

# TRACKS

- \* particle detector
- \* method to combine hits

charged particles

$\pi^+ \pi^-$   
 $e^+ e^-$   
 $\mu^+ \mu^-$   
 $\dots$

Vacuum

trajectory?  $\vec{E} \vec{B}$

Lorentz:  $\vec{F} = q(\vec{E} + \vec{v} \times \vec{B})$

$\vec{F} = \frac{d\vec{p}}{dt}$        $\vec{p} = m_0 \gamma \vec{v} = m \vec{v}$

$\vec{v} = \frac{d\vec{R}}{dt}$

$\Rightarrow \vec{F} = m_0 \gamma \frac{d\vec{v}}{dt} = m_0 \gamma \frac{d^2\vec{R}}{dt^2}$

$\Rightarrow m_0 \gamma \frac{d^2\vec{R}}{dt^2} = e \frac{d\vec{R}}{dt} \times \vec{B}$

$ds = v dt$

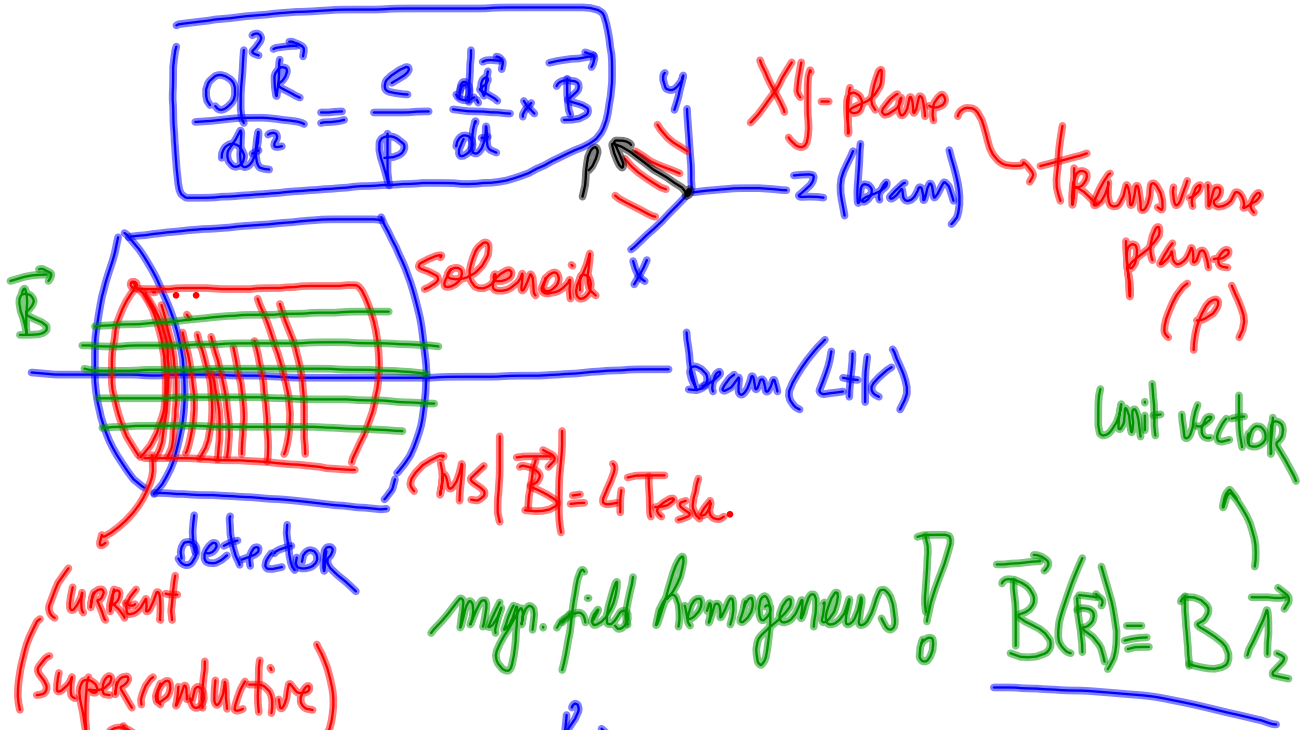
$\Rightarrow dt = \frac{ds}{v}$

$m_0 \gamma \frac{d^2\vec{R}}{ds^2} \frac{ds}{dt} = e \frac{d\vec{R}}{ds} \times \vec{B}$

$|\vec{p}| \equiv p$

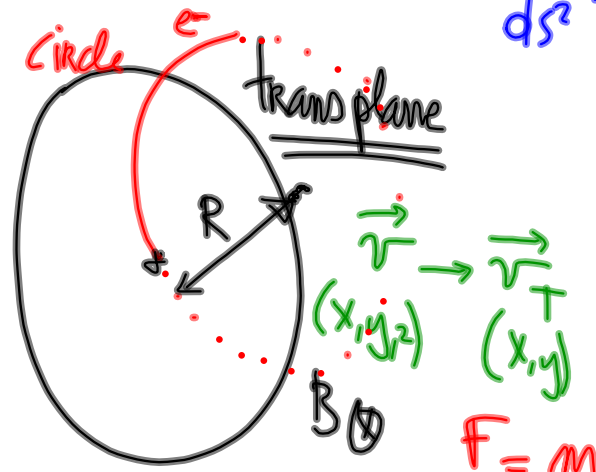
$\frac{d^2\vec{R}}{ds^2} = \frac{e}{p} \cdot \frac{d\vec{R}}{ds} \times \vec{B}(\vec{R})$

$\rightarrow$  diff. eq.



