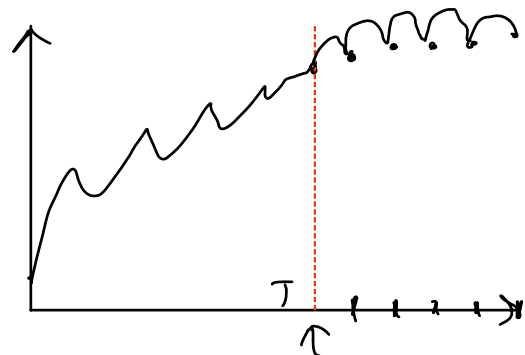


Смещение

1) Регрессионная.

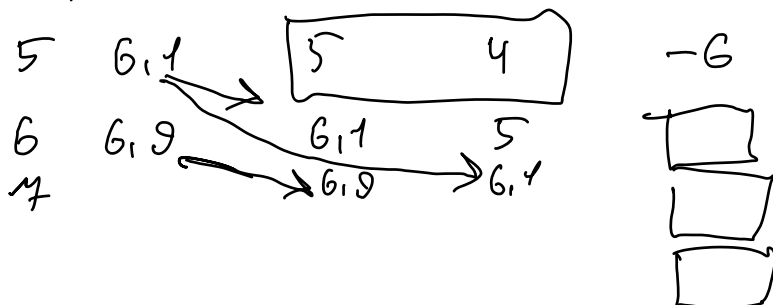
forecast horizon = 5



$$\hat{y}_{T+h|T} = \hat{f}(\underbrace{\tilde{y}_{T+h-1}, \dots, \tilde{y}_{T+h-k+1}}_{k \text{ ларов в регрессорах}})$$

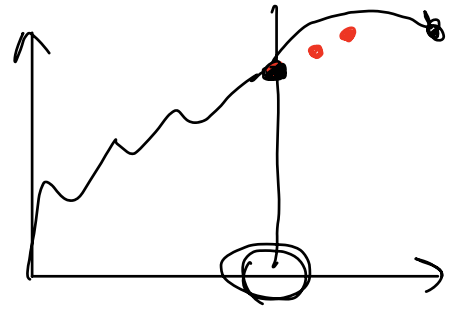
$$\tilde{y}_t = \begin{cases} y_t, & t \leq T \\ \hat{y}_t, & t > T \end{cases}$$

t	y_t	y_{t-1}	y_{t-2}	x_{t-1}
0	1	—	—	-1
1	2	1	—	-2
2	3	2	1	-3
3	4	3	2	-4
4	5	4	3	-5



Direct

$$\hat{y}_{T+h} = \hat{f}_h(y_T, \dots, y_{T-k+1})$$



$h=1$

t	y_t	y_{t-1}	y_{t-2}	x_{t+1}
0	1	—	—	
1	2	1	—	
2	3	2	1	
3	4	3	2	
4	5	4	3	—5

$h=2$

t	y_t	y_{t-2}	y_{t-3}
0	1	—	—
1	2	—	—
2	3	1	—
3	4	2	1
4	5	3	2

Dirvec

Одноступенчатая модель марковской цепи

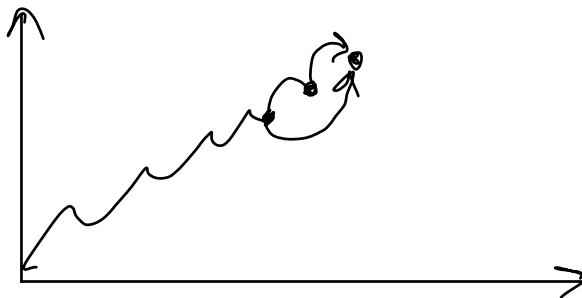
$h=1$

t	y_t	y_{t-1}	y_{t-2}	x_{t+1}
0	1			
1	2	1	—	
2	3	2	1	
3	4	3	2	
4	5	4	3	—5

→ \hat{y}_t

$h=2$

t	y_t	y_{t-2}	y_{t-3}
0	1	—	—
1	2	—	—
2	3	1	—
3	4	2	1
4	5	3	2



4) MIMO

$$\underbrace{[y_{T+h}, \dots, y_{T+1}]} = F(y_T, \dots, y_{T-k+1})$$

5) DirMo

$$h = \underbrace{[1, 2, 3, 4, 5]}_{\text{MIMO}} \parallel \underbrace{[6, 7, 8, 9, 10]}_{\text{MIMO}}$$

