

# Simple 1D KF

$$\begin{cases} x_t = Ax_{t-1} + n_t \\ z_t = Hx_t + n_s \end{cases}$$

# Kalman Filter : Bayes Filter

$$A: n \times n$$

$$H: m \times n$$

predictive

w/ Linear dyn. and measurements starting w Gaussian  $P(x_0)$

$$P(x_t | Z^t)$$

"posterior"

$$\propto P(z_t | x_t) P(x_t | Z^{t-1})$$

filtering density

$$\propto L(x_t; z_t) \int P(x_t | x_{t-1}) P(x_{t-1} | Z^{t-1})$$

likelihood (3)

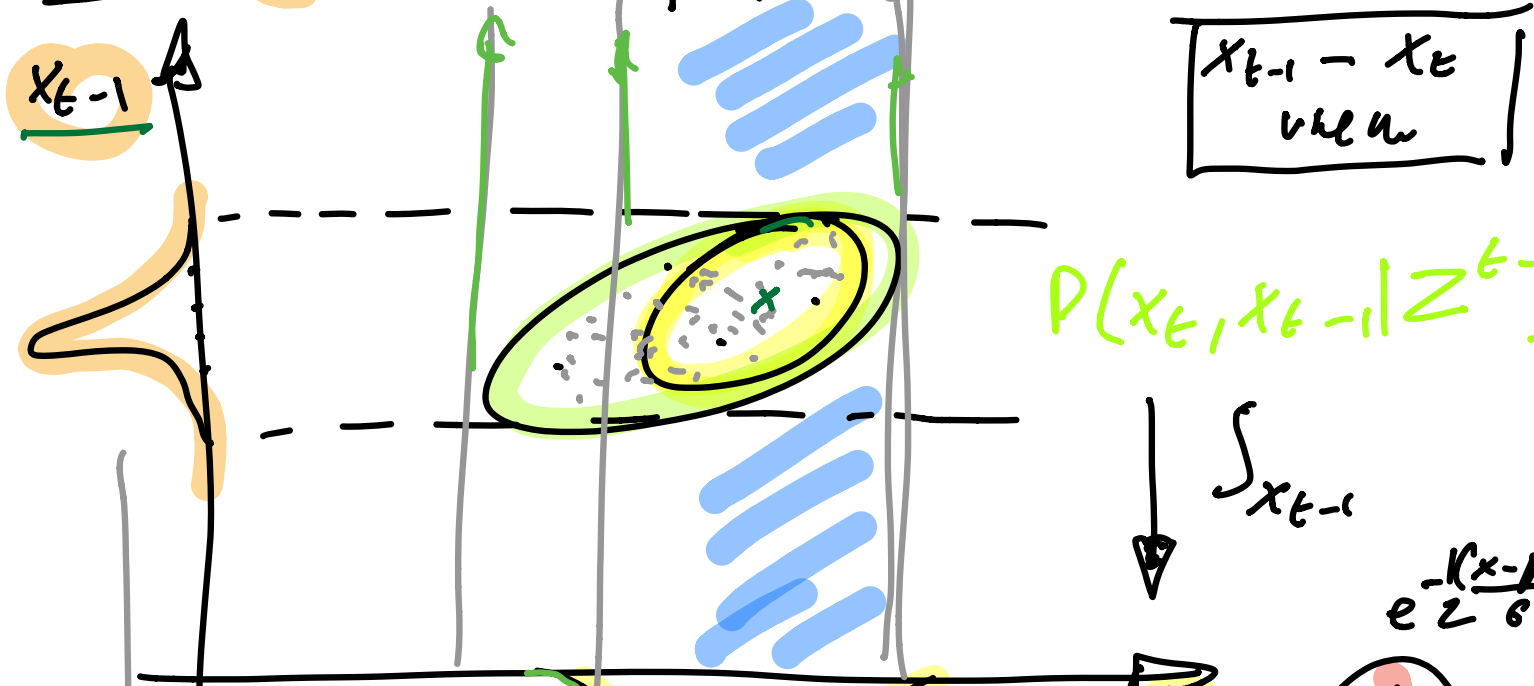
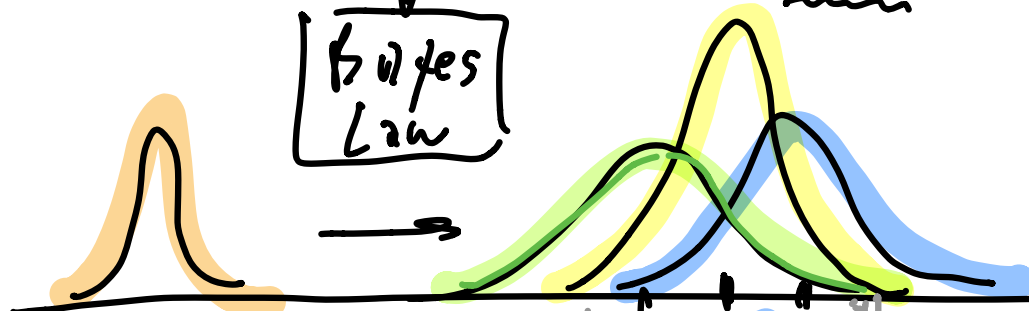
$x_{t-1}$

