# Solutions to Exercises

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# 2.9 Exercises

#### Exercise 2.1

a.  $ETS(A,A_d,N)$ 

$$y_t = \ell_{t-1} + \phi b_{t-1} + \varepsilon_t$$
  
$$\ell_t = \ell_{t-1} + \phi b_{t-1} + \alpha \varepsilon_t$$
  
$$b_t = \phi b_{t-1} + \beta \varepsilon_t$$

b.

$$\mathbf{x}_{t} = \begin{bmatrix} \ell_{t} & b_{t} \end{bmatrix}'$$

$$\mathbf{y}_{t} = \begin{bmatrix} 1 & \phi \end{bmatrix} \mathbf{x}_{t-1} + \varepsilon_{t}$$

$$\mathbf{x}_{t} = \begin{bmatrix} 1 & \phi \\ 0 & \phi \end{bmatrix} \mathbf{x}_{t-1} + \begin{bmatrix} \alpha \\ \beta \end{bmatrix} \varepsilon_{t}$$

$$egin{aligned} m{w}(m{x}_{t-1}) &= egin{bmatrix} 1 & \phi \end{bmatrix} m{x}_{t-1} & m{r}(m{x}_{t-1}) &= 1 \ m{f}(m{x}_{t-1}) &= egin{bmatrix} 1 & \phi \ 0 & \phi \end{bmatrix} m{x}_{t-1} & m{g}(m{x}_{t-1}) &= egin{bmatrix} lpha \ eta \end{bmatrix} \end{aligned}$$

c. ETS(A,A,A)

$$y_t = \ell_{t-1} + b_{t-1} + s_{t-m} + \varepsilon_t$$
  

$$\ell_t = \ell_{t-1} + b_{t-1} + \alpha \varepsilon_t$$
  

$$b_t = b_{t-1} + \beta \varepsilon_t$$
  

$$s_t = s_{t-m} + \gamma \varepsilon_t$$

$$\mathbf{x}_{t} = \begin{bmatrix} \ell_{t} & b_{t} & s_{t} \end{bmatrix}'$$

$$\mathbf{y}_{t} = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix} \mathbf{x}_{t-1} + \varepsilon_{t}$$

$$\mathbf{x}_{t} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \mathbf{x}_{t-1} + \begin{bmatrix} \alpha \\ \beta \\ \gamma \end{bmatrix} \varepsilon_{t}$$

## $d. ETS(M,A_d,N)$

$$y_t = (\ell_{t-1} + \phi b_{t-1})(1 + \varepsilon_t)$$
  

$$\ell_t = (\ell_{t-1} + \phi b_{t-1})(1 + \alpha \varepsilon_t)$$
  

$$b_t = \phi b_{t-1} + \beta (\ell_{t-1} + \phi b_{t-1})\varepsilon_t$$

$$\begin{aligned} \boldsymbol{x}_{t} &= \begin{bmatrix} \ell_{t} & b_{t} \end{bmatrix}' \\ y_{t} &= \begin{bmatrix} 1 & \phi \end{bmatrix} \boldsymbol{x}_{t-1} (1 + \varepsilon_{t}) \\ \boldsymbol{x}_{t} &= \begin{bmatrix} 1 & \phi \\ 0 & \phi \end{bmatrix} \boldsymbol{x}_{t-1} + \begin{bmatrix} 1 & 1 \end{bmatrix} \boldsymbol{x}_{t-1} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} \varepsilon_{t} \end{aligned}$$

$$egin{aligned} oldsymbol{w}(oldsymbol{x}_{t-1}) &= egin{bmatrix} 1 & \phi \end{bmatrix} oldsymbol{x}_{t-1} \ oldsymbol{f}(oldsymbol{x}_{t-1}) &= egin{bmatrix} 1 & \phi \end{bmatrix} oldsymbol{x}_{t-1} \ oldsymbol{g}(oldsymbol{x}_{t-1}) &= egin{bmatrix} 1 & \phi \end{bmatrix} oldsymbol{x}_{t-1} \ oldsymbol{\phi} \ oldsymbol{\beta} \end{aligned}$$