(CURRENT  
Superconductive)  
 $R=0$

$$\frac{d^2 \vec{R}}{ds^2} = \frac{e}{p} \frac{d\vec{R}}{ds} \times B \vec{1}_z$$



$$\vec{F} = q \vec{v}_T \times \vec{B} = \frac{d\vec{p}_T}{dt}$$

centripetal force  $\frac{F = m v^2}{R}$

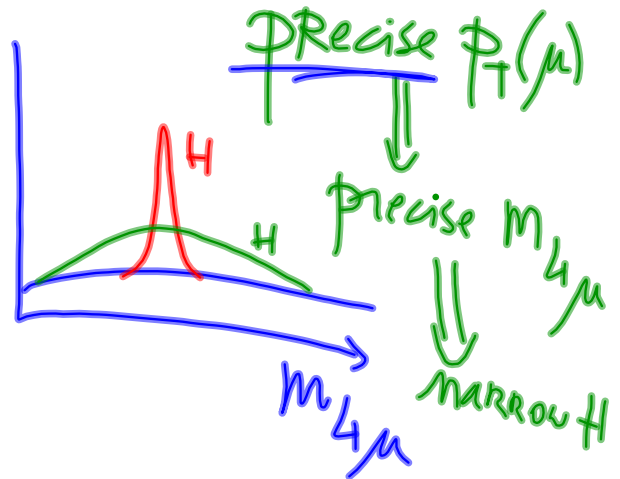
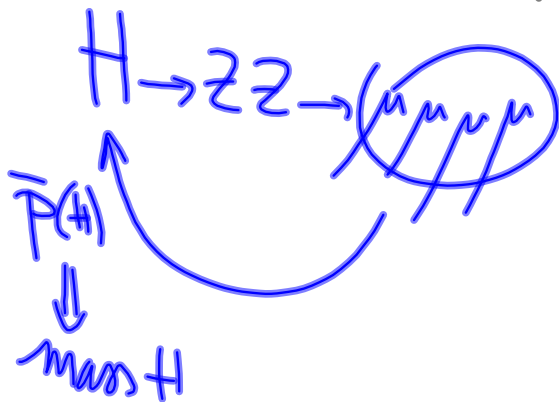
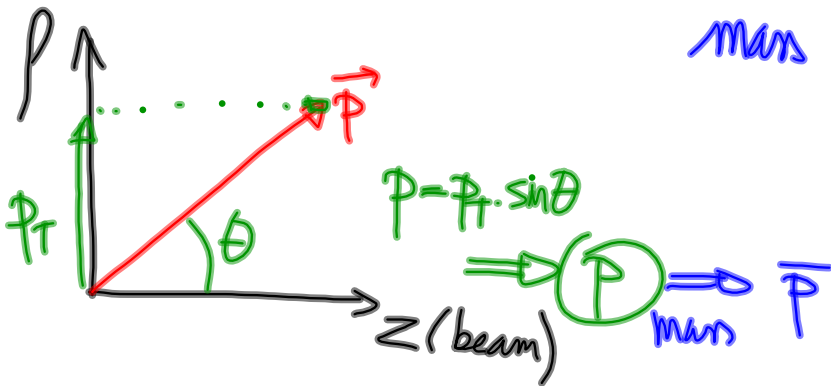
$$\Rightarrow q v_T \cdot B = \frac{m v_T^2}{R}$$

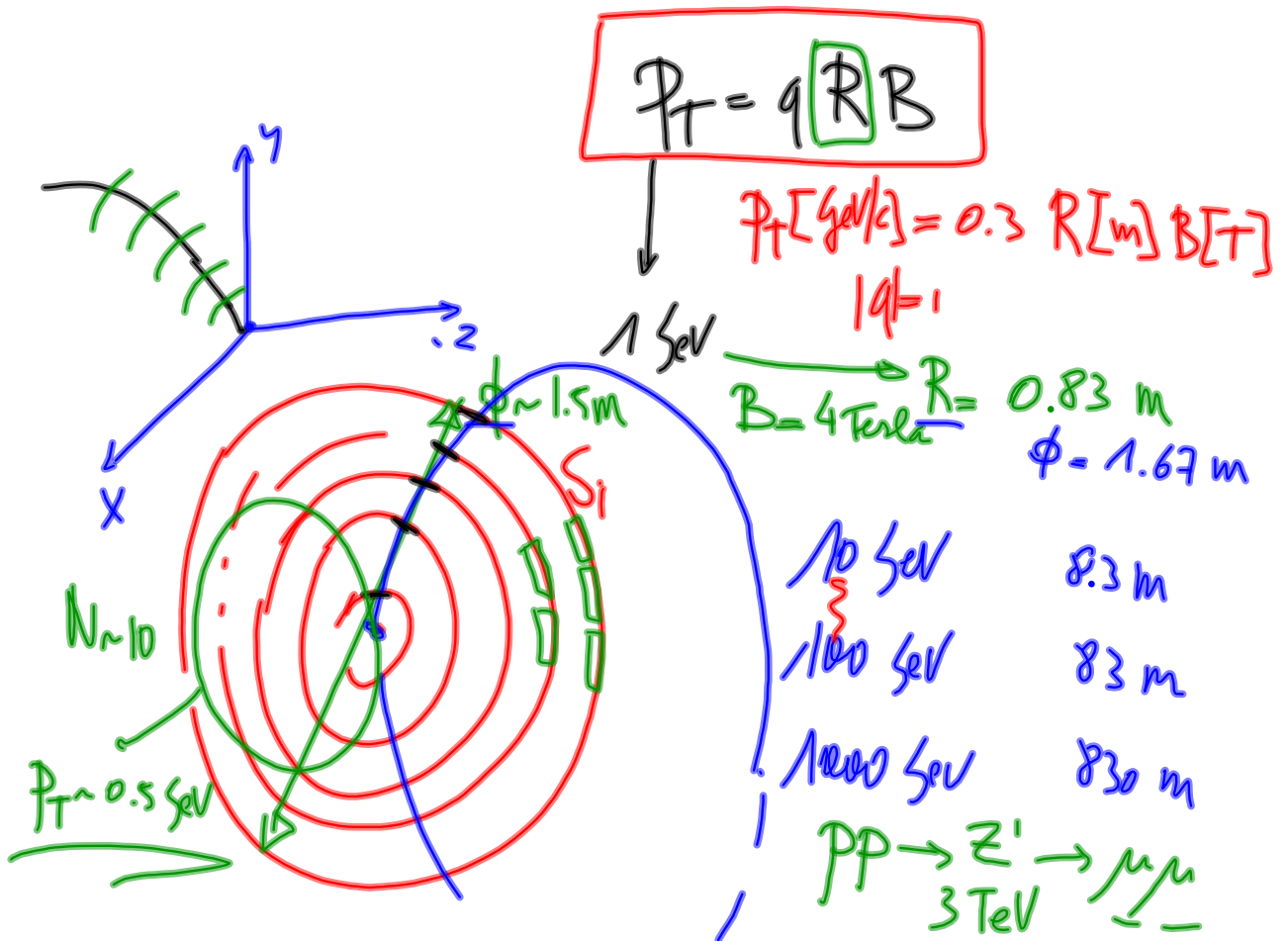
$$R = \frac{m v_T}{q B} \Rightarrow \boxed{p_T = q R B}$$

