b\right] = E[S_n^{(1)}]$

②: $A \stackrel{\text{let}}{=} \sum_{j=0}^{\infty} (1-\alpha)^j j = 0 + (1-\alpha) + 2 \cdot (1-\alpha)^2 + \dots$

$\alpha \cdot A = \quad \quad \quad (1-\alpha)^2 + \dots$

$(1-\alpha)A = (1-\alpha) + (1-\alpha)^2 + \dots$

$\therefore A = \frac{1-\alpha}{\alpha^2}$

$\therefore \textcircled{2} = \alpha \cdot b = \frac{1-\alpha}{\alpha^2} = b \cdot \frac{1-\alpha}{\alpha}$

따라서 $E[S_T^{(2)}] = E[S_T^{(1)}] - \frac{1-\alpha}{\alpha} b$

3. Consider series J05 (gold prices in US dollars).

a) Obtain double moving averages with $N=3$ and $N=6$, respectively and make graphs comparing with the original series.

b) Obtain double exponential smoothing with $\alpha=0.1$ and $\alpha=0.3$, respectively and make graphs comparing with the original series.

c) Apply Holt's model with $\alpha=0.1$ to the data until 1999 and make one-step-ahead forecasts from 2000 to 2017. Calculate forecasting performance measures based on the forecast errors.

3-a) see the R code (HW1-3.r)

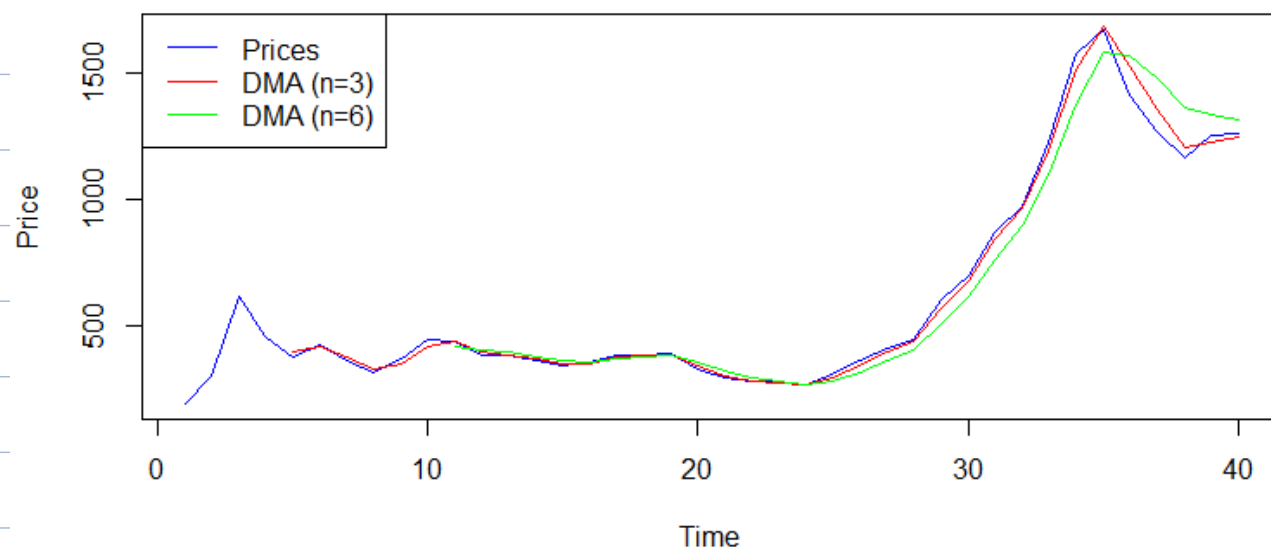
DMA ($n=3$)

```
> df_ts_dma.3
[1] NA NA NA NA 437.7911 453.9689 429.6689 391.0778
[9] 367.4900 364.3378 380.9144 405.1333 412.8967 399.0867 379.7389 364.7156
[17] 360.4133 364.6633 374.6678 376.3667 363.6144 335.6011 307.6622 287.1578
[25] 282.2544 292.5078 320.6700 360.3500 417.3889 490.8822 596.9022 717.1589
[33] 864.4056 1041.8800 1255.8044 1431.6822 1495.9289 1426.2256 1317.9511 1242.5511
```

DMA ($n=6$)

```
> df_ts_dma.6
[1] NA NA NA NA NA NA NA 414.4344
[9] 410.7294 397.0033 385.9961 386.3117 388.6172 390.0006 392.4361 388.8061
[17] 379.7500 372.2011 369.6917 368.3900 364.1389 355.1344 342.0144 325.3861
[25] 308.9278 300.0850 303.9139 321.3022 354.9483 405.7761 478.6261 567.2739
[33] 677.6439 819.3911 986.4817 1148.0439 1268.9044 1341.0150 1374.8167 1369.2400
```

Graph



3-b) See the R code (HW1-3.r)

