

Reconstruction methods in high energy physics experiments

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Tata Institute of Fundamental Research

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Outline of the Talk

- ① Detection principles in HEP-Ex
 - Importance of observation in science
 - Rutherford's gold foil experiment
 - Historical development of detectors
 - Design of a modern HEP-Ex detector
- ② What is reconstruction?
- ③ Reconstruction techniques
 - Pattern recognition & track finding.
 - Fitting and/or calibration.

If you see it, it's there



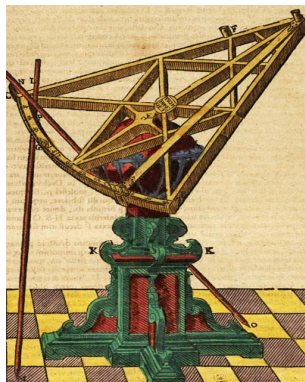
- Check what you believe
- Reproducibility

Theory reflects reality

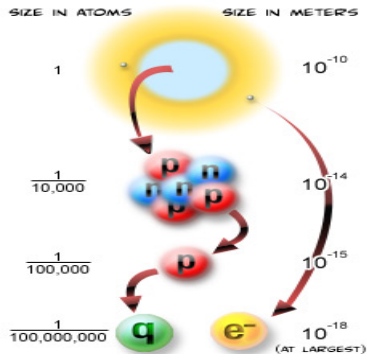
- World as we see it: celestial bodies to microscopic objects
- Build up a theory to understand physical world
- How do we know a theory is correct?
- Test the theory with experiment



An example



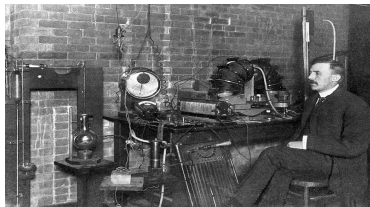
On a smaller scale



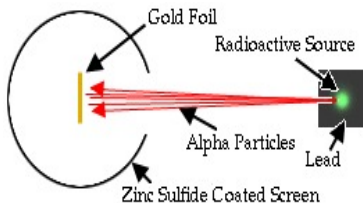
(a)

Figure : (a) Quest for fundamental particles

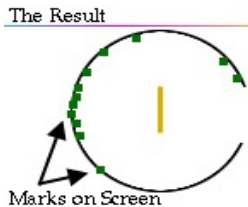
Break the indivisible



(a)

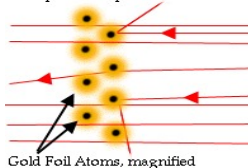


(b)



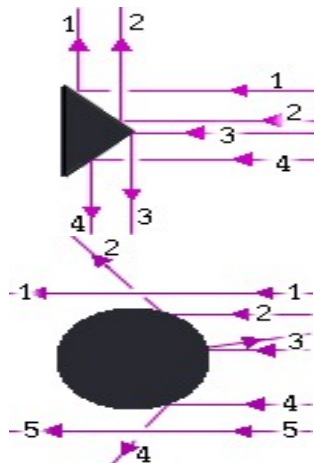
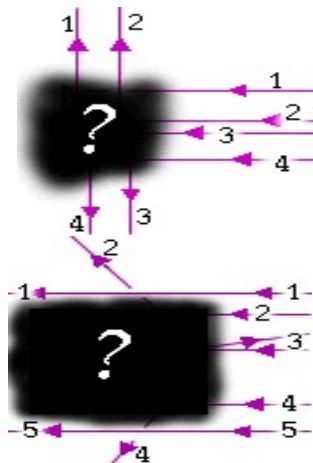
(c)

The Positive Nucleus Theory
Explains Alpha Deflection

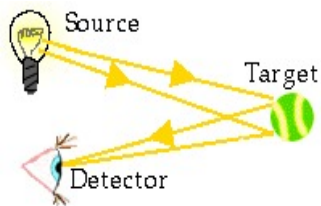


(d)