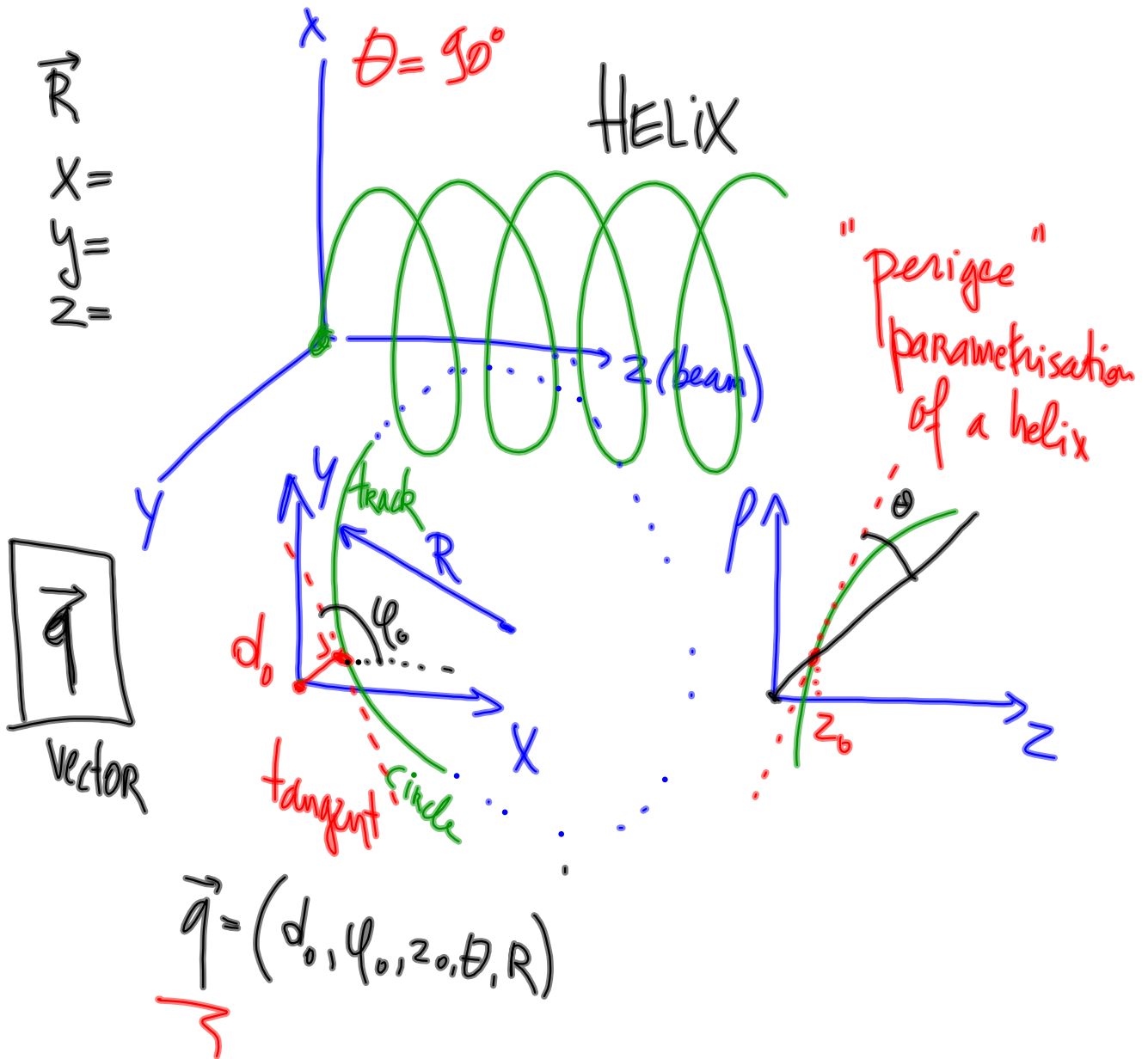
$$\mathbb{P}_T = q \vec{R} \cdot \vec{B}$$