$d=0.1$, $n=23$ 설정

```
> dema.0.1
```

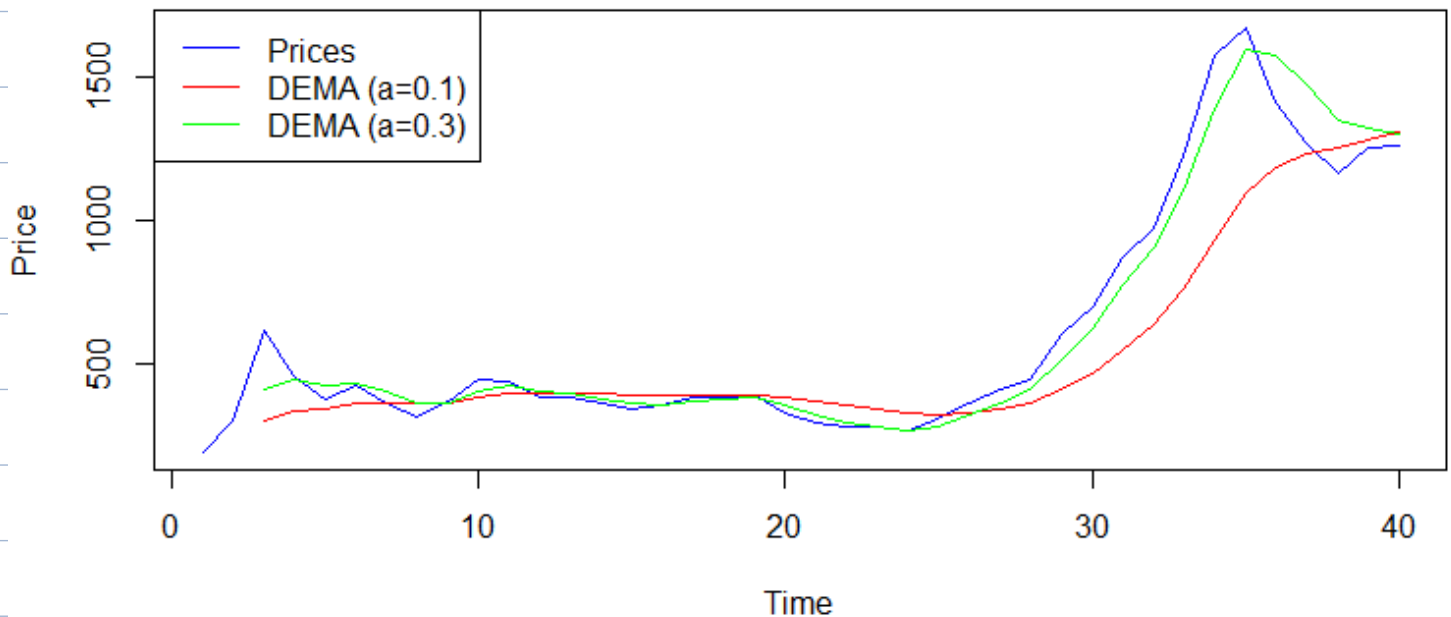
[1]	NA	NA	303.8760	335.0434	345.5788	363.5981	366.8471	361.1570
[9]	365.7139	384.3167	398.2255	399.2304	400.3809	397.0976	390.6109	387.9053
[17]	390.0556	391.7193	393.7160	384.5247	369.4794	353.5900	340.0700	326.9778
[25]	323.1553	330.1159	344.7976	364.0367	410.6952	467.8272	549.9281	638.7242
[33]	761.8976	932.2243	1095.0992	1183.7943	1230.3939	1248.2950	1279.1532	1305.0713

$d=0.3$, $n=23$ 설정

```
> dema.0.3
```

[1]	NA	NA	413.5080	448.3529	423.5889	431.7722	402.9831	362.9033
[9]	364.9405	406.1692	425.2023	406.9973	397.1183	380.1943	360.9494	358.0831
[17]	369.2433	376.0024	381.9868	356.5929	322.9006	295.8832	281.3568	269.9251
[25]	284.1304	320.7468	365.9039	409.9177	515.8455	622.3896	771.2169	904.4081
[33]	1104.3932	1390.1673	1596.2230	1572.2859	1472.1984	1350.4003	1319.8660	1301.9267

Graph



3-C) see the R code (HW1-3.r)

```
> forecast$accurate
```

	SST	SSE	MSE	RMSE	MAPE	MPE
1.366073e+05	3.418239e+05	1.709119e+04	1.307333e+02	2.389962e+01	-3.634062e+00	
	MAE	ME	R.squared	R.adj.squared	RW.R.squared	AIC
9.260834e+01	1.205199e+00	-1.502238e+00	-1.627350e+00	1.597243e+19	2.163222e+02	
	SBC	APC				
2.185043e+02	1.864494e+04					

4. Consider series J04 (electricity consumption in residential & commercial use). We would like to apply Winters' model.

- Obtain suitable initial values for this model using data in first three years.
- Obtain smoothed values of level, trend, and seasonality using data from 2000 to 2004.
- Make one-step-ahead forecasts from year 2004 to 2017 and calculate forecast errors.

4-a) See the R code (HW1-4.r)

$$L_0 = 1583.917$$

$$b_0 = 21.968$$

4-b) See the R code (HW1-4.r)

$$L_0 = 2957.474$$

$$b_0 = 64.375$$

$$s_1 = 307.473$$

$$s_2 = -211.33$$

$$s_3 = -51.776$$

$$s_4 = -202.47$$

4-c) See the R code.

```
> (x.2004_2017-hw.2004_2017.pred)
```

	Qtr1	Qtr2	Qtr3	Qtr4
6	728.677859	497.108754	456.177797	594.500000
7	774.177859	466.608754	563.677797	487.000000
8	723.677859	235.108754	233.177797	191.500000
9	490.177859	6.608754	101.677797	45.000000
10	203.677859	-211.891246	-89.822203	-262.500000
11	-228.822141	-560.391246	-505.322203	-466.000000
12	-369.322141	-735.891246	-613.822203	-756.500000
13	-521.822141	-943.391246	-536.322203	-872.000000
14	-778.322141	-1145.891246	-710.822203	-984.500000