(2)

(1)

Bayes Law

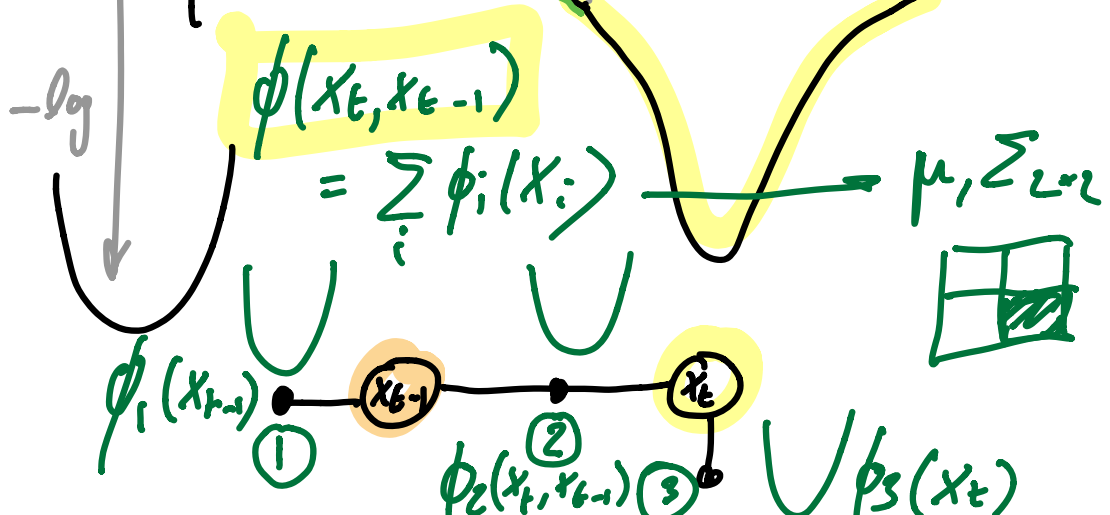
x-view  
1D



$$P(x_t, x_{t-1} | Z^{t-1})$$

$$\int_{x_{t-1}}$$

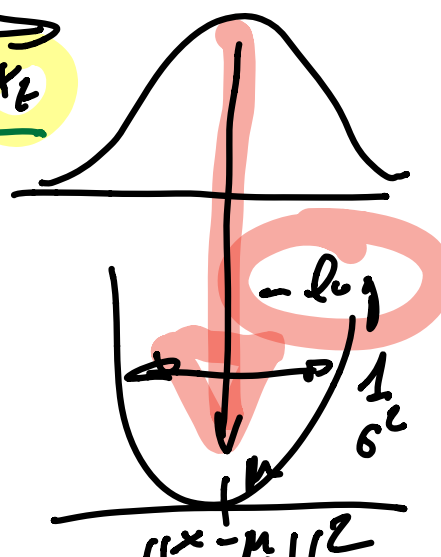
$$e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$