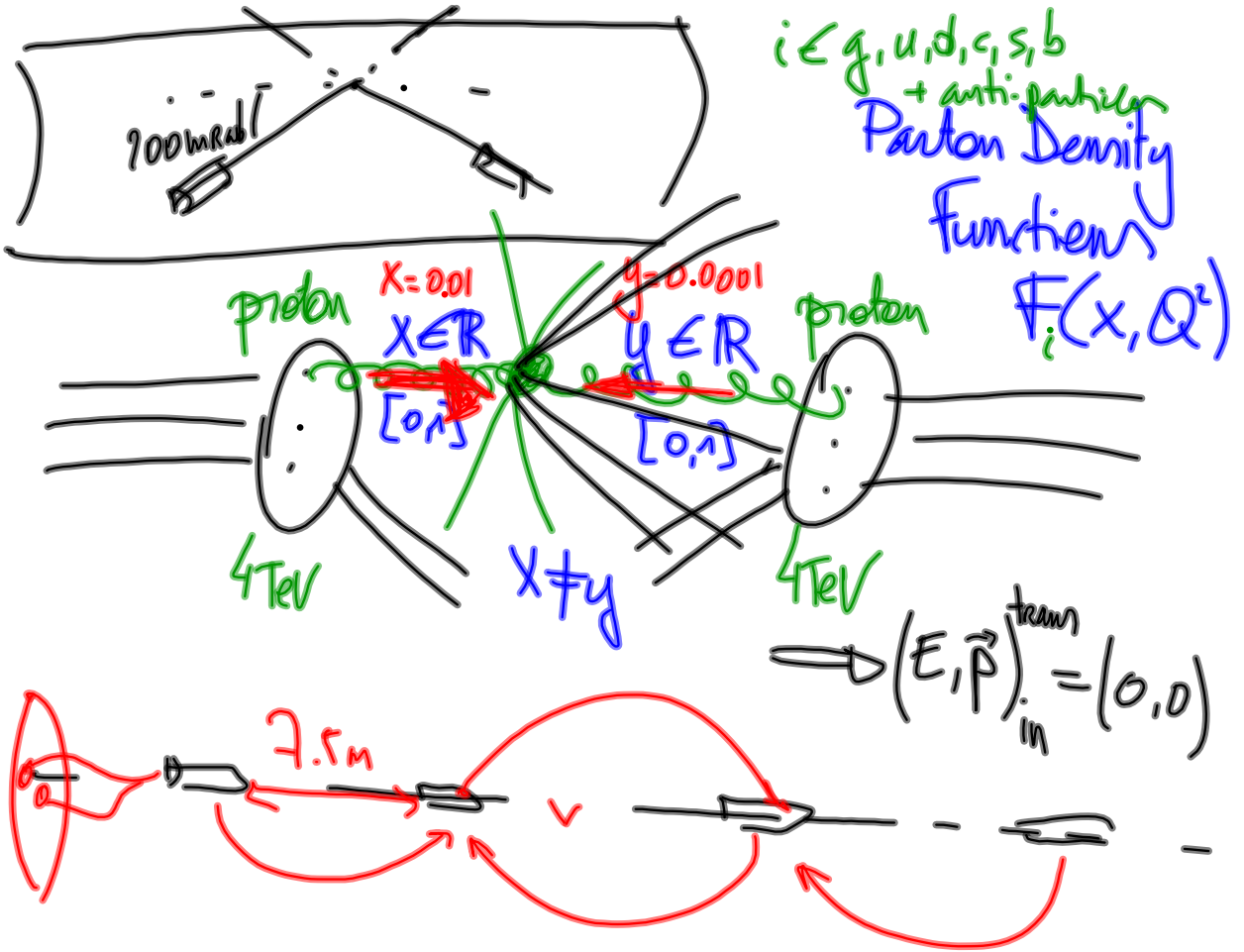
$$\vec{P} = (E, p_x, p_y, p_z)$$

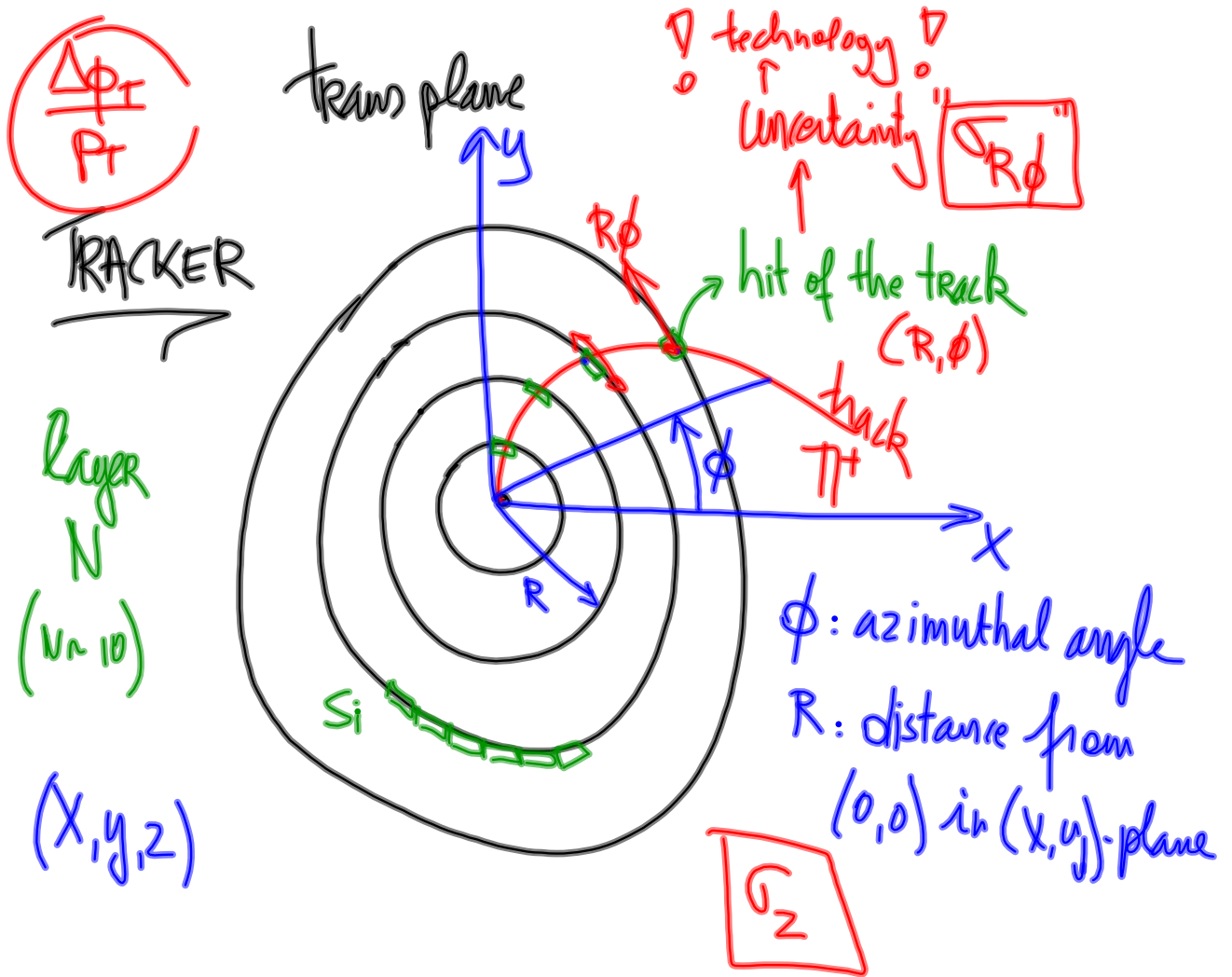
$$\text{mass } m^2 = E^2 - |\vec{P}|^2$$