ETS

$$(y_t)_{t=1}^T$$

~ 1940

$$\hat{y}_{T+1|T} = y_T$$

$$\hat{y}_{T+h|T} = y_T$$

$$\hat{y}_{T+1|T} = \frac{1}{T} \sum_{t=1}^T y_t$$

$$\hat{y}_{T+1|T} = \sum_{i=0}^{T-1} \alpha q^i y_{T-i}$$

$$\alpha \in [0; 1]$$

$$\alpha + \alpha q + \alpha^2 q^2 + \dots + q^{T-1} \alpha = 1$$

$$\frac{\alpha(q^T - 1)}{q - 1} = 1$$

$$\frac{\alpha}{1-\alpha} = 1 \Rightarrow \alpha = 1-\alpha$$

$$\alpha = 1-\alpha$$

$$\hat{y}_{T+1|T} = \sum_{i=0}^{T-1} \alpha(1-\alpha)^i y_{T-i}$$

$$\hat{y}_{T+h|T} = \left(\sum_{i=0}^{T-1} \alpha(1-\alpha)^i y_{T-i} \right)$$

$$\sum_{t=1}^T (y_t - \hat{y}_{t|t-1})^2 \rightarrow \min_{\alpha}$$

$$\begin{aligned} \hat{y}_{T+1|T} &= \alpha y_T + (1-\alpha) \hat{y}_{T|T-1} = \\ &= \alpha (y_T - \hat{y}_{T|T-1}) + \hat{y}_{T|T-1} = \\ &= \alpha e_T + \hat{y}_{T|T-1} \end{aligned}$$

модель коррекции ошибки

VECM

$$\begin{aligned} \hat{y}_{T+1|T} &= \alpha y_T + \alpha(1-\alpha) y_{T-1} + \alpha(1-\alpha)^2 y_{T-2} \dots \\ &= \alpha y_T + (1-\alpha) [\alpha y_{T-1} + \alpha(1-\alpha) y_{T-2}] = \\ &= \boxed{\alpha y_T} + \boxed{(1-\alpha) \hat{y}_{T|T-1}} \quad \hat{y}_{T|T-1} \end{aligned}$$

$$t \in \overline{1, T} \quad \left\{ \hat{y}_{T+1|T} = \alpha y_T + (1-\alpha) \hat{y}_{T|T-1} \right.$$

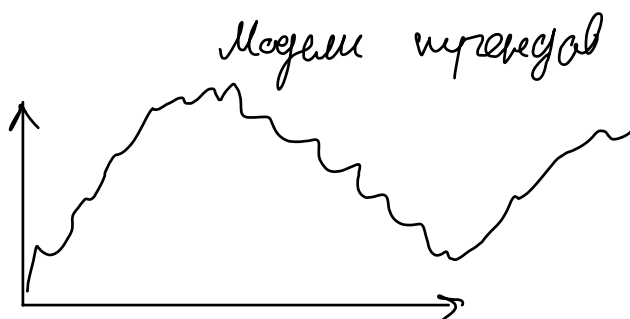
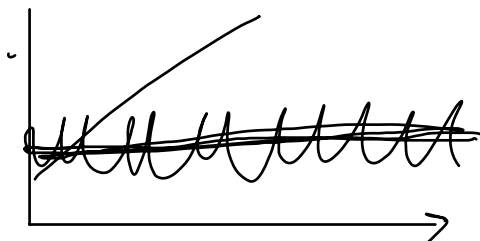
$$\hat{y}_{2|1} = \alpha y_1 + (1-\alpha) \underline{l_0}$$

$$\hat{y}_{T+1|T} = \sum_{i=0}^{T-1} \alpha(1-\alpha)^i y_{T-i} + (1-\alpha)^T l_0$$

$$\sum_{t=1}^T (y_t - \hat{y}_{t|t-1})^2 \rightarrow \min_{\alpha, l_0}$$

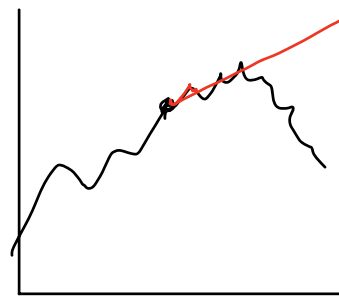
Каким образом

$$\begin{aligned} y_t \text{ - прогноза } \hat{y}_{t+1|t} &= l_t \\ y_t \text{ - фактический } l_t &\rightarrow = \alpha y_t + (1-\alpha) \hat{l}_{t-1} \end{aligned}$$