$$\phi(x_t, x_{t-1})$$

$$= \prod_i \phi_i(x_i)$$

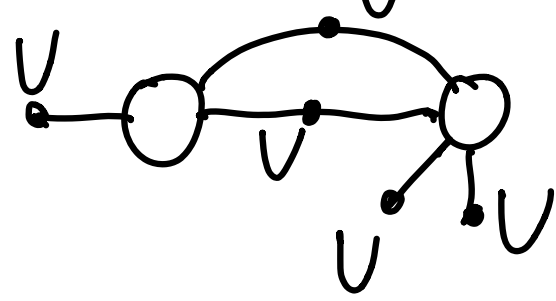
$\mu, \Sigma, \dots$



**LLS**

FACTOR GRAPH

$$\psi(x_t, x_{t-1}) = \prod_i \psi(x_i)$$



$$\phi(x_t, x_{t-1}) = \sum_i \phi(x_i)$$

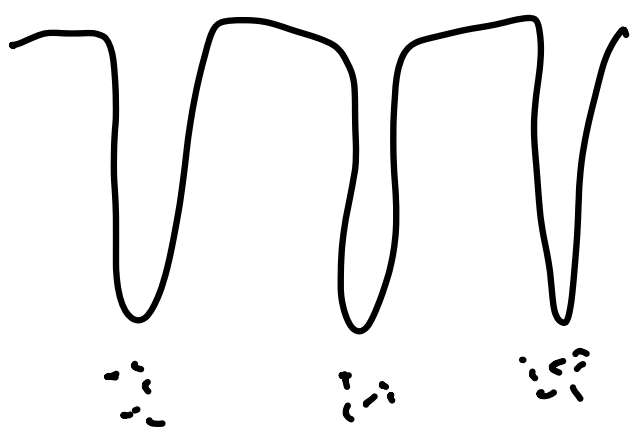
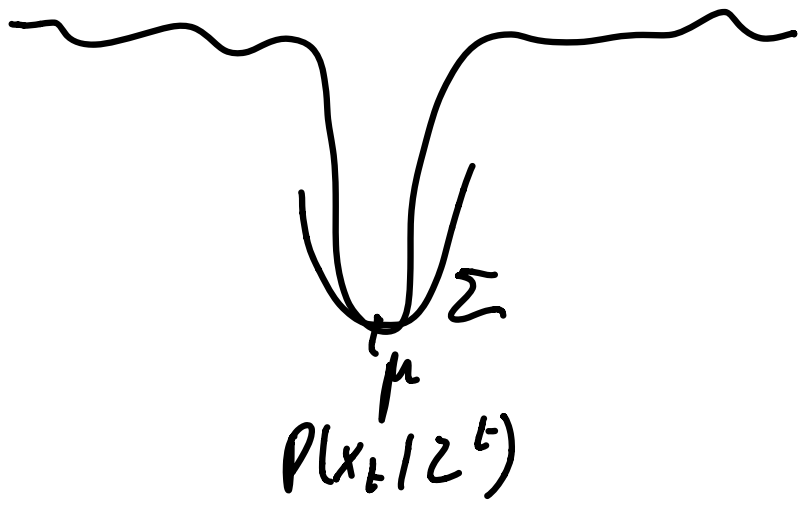
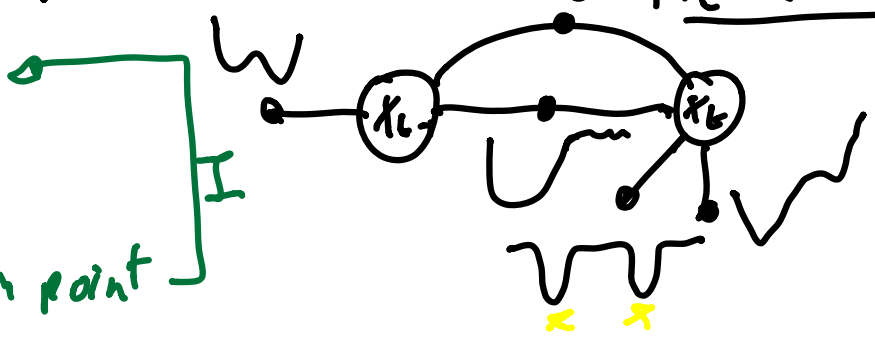
Iterated Extended Kalman Filter