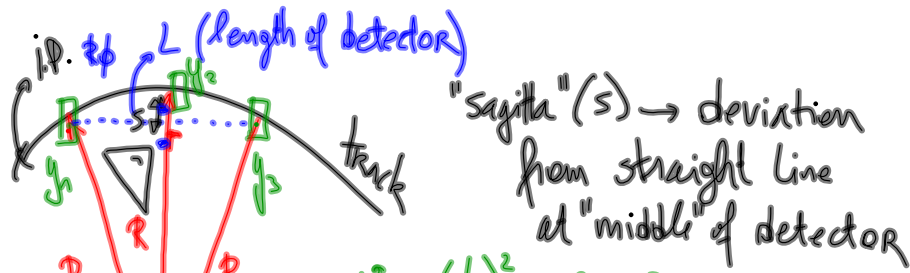












"sagitta" (s) → deviation from straight line at "middle" of detector

$$R^2 = \left(\frac{L}{2}\right)^2 + (R-s)^2$$

$$\rightarrow R^2 = \frac{L^2}{4} + R^2 - 2sR + s^2$$

$$\Rightarrow R = \frac{L^2}{8s} + \frac{s}{2}$$

high momentum  $R \gg s$

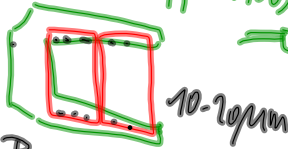
$$\Rightarrow R \approx \frac{L^2}{8s}$$

$$P_T = q \cdot \frac{L^2}{8s} \cdot B$$

$$B = 4T \quad L = 1m$$

$$P_T = 106eV$$

$$\Rightarrow s \approx 0.7 \text{ cm} \quad 700 \mu m$$



$$\left| \frac{dP_T}{ds} \right| = q \cdot \frac{L^2}{8s^2} B = \frac{P_T}{s}$$

$$\Rightarrow \frac{\Delta P_T}{P_T} = \frac{\Delta s}{s} = 8 \frac{\Delta s}{L^2} P_T \frac{1}{q B}$$

$$s = \frac{q L^2 B}{8 P_T}$$

$$s = y_2 - \frac{y_1 + y_3}{2}$$

$$\Delta y = \Delta y_1 = \Delta y_2 = \Delta y_3 \quad (\Delta s)^2 = \sum_{i=1}^3 \left( \frac{\partial s}{\partial y_i} \right)^2 (\Delta y_i)^2$$

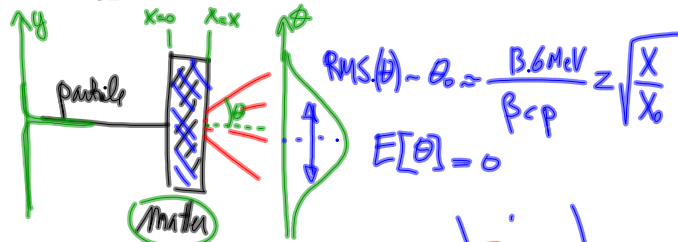
$$\Rightarrow (\Delta s)^2 = \frac{1}{4} (\Delta y)^2 + (\Delta y)^2 + \frac{1}{4} (\Delta y)^2$$

$$= \frac{3}{2} (\Delta y)^2 \Rightarrow \Delta s = \sqrt{\frac{3}{2}} \cdot \sigma_{rd}$$

$$\frac{\Delta P_T}{P_T} = 8 \sqrt{\frac{3}{2}} \frac{\sigma_{rd} \cdot P_T}{q L^2 B}$$

$$\frac{\Delta P_T}{P_T} = a_n \frac{\sigma_{rd} P_T}{0.3 B L^2} \quad \text{with } a_n = \sqrt{\frac{770}{N+4}}$$

