

$H_i \Downarrow$

ETS(222) model
 | L seasonality
 | trend
 | over

ETS(AAA) model

ARIMA

source of randomness
 $u_t \sim N(0; \sigma^2)$
 indep.

$$y_t = c + \alpha_1 y_{t-1} + \alpha_2 y_{t-2} + \dots + u_t + \alpha_1 u_{t-1} + \alpha_2 u_{t-2} + \dots$$

Google: Forecasting principles and practice by Hyndman.
 \rightarrow ETS models

$u_t \sim N(0; \sigma^2)$ independent

$$y_t = l_{t-1} + b_{t-1} + s_{t-12} + u_t$$

seas. effect

$$s_t = s_{t-12} + \gamma \cdot u_t$$

y_t — observed
 l_t, b_t, s_t are
 unknown

level/
 deseasoned y_t /
 smoothed y_t /
 trend

$$l_t = l_{t-1} + b_{t-1} + \alpha \cdot u_t$$

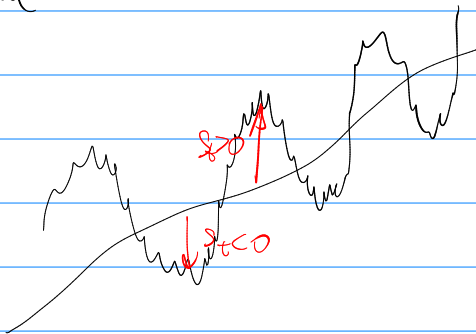
+ initial conditions:

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

$b_0, l_0, s_0, s_1, s_2, s_3, \dots, s_{11}$

+ 1 const.

local slope of trend



Q. How many free param-s
 do we have?

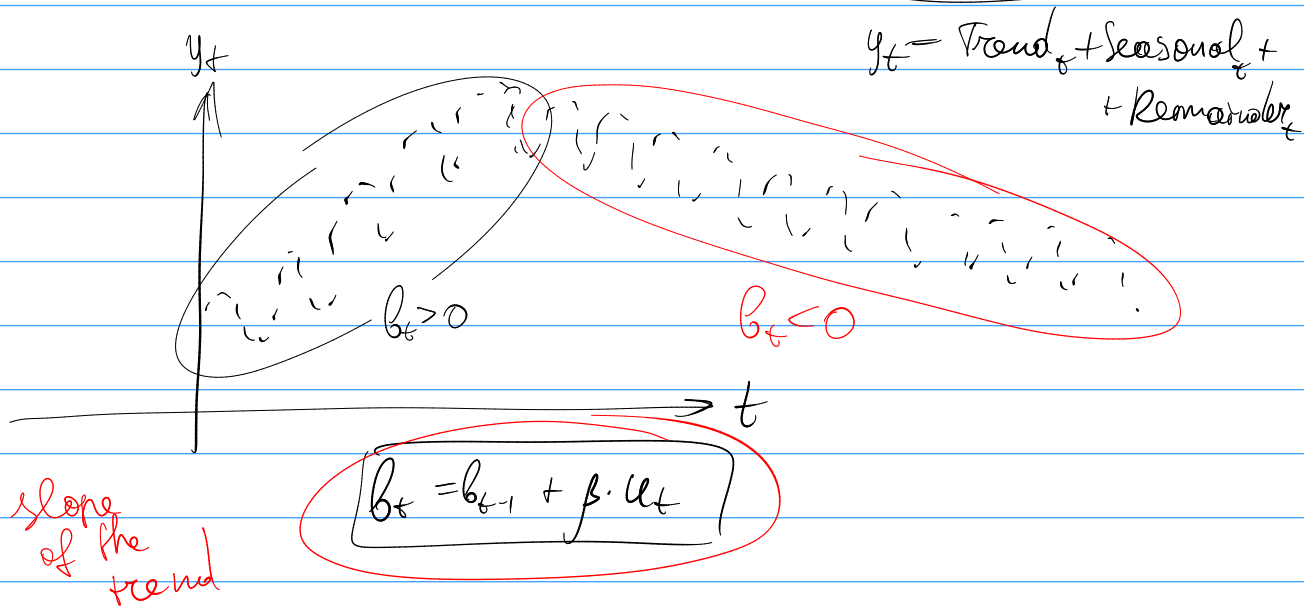
$\sigma^2, \alpha, \beta, \gamma, b_0, l_0, s_0, s_1, s_2, \dots, s_{11}$
 s.t. $s_0 + s_1 + s_2 + \dots + s_{11} = 0$

A. 17 free param-s (18 param-s
 — one const.)

To estimate ETS model we need $\geq 4-5$ years.