$$\mu_2 \Sigma_{2 \times 2}$$

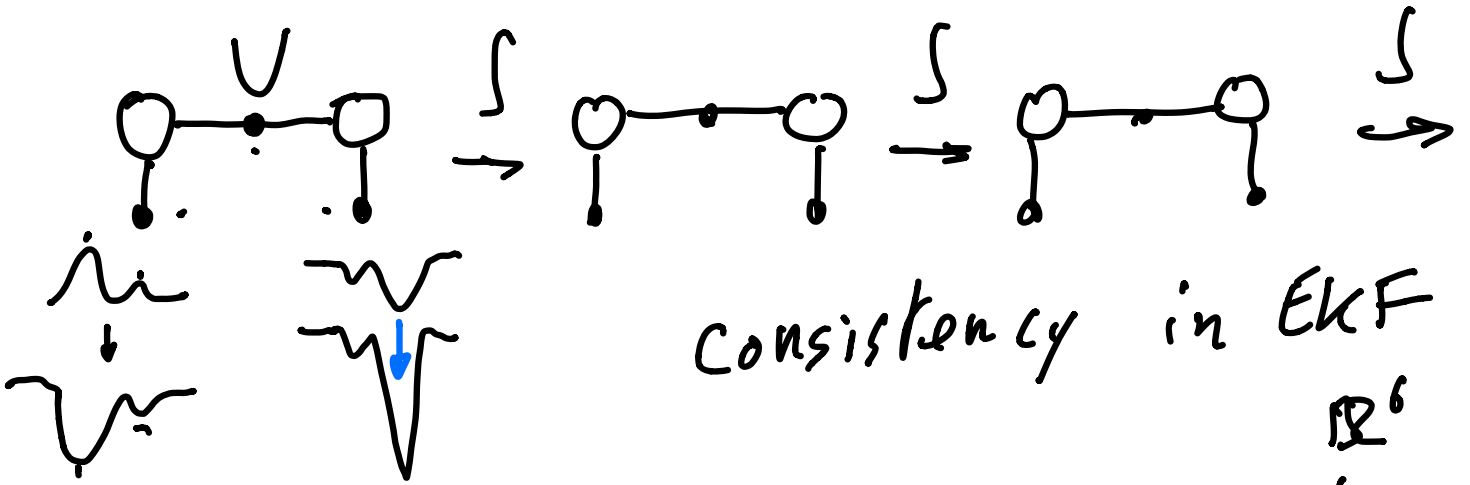
$W K(\Pi(\dots))$

**NLS**

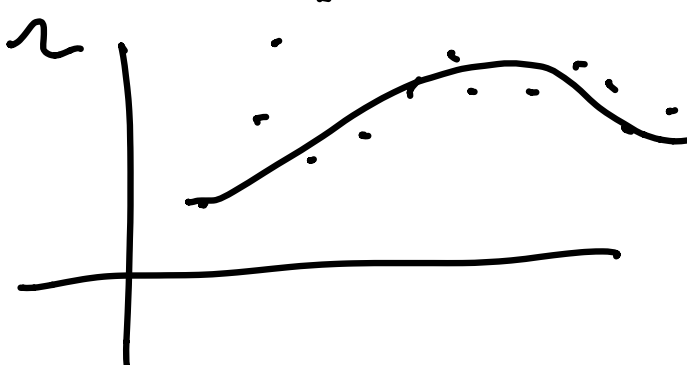
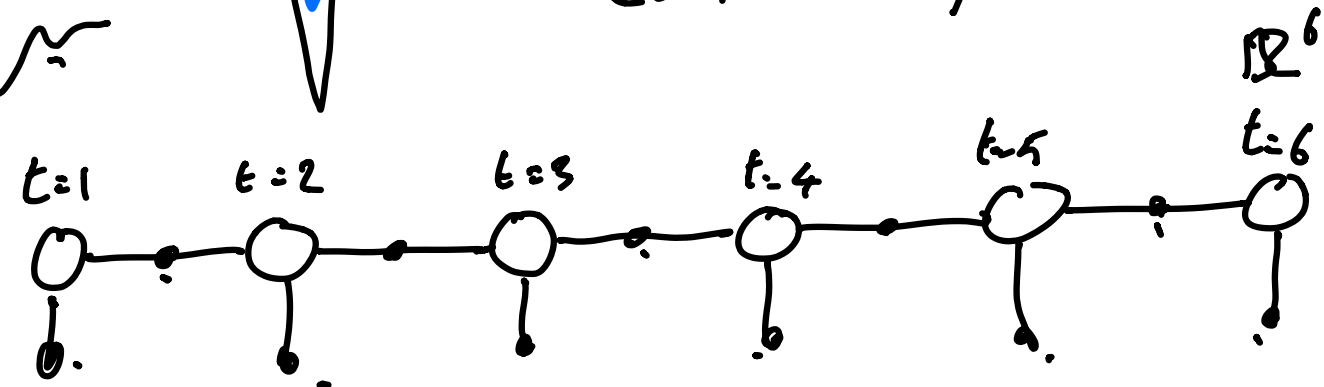
- 1) linearize  $\rightarrow \Delta x$
- 2) LLS  $\Delta x$
- 3) update lin point



Kalman Smoothing  $\leftrightarrow$  Filtering.