#### e. $ETS(M, A_d, A)$

$$y_{t} = (\ell_{t-1} + \phi b_{t-1} + s_{t-m})(1 + \varepsilon_{t})$$

$$\ell_{t} = \ell_{t-1} + \phi b_{t-1} + \alpha(\ell_{t-1} + \phi b_{t-1} + s_{t-m})\varepsilon_{t}$$

$$b_{t} = \phi b_{t-1} + \beta(\ell_{t-1} + \phi b_{t-1} + s_{t-m})\varepsilon_{t}$$

$$s_{t} = s_{t-m} + \gamma(\ell_{t-1} + \phi b_{t-1} + s_{t-m})\varepsilon_{t}$$

$$\begin{aligned} & \boldsymbol{x}_t = \begin{bmatrix} \ell_t & b_t & s_t & s_{t-1} & \dots & s_{t-m+2} & s_{t-m+1} \end{bmatrix}' \\ & \boldsymbol{y}_t = \begin{bmatrix} 1 & \phi & 0 & 0 & \dots & 0 & 0 \end{bmatrix} \boldsymbol{x}_{t-1} (1 + \varepsilon_t) \\ & \boldsymbol{x}_t = \begin{bmatrix} 1 & \phi & 0 & 0 & \dots & 0 & 0 \\ 0 & \phi & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 1 \\ 0 & 0 & 1 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 1 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \dots & 1 & 0 \end{bmatrix} \boldsymbol{x}_{t-1} + \begin{bmatrix} 1 & \phi & 0 & 0 & \dots & 0 & 1 \end{bmatrix} \boldsymbol{x}_{t-1} \begin{bmatrix} \alpha \\ \beta \\ \gamma \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \varepsilon_t \end{aligned}$$

#### Exercise 2.2

a. ETS(A,N,N)

$$y_t = \ell_{t-1} + \varepsilon_t$$
$$\ell_t = \ell_{t-1} + \alpha \varepsilon_t$$

$$\hat{y}_{t+1|t} = E[y_{t+1} \mid \boldsymbol{x}_t] = E[\ell_t + \varepsilon_{t+1} \mid \boldsymbol{x}_t] = E[\ell_t \mid \boldsymbol{x}_t] = \ell_t$$

$$\hat{y}_{t+2|t} = E[y_{t+2} \mid \boldsymbol{x}_t] = E[\ell_{t+1} + \varepsilon_{t+2} \mid \boldsymbol{x}_t] = E[\ell_{t+1} \mid \boldsymbol{x}_t] = \ell_t$$

$$\dots$$

$$\hat{y}_{t+h|t} = E[y_{t+h} \mid \boldsymbol{x}_t] = E[\ell_{t+h-1} + \varepsilon_{t+h} \mid \boldsymbol{x}_t] = E[\ell_{t+h-1} \mid \boldsymbol{x}_t] = \ell_t$$

$$v_{t+1|t} = \operatorname{var}(y_{t+1} \mid \boldsymbol{x}_t) = \operatorname{var}(\ell_t + \varepsilon_{t+1} \mid \boldsymbol{x}_t) = \operatorname{var}(\varepsilon_{t+1} \mid \boldsymbol{x}_t) = \sigma^2$$

$$v_{t+2|t} = \operatorname{var}(y_{t+2} \mid \boldsymbol{x}_t) = \operatorname{var}(\ell_{t+1} + \varepsilon_{t+2} \mid \boldsymbol{x}_t) = \operatorname{var}(\ell_t + \alpha \varepsilon_{t+1} + \varepsilon_{t+2} \mid \boldsymbol{x}_t)$$

$$= (1 + \alpha^2)\sigma^2$$

$$v_{t+3|t} = \operatorname{var}(y_{t+3} \mid \boldsymbol{x}_t) = \operatorname{var}(\ell_{t+2} + \varepsilon_{t+3} \mid \boldsymbol{x}_t) = \operatorname{var}(\ell_t + \alpha \varepsilon_{t+1} + \alpha \varepsilon_{t+2} + \varepsilon_{t+3} \mid \boldsymbol{x}_t)$$

$$= \alpha^2 \sigma^2 + \alpha^2 \sigma^2 + \sigma^2 = (1 + 2\alpha^2)\sigma^2$$
...
$$v_{t+h|t} = \operatorname{var}(y_{t+h} \mid \boldsymbol{x}_t) = [1 + (h-1)\alpha^2] \sigma^2$$

b. ETS(A,A,N)