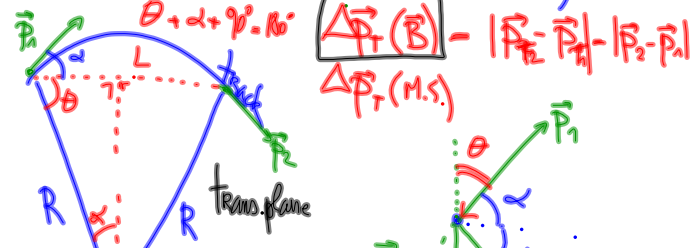
### Multiple Scattering (M.S.)



$$RMS(\theta) \sim \theta_0 \approx \frac{B \cdot 6 MeV}{\beta \cdot p} \approx \sqrt{\frac{X}{X_0}}$$

$$E[\theta] = 0$$

$X_0$ : radiation length.  $\nearrow$  SMALL  $\searrow$  LARGE  
 $E(X) = E(X_0) \cdot e^{-X/X_0}$   
 $X_0(\text{air}) \sim 300 \mu\text{m}$   
 $X_0(\text{Si}) \sim 9 \text{ cm}$



$$\Delta \vec{P}_T(B) = |\vec{P}_2 - \vec{P}_1| = |\vec{P}_2 - \vec{P}_1|$$

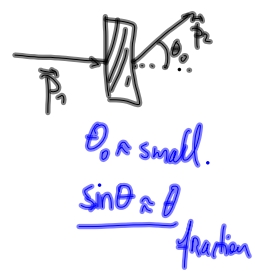
$$\Delta \vec{P}_T(M.S.)$$

$$R \cdot \cos \theta = \frac{L}{2}$$

$$\vec{P}_1 \cdot \cos \theta = \frac{\Delta \vec{P}_T}{2}$$

$$\rightarrow \vec{P}_T \cdot \frac{L}{2R} = \frac{\Delta \vec{P}_T}{2} = q \cdot B \cdot L \rightarrow \Delta \vec{P}_T(B) = q B L$$

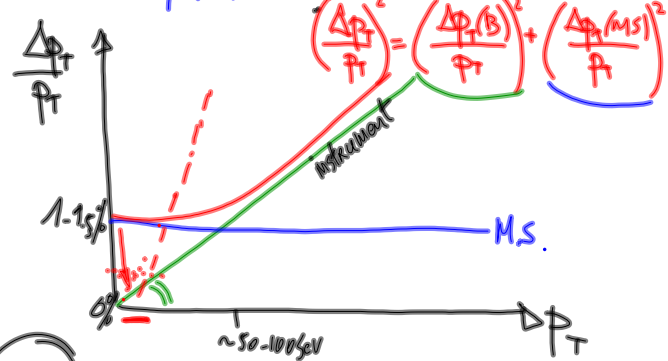
$$|\Delta \vec{P}_T(M.S.)| = |\vec{P}_2 - \vec{P}_1|$$



$$|\vec{P}_1| \cdot \sin \frac{\theta_0}{2} = \frac{\Delta \vec{P}_T}{2}$$

$$\Rightarrow P_T \cdot \frac{\theta_0}{2} = \frac{\Delta \vec{P}_T}{2}$$

$$\Rightarrow \Delta \vec{P}_T(M.S.) = P_T \cdot \theta_0$$

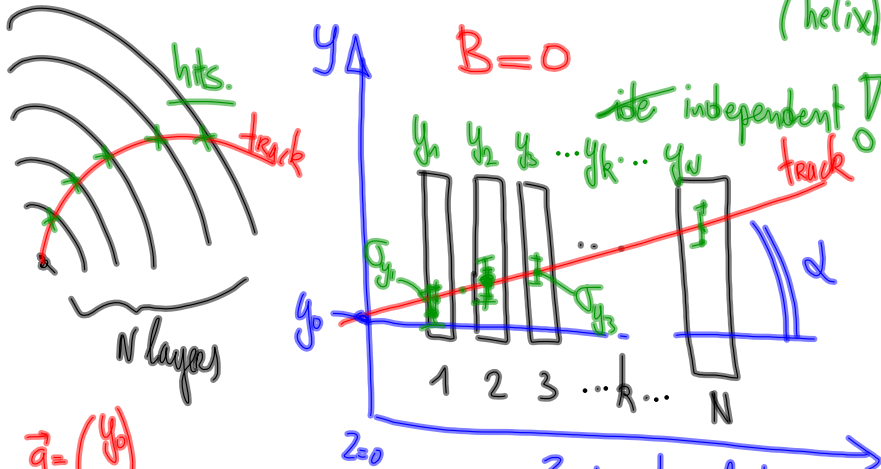


$$\left(\frac{\Delta P_T}{P_T}\right)^2 = \left(\frac{\Delta P_T(B)}{P_T}\right)^2 + \left(\frac{\Delta P_T(M.S.)}{P_T}\right)^2$$

$\frac{\Delta P_T}{P_T} \sim 1\%$  ( $P_T = 100 \text{ GeV}$ )  $\rightarrow N=10$   $L=1 \text{ m}$   $B=4 \text{ T}$   
 $\rightarrow \sigma_{RD} \sim 20 \mu\text{m}$

# TRACK FITTING

estimate track parameters.  $\vec{q} \rightarrow 5$  par. (helix)



$$\vec{q} = \begin{pmatrix} y_0 \\ \alpha \end{pmatrix}$$

Straight line:  $y = y_0 + (z_k - z_0) \cdot \alpha$   
 $z$ : position of det  
 $y$ : measurement position

distribution  $\propto \exp\left(-\frac{1}{2} \frac{(y_k - (y_0 + z_k \cdot \alpha))^2}{\sigma_{y_k}^2}\right)$  Gaussian