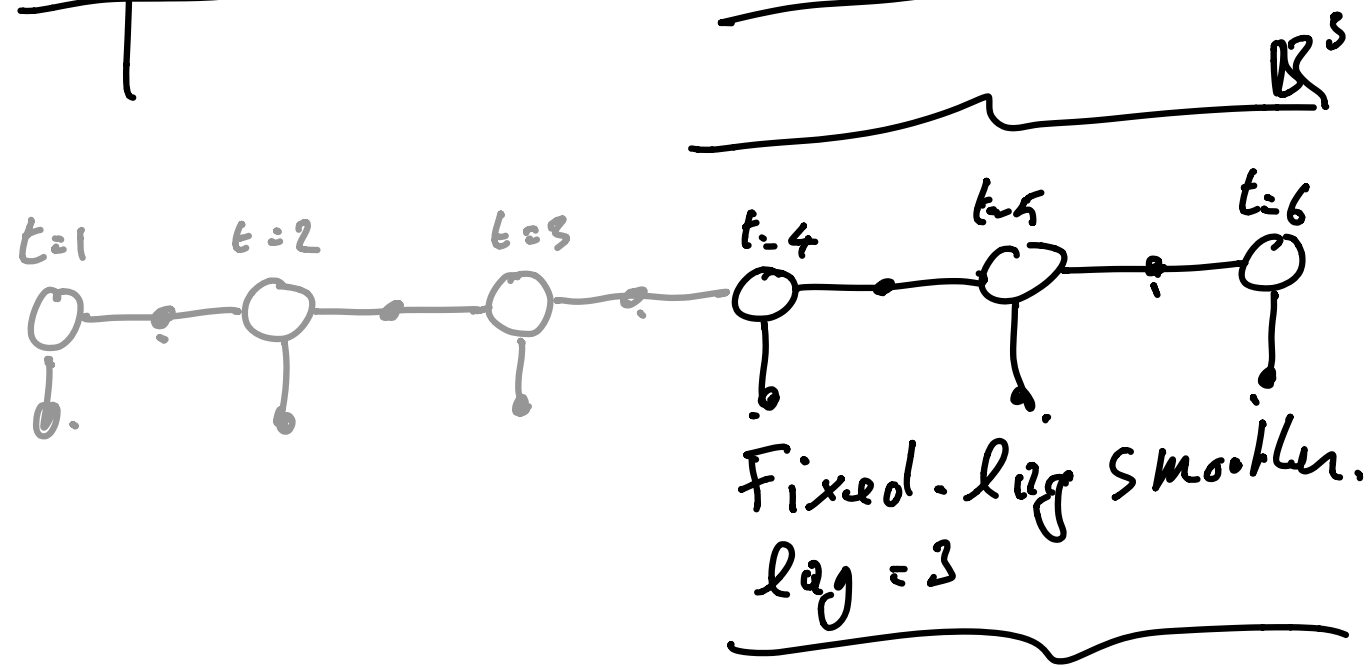


consistency in EKF



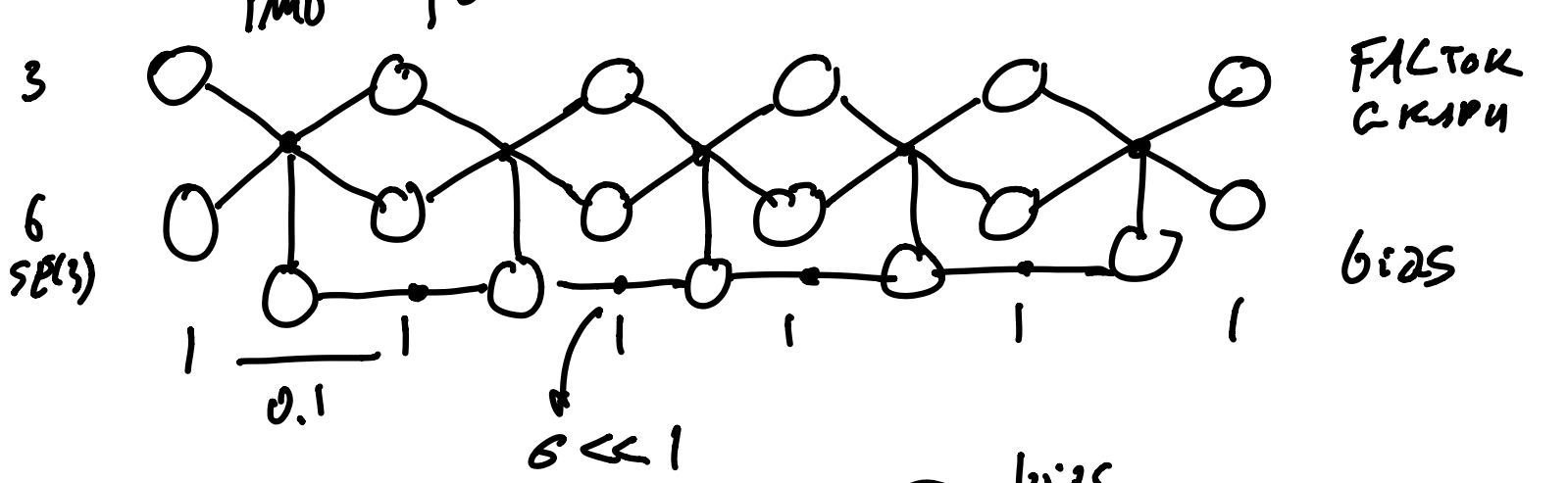
K Smoothing

IEKS  $\equiv$  NLS



$3 \times 6 = 18$

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$$a = \cancel{\ddot{x}} + n$$

$$a = \ddot{x} + \underline{b} + n$$

$$x = \iint a$$

$$e = \frac{1}{2} b_a^2$$

$$\underline{\dot{b}} = e^{-b/\sigma} + n$$

$$\underline{g} = \omega + \beta + n$$

$$\theta = \int \omega$$

$$e = b_g^T$$