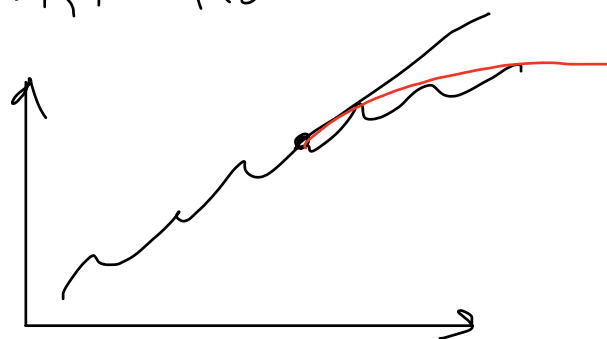
b_t - локальная скорость роста

$$\begin{cases} \hat{y}_{t+h} = l_t + h b_t \\ l_t = \alpha y_t + (1-\alpha)(l_{t-1} + b_{t-1}) \\ b_t = \beta(l_t - l_{t-1}) + (1-\beta)b_{t-1} \end{cases}$$



$$\frac{y_t - y_{t-1}}{l_t - l_{t-1}}$$

$$\alpha, \beta, b_0, l_0$$



$$\hat{y}_{t+h|t} = l_t + (\varphi + \varphi^2 + \dots + \varphi^h) b_t$$

$$l_t = \alpha y_t + (1-\alpha)(l_{t-1} + \varphi b_{t-1})$$

$$b_t = \beta(l_t - l_{t-1}) + (1-\beta)\varphi b_{t-1}$$

$$\lim_{h \rightarrow \infty} \hat{y}_{t+h|t} = l_t + \frac{\varphi b_t}{1-\varphi}$$

Сезонность.

$$K = \frac{(h-1)}{m}$$

$$\begin{cases} \hat{y}_{t+h|t} = l_t + h b_t + S_{t+h-m(K+1)} \\ l_t = \alpha(y_t - S_{t-m}) + (1-\alpha)(l_{t-1} + b_{t-1}) \\ b_t = \beta(l_t - l_{t-1}) + (1-\beta)b_{t-1} \end{cases}$$

$$m = 12$$

$$S_t = \lambda(y_t - l_{t-1} - b_{t-1}) + (1-\lambda)(S_{t-m})$$

$\lambda, \beta, \gamma,$

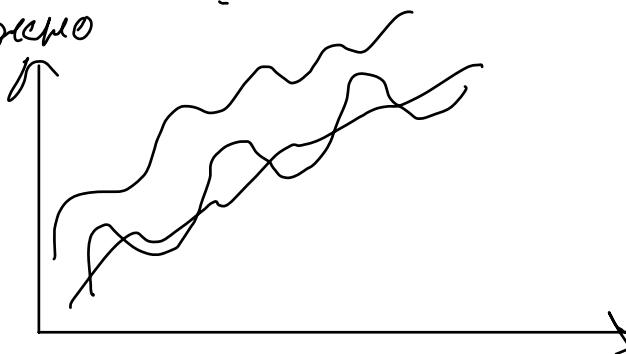
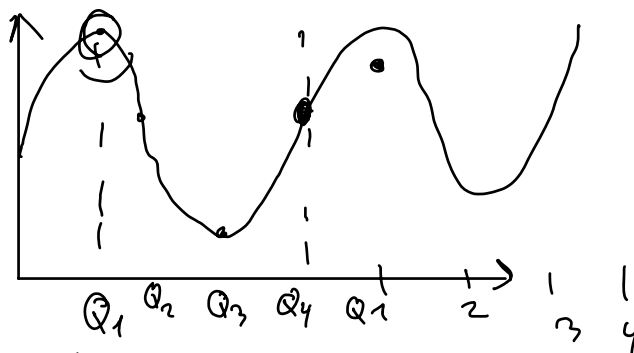
l_0, b_0

S_0, \dots, S_{m-1}

$$S_0 + \dots + S_{m-1} = 0$$

Из уравнений можно
найти одну
параметр \Rightarrow

вычислить $m-1$
параметры.



$$\begin{cases} \hat{y}_{t+1|t} = l_t \\ l_t = \lambda y_t + (1-\lambda) l_{t-1} \end{cases}$$

$$\begin{aligned} l_t &= l_{t-1} + \lambda(y_t - l_{t-1}) = \\ &= l_{t-1} + \lambda e_t \end{aligned}$$

$$e_t = y_t - l_{t-1}$$

$$y_t = l_{t-1} + e_t$$

$$\begin{cases} y_t = l_{t-1} + e_t \sim N(0, \sigma^2) \\ l_t = l_{t-1} + \lambda e_t \end{cases}$$

ETS(AAN)

$$\begin{cases} y_t = l_{t-1} + b_{t-1} + \varepsilon_t \sim N(0, \sigma^2) \\ l_t = l_{t-1} + \alpha \varepsilon_t \\ b_t = b_{t-1} + \beta \varepsilon_t \end{cases} \quad \alpha, \beta, l_0, b_0, \sigma^2$$

$$f(y_T, \dots, y_1) \neq \prod_{t=1}^T f(y_t)$$

$$f(y_T, \dots, y_1)$$

$$f(y_T | y_{T-1}, \dots, y_1) = \frac{f(y_T, \dots, y_1)}{f(y_{T-1}, \dots, y_1)} \Rightarrow$$