$$y_t = \ell_{t-1} + b_{t-1}\varepsilon_t$$
  
$$\ell_t = \ell_{t-1} + b_{t-1}\alpha\varepsilon_t$$
  
$$b_t = b_{t-1} + \beta\varepsilon_t$$

$$\hat{y}_{t+1|t} = E[y_{t+1} \mid \boldsymbol{x}_t] = E[\ell_t + b_t + \varepsilon_{t+1} \mid \boldsymbol{x}_t] = E[\ell_t + b_t \mid \boldsymbol{x}_t] = \ell_t + b_t$$

$$\hat{y}_{t+2|t} = E[y_{t+2} \mid \boldsymbol{x}_t] = E[\ell_{t+1} + b_{t+1}\varepsilon_{t+2} \mid \boldsymbol{x}_t] = \ell_t + b_t + b_t = \ell_t + 2b_t$$
...
$$\hat{y}_{t+h|t} = E[y_{t+h} \mid \boldsymbol{x}_t] = E[\ell_{t+h-1} + b_{t+h-1} + \varepsilon_{t+h} \mid \boldsymbol{x}_t] = \ell_t + hb_t$$

$$\begin{aligned} v_{t+1|t} &= \text{var}(y_{t+1} \mid \boldsymbol{x}_t) = \text{var}(\ell_t + b_t + \varepsilon_{t+1} \mid \boldsymbol{x}_t) = \sigma^2 \\ v_{t+2|t} &= \text{var}(y_{t+2} \mid \boldsymbol{x}_t) = \text{var}(\ell_{t+1} + b_{t+1} + \varepsilon_{t+2} \mid \boldsymbol{x}_t) = \text{var}(\ell_t + b_t + \alpha \varepsilon_{t+1} + b_t + \beta \varepsilon_{t+1} + \varepsilon_{t+2} \mid \boldsymbol{x}_t) \\ &= (\alpha + \beta)^2 \sigma^2 + \sigma^2 = [1 + ((\alpha + \beta)^2] \sigma^2 \\ v_{t+3|t} &= \text{var}(y_{t+3} \mid \boldsymbol{x}_t) = \text{var}(\ell_{t+2} + b_{t+2} + \varepsilon_{t+3} \mid \boldsymbol{x}_t) \\ &= \text{var}(\ell_{t+1} + b_{t+1} + \alpha \varepsilon_{t+2} + b_{t+1} + \beta \varepsilon_{t+2} + \varepsilon_{t+3}) \\ &= \text{var}[\ell_t + b_t + \alpha \varepsilon_{t+1} + 2(b_t + \beta \varepsilon_{t+1}) + \alpha \varepsilon_{t+2} + \beta \varepsilon_{t+2} + \varepsilon_{t+3}] \\ &= \text{var}[(\alpha + 2\beta)\varepsilon_{t+1} + (\alpha + \beta)\varepsilon_{t+2} + \varepsilon_{t+3}] \\ &= [1 + (\alpha + \beta)^2 + (\alpha + 2\beta)^2]\sigma^2 = \left[1 + \sum_{j=1}^2 (\alpha + j\beta^2)\right]\sigma^2 \end{aligned}$$

. . .

$$v_{t+h|t} = \left[1 + \sum_{j=1}^{h-1} (\alpha + j\beta^2)\right] \sigma^2$$

## c. ETS(M,N,N)

b = 1.1238

$$\begin{aligned} y_t &= \ell_{t-1}(1+\varepsilon_t) \\ \ell_t &= \ell_{t-1}(1+\alpha\varepsilon_t) \\ \\ \hat{y}_{t+1/h} &= E[y_{t+1} \mid x_t] = E[\ell_t(1+\varepsilon_{t+1}) \mid x_t] = \ell_t \\ \hat{y}_{t+2/h} &= E[y_{t+2} \mid x_t] = E[\ell_{t+1}(1+\varepsilon_{t+2}) \mid x_t] = E[\ell_t(1+\varepsilon_{t+1})(1+\varepsilon_{t+2})] = \ell_t \\ \\ \dots \\ \hat{y}_{t+h/h} &= E[y_{t+h} \mid x_t] = \ell_t \\ \end{aligned}$$