ARIMA model : predictions

ETS model : predictions + decomposition.



Problem 1
Parameters:
(and decomp-n)

par-S

$$\alpha = \frac{1}{2} \quad \beta = \frac{1}{2} \quad \gamma = \frac{1}{2}$$

decomp.

$$l_{100} = 20$$

$$b_{100} = 1$$

$$s_{83} = +2$$

$$s_{90} = -1$$

par

$$\sigma^2 = 16$$

95% predictive interval for y_{101}, y_{102} ?

$$y_{101} = \underbrace{l_{100} + b_{100} + s_{83}}_{\text{known at time } t=100} + u_{101}$$

$F_t = \mathcal{F}(y_t, y_{t-1}, y_{t-2}, \dots)$

indep of what we know at time $t=100$

$u_t \sim N(0, \sigma^2)$ independent

$$y_t = l_{t-1} + b_{t-1} + s_{t-12} + u_t$$

$$s_t = s_{t-12} + \gamma \cdot u_t$$

$$l_t = l_{t-1} + b_{t-1} + \alpha \cdot u_t$$

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

$$E(y_{101} | F_{100}) = l_{100} + b_{100} + s_{83} + 0 = 20 + 1 + 2 = 23$$

$$E(u_{101} | F_{100}) = E(u_{101}) = 0$$

$$\text{Var}(y_{101} | F_{100}) = \text{Var}(l_{100} + b_{100} + s_{83} + u_{101} | F_{100})$$

$$= \text{Var}(u_{101} | F_{100}) = \text{Var}(u_{101}) = \sigma^2 = 16$$

over known

PI : $[23 - 1.96\sqrt{16} ; 23 + 1.96\sqrt{16}]$

future = past (known) part + random (unpredict.) part

$$y_{102} = \underbrace{l_{101}}_{\text{past}} + \underbrace{b_{101}}_{\text{past}} + s_{90} + u_{102} =$$

$$= (l_{100} + b_{100} + \alpha \cdot u_{101}) + (b_{100} + \beta \cdot u_{101}) + s_{90} + u_{102} =$$

$$= \underbrace{(l_{100} + 2b_{100} + s_{90})}_{\text{predictable at time 100}} + \underbrace{(\alpha + \beta) \cdot u_{101} + u_{102}}_{\text{independent of what we know at time 100}}$$

$$E(y_{102} | F_{100}) = l_{100} + 2b_{100} + s_{90} = 20 + 2 \cdot 1 - 1 = 21$$

$$\text{Var}(y_{102} | F_{100}) = \text{Var}((\alpha + \beta)u_{101} + u_{102}) = (\alpha + \beta)^2 \sigma^2 + \sigma^2 =$$

$$= \left(\frac{1}{2} + \frac{1}{2}\right)^2 \cdot 16 + 16 = 32$$

$$(y_{102} | F_{100}) \sim N(l_{100} + 2b_{100} + s_{90}, (\alpha + \beta)^2 \sigma^2 + \sigma^2)$$

y_{102} PI: $[21 - 1.96\sqrt{32}; 21 + 1.96\sqrt{32}]$

3 tasks:

data → estimate parameters.

→ data + est. par-s → decomposition of y_t

decomp + estim. of par-s → forecasts

$$\alpha = \frac{1}{2} \quad \beta = \frac{1}{2} \quad \gamma = \frac{1}{2} \quad l_0 = 10 \quad b_0 = 2$$

$$s_0 = s_1 = s_2 = \dots = s_5 = -3$$

$$s_6 = s_7 = \dots = s_{11} = +3$$

t	y_t	l_t	b_t	s_t	u_t
t=0	NA	(10)	(2)	-3	NA
t=1	(12)	10.5	0.5	1.5	(-3)
t=2	14	11	0.5	3	0
t=3	10				
⋮	⋮				

① $u_t = y_t - l_{t-1} - b_{t-1} - s_{t+2}$
 $u_1 = y_1 - l_0 - b_0 - s_{1+2} = 12 - 10 - 2 - (-3) = 3$