$\Rightarrow$  likelihood  $L(y_0, \alpha) = \prod_{k=1}^N \frac{1}{\sigma_{y_k}} \exp\left(-\frac{1}{2} \frac{(y_k - (y_0 + z_k \cdot \alpha))^2}{\sigma_{y_k}^2}\right)$

maximize  $L(y_0, \alpha) = \prod_{k=1}^N \frac{1}{\sigma_{y_k}} \cdot \exp\left[-\frac{1}{2} \left(\frac{y_k - (y_0 + z_k \cdot \alpha)}{\sigma_{y_k}}\right)^2\right]$

minimize  $-\ln L(y_0, \alpha) = c + \sum_{k=1}^N \frac{1}{2} \left(\frac{y_k - (y_0 + z_k \cdot \alpha)}{\sigma_{y_k}}\right)^2$

$\Rightarrow \chi^2(y_0, \alpha) = \sum_{k=1}^N \left(\frac{y_k - (y_0 + z_k \cdot \alpha)}{\sigma_{y_k}}\right)^2$

estimate  $\hat{x}$ : estimator of  $x$

$$\begin{cases} \frac{\partial \chi^2(y_0, \alpha)}{\partial y_0} \\ \frac{\partial \chi^2(y_0, \alpha)}{\partial \alpha} \end{cases} \Big|_{y_0 = \hat{y}_0, \alpha = \hat{\alpha}} = 0$$

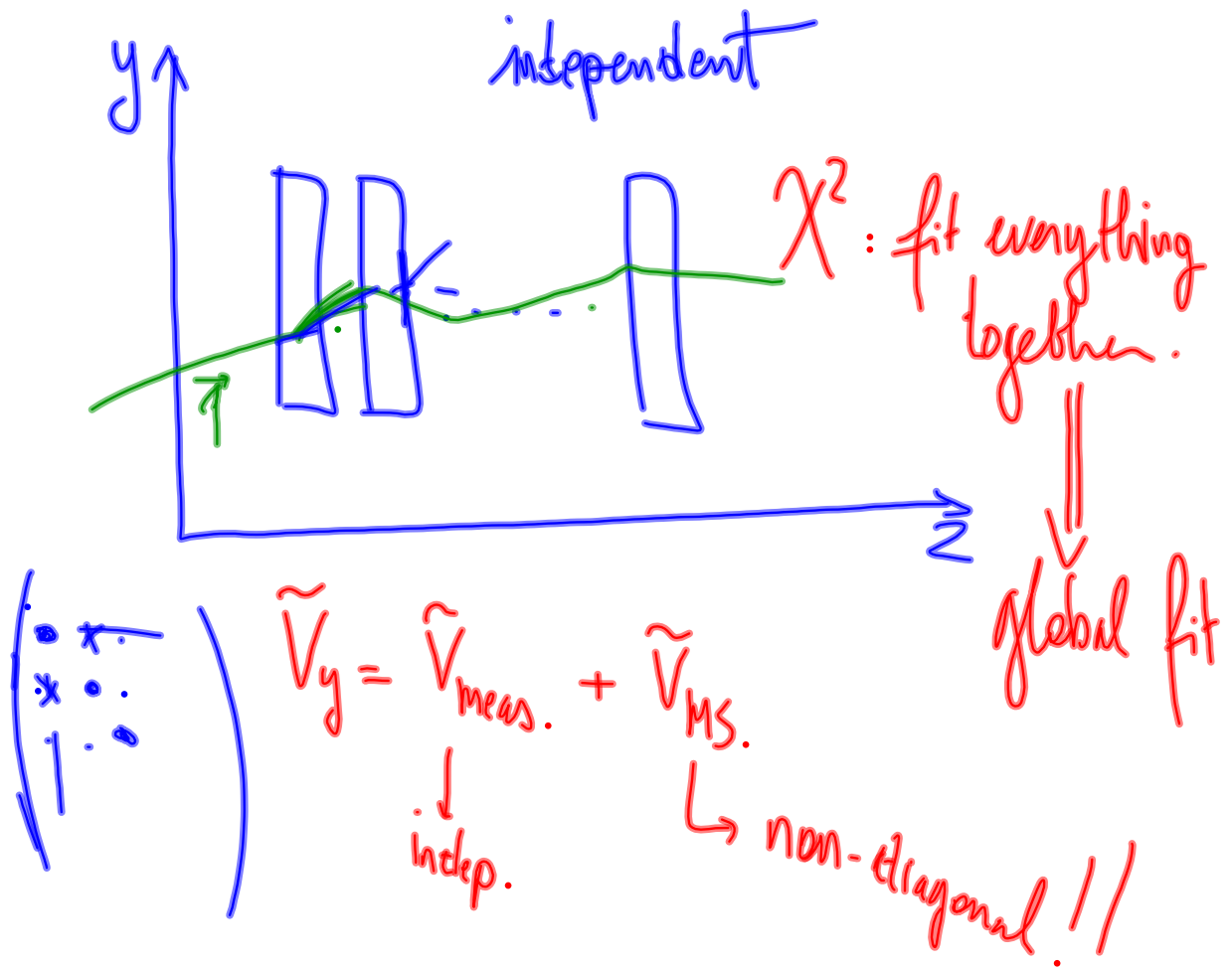
$\Rightarrow \begin{cases} \hat{y}_0 = \dots \\ \hat{\alpha} = \dots \end{cases}$