$$v_{t+1|t} &= \text{var}(y_{t+1} \mid x_t) = \text{var}(\ell_t(1+\varepsilon_{t+1}) \mid x_t) = \ell_t^2 \sigma^2 \\ v_{t+2|t} &= \text{var}(y_{t+2} \mid x_t) = \text{var}(\ell_{t+1}(1+\varepsilon_{t+2}) \mid x_t) = \text{var}(\ell_t(1+\alpha\varepsilon_{t+1})(1+\varepsilon_{t+2}) \mid x_t] \\ &= \ell_t^2 \text{var}[(1+\alpha\varepsilon_{t+1})(1+\varepsilon_{t+2})] \\ &= \ell_t^2 \text{var}[(1+\alpha\varepsilon_{t+1})(1+\varepsilon_{t+2})] \\ &= \ell_t^2 (\alpha^2 var(\varepsilon_{t+1}) + var(\varepsilon_{t+2}) + \alpha^2 var(\varepsilon_{t+1}\varepsilon_{t+2}) + 2\alpha^2 \text{cov}(\varepsilon_{t+1}, \varepsilon_{t+1}\varepsilon_{t+2}) + 2\alpha \text{cov}(\varepsilon_{t+2}, \varepsilon_{t+1}\varepsilon_{t+2})] \\ &= \ell_t^2 (\alpha^2 va^2 + \sigma^2 + \alpha^2 \sigma^2 \sigma^2) \\ &= \ell_t^2 (1+\alpha^2 \sigma^2)(1+\sigma^2) - 1] \end{aligned}$$
Exercise 2.3

> (bonds.ets <-ets(bonds))

ETS(A, Ad, N)

Call: ets(y = bonds)

Smoothing parameters: alpha = 0.9999 beta = 0.1608 phi = 0.8

Initial states: 1 = 5.5163 b = 0.2967 signs: 0.2394

Alic Alc Bic Elc Elc 256.6683 270.3056 > (usnet.ets <-ets(usnetelec)) ETS(M, M, N)

Call: eta(y = usnetelec)

Smoothing parameters: alpha = 0.9999 beta = 1e-04 phi = 0.9638 Initial states: 1 = 262.6421

```
sigma: 0.0236
     AIC
           AICc
628.1943 629.4188 638.2310
> (ukc.ets <- ets(ukcars))</pre>
ETS(A,N,A)
Call:
 ets(y = ukcars)
  Smoothing parameters:
    alpha = 0.6267
    gamma = 2e-04
  Initial states:
    1 = 338.4757
    s=-0.5313 -45.3246 20.6084 25.2476
  sigma: 25.3264
    AIC
            AICc
1276.592 1277.385 1292.957
> (visit.ets <- ets(visitors))</pre>
ETS(M,A,M)
Call:
 ets(y = visitors)
  Smoothing parameters:
    alpha = 0.6244
    beta = 1e-04
    gamma = 0.1832
  Initial states:
   1 = 86.3534
    b = 2.0306
    s=0.942 1.076 1.0515 0.9568 1.3621 1.1157
           1.011 0.8294 0.9336 1.0017 0.8649 0.8554
  sigma: 0.0515
```

AIC

AICc

2598.193 2600.632 2653.883

BIC

# Exercise 2.4

### > forecast(bonds.ets,h=4,level=80)

```
      Jun
      2004
      Forecast
      Lo 80
      Hi 80

      Jun
      2004
      4.791887
      4.485047
      5.098727

      Jul
      2004
      4.865425
      4.364764
      5.366085

      Aug
      2004
      4.924255
      4.266113
      5.582396

      Sep
      2004
      4.971319
      4.172657
      5.769980
```

### > forecast(usnet.ets,h=4,level=80)

	Point	Forecast	Lo 80	Hi 80
2004		3905.517	3789.206	4026.131
2005		3963.887	3793.446	4133.019
2006		4020.971	3804.828	4224.363
2007		4076.769	3829.443	4314.189

## > forecast(ukc.ets,h=4,level=80)

		${\tt Point}$	Forecast	Lo 80	Hi 80
2005	Q2		426.8056	394.3485	459.2626
2005	QЗ		360.8705	322.5657	399.1753
2005	Q4		405.6569	362.2828	449.0310
2006	Q1		431.4437	383.5363	479.3510

## > forecast(visit.ets,h=4,level=80)

		${\tt Point}$	${\tt Forecast}$	Lo 80	Hi 80
May	2005		361.1182	337.3021	384.9342
Jun	2005		396.1179	365.3447	426.8912
Jul	2005		494.4950	451.0785	537.9114
Aug	2005		428.0406	386.6065	469.4748