Analysis principle



Analysis principle

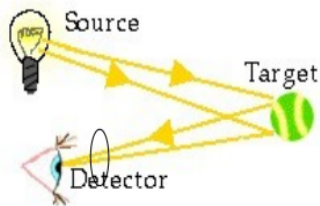


Beam

target

detector

Betterment of detector

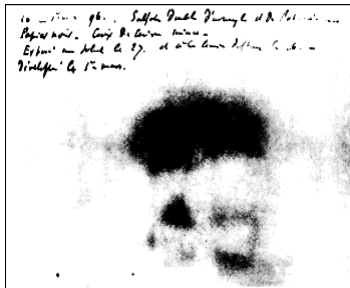


Beam

target

detector

α , β , γ , X-rays



(a)



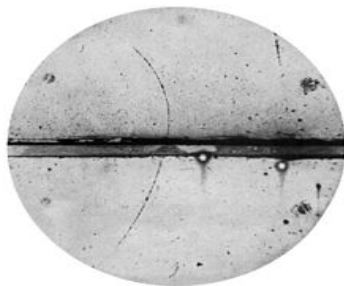
(b)

Figure : (a) Becquerel's photographic plate, fogged by exposure to radiation from a uranium salt; (b) X-ray picture taken by Wilhelm Röntgen of Albert von Kölliker's hand at a public lecture on 23 January 1896.

Cloud Chamber



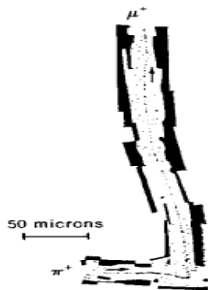
(a)



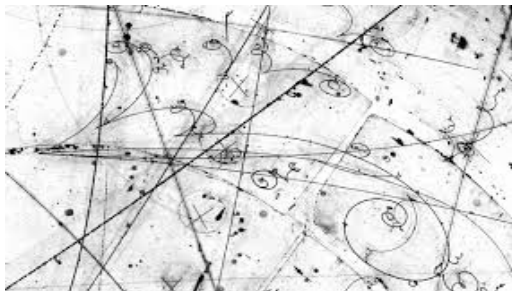
(b)

Figure : (a) Trails of particles in a cloud chamber; (b) Discovery of e^+ by Anderson

Nuclear emulsion and Bubble chamber



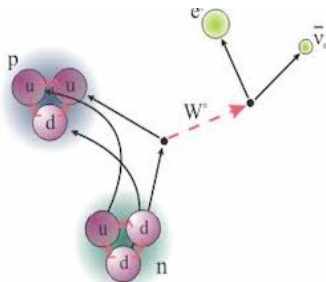
(a)



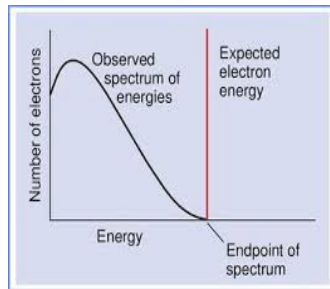
(b)

Figure : (a) Muon discovery via nuclear emulsion technique; (b) Bubble chamber

Importance of energy measurement



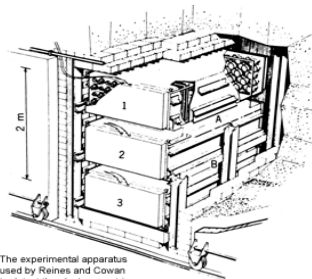
(a)



(b)

Figure : (a) β decay and (b) energy spectrum of emitted electrons

Neutrinos

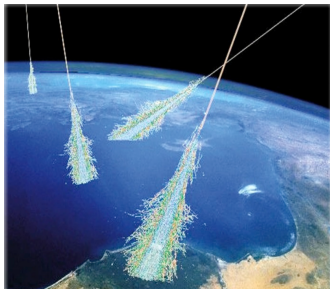


The experimental apparatus
used by Reines and Cowan
to detect the electron neutrino.