← Случ. величины

$$\ominus f(y_T, \dots, y_1) = f(y_T | y_{T-1}, \dots, y_1) \cdot \underbrace{f(y_{T-1}, \dots, y_1)}_{(y_T | y_{T-1} = y_{T-1}, \dots)}$$

$$= f(y_T | y_{T-1}, \dots, y_1) \cdot f(y_{T-1} | \underbrace{y_{T-2}, \dots, y_1}_{\mathcal{F}_{T-2}})$$

$$f(y_{T-2}, \dots, y_1) = \dots \quad \mathcal{F}_{T-2} =$$

$$= \underbrace{\quad \quad \quad} \cdot f(y_2 | y_1) \underbrace{f(y_1)}$$

$$= \prod_{t=1}^T f(y_t | \mathcal{F}_{t-1})$$

$$\begin{cases} y_t = l_{t-1} + b_{t-1} + \varepsilon_t \sim N(0, \sigma^2) \\ l_t = l_{t-1} + b_{t-1} + \alpha \varepsilon_t \\ b_t = b_{t-1} + \beta \varepsilon_t \end{cases} \quad \alpha, \beta, l_0, b_0, \sigma^2$$

$$y_1 = l_0 + b_0 + \varepsilon_1 \sim N(l_0 + b_0, \sigma^2)$$

$$y_2 | y_1 = l_1 + b_1 + \varepsilon_2 =$$

$$= l_0 + b_0 + \alpha \hat{\varepsilon}_1 + b_0 + \beta \hat{\varepsilon}_1 + \varepsilon_2 \quad \Leftrightarrow$$

$$\hat{y}_1 = l_0 + b_0 + \hat{\varepsilon}_1 \Rightarrow \hat{\varepsilon}_1 = \hat{y}_1 - l_0 - b_0$$

$$\Leftrightarrow l_0 + 2b_0 + \alpha(\hat{y}_1 - l_0 - b_0) + \beta(\hat{y}_1 - l_0 - b_0) + \varepsilon_2$$

$$= \underbrace{l_0 + 2b_0}_{\text{}} + \underbrace{(\alpha + \beta)(\hat{y}_1 - l_0 - b_0)}_{\text{}} + \underbrace{\varepsilon_2}_{\text{}}$$

$$y_2 | y_1 \sim N(l_0 + 2b_0 + (\alpha + \beta)(y_1 - l_0 - b_0), \sigma^2)$$

Задача.

$$\varepsilon_t \sim N(0, 1)$$

$$\begin{cases} y_t = l_{t-1} + b_{t-1} + s_{t-2} + \varepsilon_t \\ l_t = l_{t-1} + b_{t-1} + 0.3 \varepsilon_t \\ b_t = b_{t-1} + 0.2 \varepsilon_t \\ s_t = s_{t-2} + 0.1 \varepsilon_t \end{cases}$$

$$S_{100} = 2, \quad S_{99} = -1,9 \quad b_{100} = 0,5, \quad l_{100} = 4$$

$$y_{102}$$

$$y_{101} = l_{100} + b_{100} + S_{99} + \varepsilon_t$$

$$(y_{101} | \mathcal{F}_{100}) \sim \mathcal{N}(l_{100} + b_{100} + S_{99}, 4)$$

2,6

$$E(y_{t+h} | \mathcal{F}_t) - \text{математическое ожидание}$$

$$\text{Var}(y_{t+h} | \mathcal{F}_t) - \text{дисперсия}$$

$$E(y_{101} | \mathcal{F}_{100}) = 2,6$$

95%

$$\text{Var}(y_{101} | \mathcal{F}_{100}) = 4$$

$$[2,6 - 1,96 \cdot 2; 2,6 + 1,96 \cdot 2]$$

$$(y_{102} | \mathcal{F}_{100}) \sim$$

$$y_{102} = l_{101} + b_{101} + \underbrace{(S_{100})}_{0,5} + \varepsilon_{102}$$

$$l_{101} = \underbrace{l_{100} + b_{100}}_{4,5} + 0,3 \varepsilon_{101}$$

$$b_{101} = 0,5 + 0,2 \varepsilon_{101}$$

$$y_{102} | \mathcal{F}_{100} = 4 + 0,5 \varepsilon_{101} + \varepsilon_{102}$$

$$E(y_{102} | \mathcal{F}_{100}) = 4$$

$$y_{102} | \mathcal{F}_{100} \sim \mathcal{N}(4, 5)$$

$$\text{Var}(y_{102} | \mathcal{F}_{100}) = 1 + 4 = 5$$

$$[7 - 1.96 \sqrt{5}; 7 + 1.96 \sqrt{5}]$$