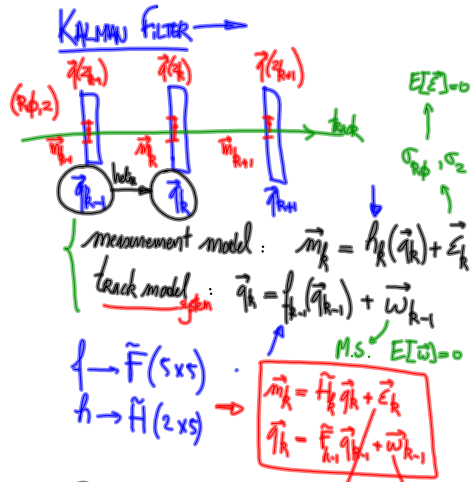
$\vec{q}$  (helix)  $\rightarrow$   $N$  measurements  $y_k$   
 $M$  parameters ( $M=5$ )  
 relation:  $y_k = f_k(\vec{q}) \quad k \in \{1 \dots N\}$   
 $\hookrightarrow$  helix (non-linear)  
 $\vec{q} = \vec{q}_A$  (numbers)  $\rightarrow$  Linearize it (TAYLOR)  
 $\Rightarrow y_k = f_k(\vec{q}_A) + \sum_{i=1}^5 \underbrace{\left( \frac{\partial f_k(\vec{q})}{\partial q_i} \right)_{\vec{q}=\vec{q}_A}}_{A_{ki}} \cdot (q_i - q_{A,i}) + \dots$

$\Rightarrow \chi^2 = \sum_{k=1}^N \left( \frac{y_k - f_k(\vec{q}_A) - \sum_{i=1}^5 A_{ki} \cdot (q_i - q_{A,i})}{\sigma_{y_k}} \right)^2$   
*note  $\sum_{i=1}^5 A_{ki}$*   
 $\chi^2(\vec{q})$

$\rightarrow \chi^2 = [\underbrace{\vec{y} - \vec{f}(\vec{q}_A)}_{\Delta \vec{y}} - \underbrace{\tilde{A}(\vec{q} - \vec{q}_A)}_{\text{matrix}}]^T \underbrace{(\tilde{V}_y^{-1})}_{\text{covariance matrix}} [\underbrace{\vec{y} - \vec{f}(\vec{q}_A)}_{\Delta \vec{y}} - \underbrace{\hat{A}(\vec{q} - \vec{q}_A)}_{\text{matrix}}]$   
 $\tilde{V}_y = \begin{pmatrix} 1/\sigma_{y_1}^2 & & & \\ & 1/\sigma_{y_2}^2 & & \\ & & \dots & \\ & & & 1/\sigma_{y_N}^2 \end{pmatrix}$   
*diagonal*

$\rightarrow \chi^2(\vec{q}) = [\Delta \vec{y} - \tilde{A}(\vec{q} - \vec{q}_A)]^T \tilde{V}_y^{-1} [\Delta \vec{y} - \hat{A}(\vec{q} - \vec{q}_A)]$   
 $\left. \frac{\partial \chi^2(\vec{q})}{\partial q_i} = 0 \right\}_{(5)} \Rightarrow \hat{\vec{q}} = \vec{q}_A + (\tilde{A}^T \tilde{V}_y^{-1} \tilde{A})^{-1} \tilde{A}^T \tilde{V}_y^{-1} \Delta \vec{y}$   
 $\hat{V}_q = (\tilde{A}^T \tilde{V}_y^{-1} \tilde{A})^{-1}$





- ① predict  $\hat{x}_{k-1} \rightarrow \hat{x}_k$
  - ② filter  $\hat{x}_k \rightarrow \hat{m}_k \rightarrow \hat{x}_k$
  - ③ smoothing
- $\hat{x}_{k,t} = \tilde{F}_{k,t} \hat{x}_{k-1,t} + \tilde{w}_{k,t}$   
 $\hat{m}_k = \tilde{F}_k \hat{x}_k + \tilde{e}_k$   
 $\tilde{C}_k = \text{cov}(\hat{x}_k - \hat{x}_{k,t})$  (error)  
 Residuals:  $\tilde{r}_k = \hat{m}_k - \tilde{H}_k \hat{x}_k$   
 $\text{cov}(\tilde{r}_k) = \tilde{R}_k$

① Prediction (k-1) → (k)

$\hat{x}_k = \tilde{F}_k \hat{x}_{k-1} + 0$  (extrapolate  $\hat{x}$ )  
 $\tilde{C}_k = \tilde{F}_k \tilde{C}_{k-1} \tilde{F}_k^T + \tilde{Q}_k$  (extrapol. of  $\tilde{C}$ )  
 $\tilde{R}_k = \tilde{H}_k \tilde{C}_k \tilde{H}_k^T + \tilde{R}_k$   
 $\tilde{V}_k = \tilde{V}_{k-1} + \tilde{F}_k \tilde{C}_{k-1} \tilde{F}_k^T$

② Filtering

$\hat{x}_k = \hat{x}_k^* + K_k (\hat{m}_k - \tilde{H}_k \hat{x}_k^*)$  (correction proposed)  
 $K_k = \tilde{C}_k \tilde{H}_k^T (\tilde{V}_k + \tilde{H}_k \tilde{C}_k \tilde{H}_k^T + \tilde{R}_k)^{-1}$  (Kalman Gain Matrix)  
 Unc. prediction  $\nabla$     Unc. on correction  $\nabla$

Min. Variance Bound.  $\hat{\theta} \rightarrow \{x_i\} \rightarrow$  smallest variance an estimator of  $\theta$  can have

$\tilde{C}_k = (I - K_k \tilde{H}_k) \tilde{C}_{k-1}$   
 $\tilde{R}_k = \hat{m}_k - \tilde{H}_k \hat{x}_k^* = \hat{m}_k - \tilde{H}_k (\hat{x}_{k-1}^* + K_k \tilde{r}_{k-1})$   
 $\tilde{R}_k = (I - \tilde{H}_k K_k) \tilde{R}_{k-1}$   
 $\tilde{R}_k = (I - \tilde{H}_k K_k) \tilde{R}_{k-1}$

③ Smoothing (skip)