② $l_t = l_0 + b_0 + \alpha \cdot u_1 = 12 - \frac{3}{2}$
 $b_t = b_0 + \beta \cdot u_1 = 2 - \frac{3}{2} = \frac{1}{2}$
 $s_1 = s_{11} + \gamma \cdot u_1 = 3 + \frac{1}{2}(-3) = 1.5$

Max Likelihood

observables
 $y = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_T \end{pmatrix}$

parameters

$\theta = \begin{pmatrix} \alpha \\ \beta \\ b_0 \\ s_0 \\ s_1 \\ \vdots \\ s_{10} \end{pmatrix}$

$$\ln L = \ln f(y_1, y_2, \dots, y_T) = \ln L(y, \theta)$$

$$= \ln f(y_1) + \ln f(y_2 | y_1) + \ln f(y_3 | y_2, y_1) + \dots + \ln f(y_T | y_{T-1}, y_{T-2}, \dots)$$

$$s_{-11} = 0 - \sum_{t=10}^0 \alpha_t$$

$$(y_t | y_{t-1}, y_{t-2}, \dots) \sim N(l_t + b_t + s_{t-12}; \sigma^2) \rightarrow \max_{\theta}$$

$$N(\mu, \sigma^2) \Rightarrow \ln f = \ln \frac{1}{\sqrt{2\pi\sigma^2}} - \frac{1}{2} \exp\left(-\frac{1}{2\sigma^2} (y_t - \mu)^2\right) = -\frac{1}{2} \ln(2\pi\sigma^2) - \frac{1}{2} \exp\left(-\frac{1}{2\sigma^2} (y_t - \mu)^2\right)$$

RHS: param-s

$$\ln f(y_1) = -\frac{1}{2} \ln(2\pi\sigma^2) - \frac{1}{2} \exp\left(-\frac{1}{2\sigma^2} (y_1 - (l_0 + b_0 + s_{-11}))^2\right)$$

RHS: param-s and y,

$$\ln f(y_2 | y_1) = -\frac{1}{2} \ln(2\pi\sigma^2) - \frac{1}{2} \exp\left(-\frac{1}{2\sigma^2} (y_2 - (l_1 + b_1 + s_{10}))^2\right)$$

$$\text{where } l_1 = l_0 + b_0 + \alpha \cdot \tilde{u}_1 =$$

$$= l_0 + b_0 + \alpha (y_1 - l_0 - b_0 - s_{-11})$$

$$b_1 = b_0 + \beta \cdot u_1 = b_0 + \beta (y_1 - l_0 - b_0 - s_{-11})$$

$$\ln f(y_3 | y_2, y_1) = -\frac{1}{2} \ln(2\pi\sigma^2) - \frac{1}{2} \exp\left(-\frac{1}{2\sigma^2} (y_3 - (l_2 + b_2 + s_{-9}))^2\right)$$

$$l_2 = l_1 + b_1 + \alpha \cdot u_2 =$$

$$= l_1 + b_1 + \alpha (y_2 - l_1 - b_1 - s_{10}) =$$

$$= \dots = \dots = \text{only } l_0, b_0, s_0, s_1, s_2, \dots, y_1, y_2$$

One may use AKAIKE inf. crit.

$$AIC = -2 \ln L + 2 \cdot k$$

max of log likelihood

number of free parameters.
ETS(AAA)?

$$k = 17$$

ETS
ETS(AAN)

No season.
Add trend
Add error.

ETS(ANA)

No trend
Additive error
Additive season - γ

Ex
how many free parameters?

$S_t = 0$
AAN: $\begin{bmatrix} \gamma^2, \alpha, \beta \\ l_0, b_0 \end{bmatrix}$
 $k = ?$ **5**

$b_t = 0$
ANA: $\begin{bmatrix} \gamma^2, \alpha, \beta \\ l_0, b_0 \end{bmatrix}$
 $k = ?$ **15**

$u_t \sim N(0; \gamma^2)$ independent
 $y_t = l_{t-1} + b_{t-1} + s_{t-12} + u_t$

y_t — observed
 l_t, b_t, s_t are unknowns

$s_t = s_{t-12} + \gamma \cdot u_t$

$l_t = l_{t-1} + b_{t-1} + \alpha u_t$

+ initial conditions:

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

$b_0, l_0, s_0, s_1, s_2, s_3, \dots, s_{11}$
+ 1 const.