(a)

Figure : (a) Discovery of neutrinos

Importance of angle measurement



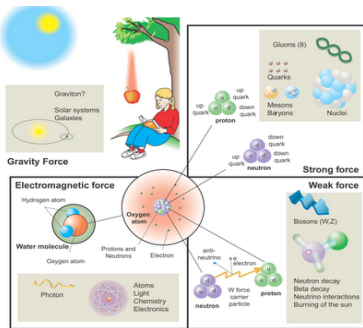
(a)



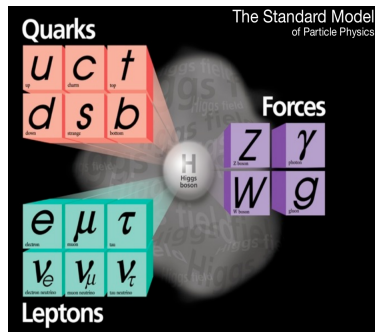
(b)

Figure : (a) Cosmic rays bombarding the upper atmosphere of the earth and (b) TIFR-based HAGAR telescope at Hanle, Ladakh

What did all these do for us?



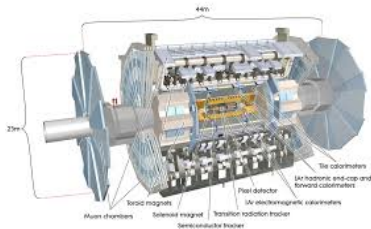
(a)



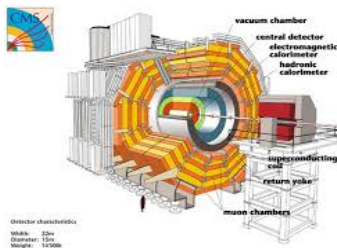
(b)

Figure : (a) four basic forces and (b) Fundamental building blocks of nature

Modern day detectors



(a)



(b)

Figure : Modern day complex detectors (a) ATLAS and (b) CMS

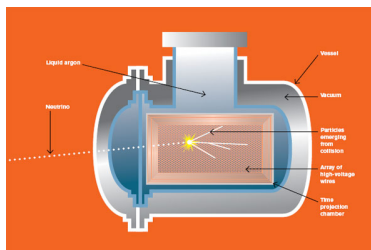
INO RPC lab



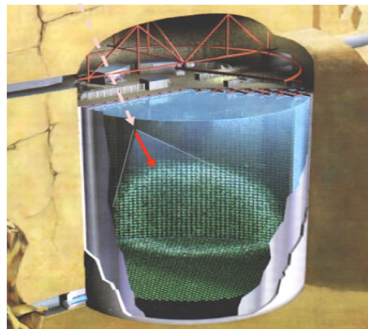
Design of a detector

- Need to know energy, direction, charge etc.
- Detector comprises:
 - Passive element (ex. Pb, Fe, LAr, H₂O etc.)
 - Active element (ex. Scintillating material + PMT, RPC etc.)

Examples of passive material



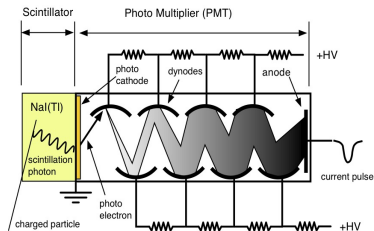
(a)



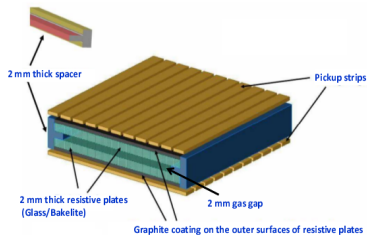
(b)

Figure : (a) a Liquid Argon TPC detector and (b) Neutrino event at Super Kamiokande

Examples of active material



(a)



(b)

Figure : (a) Scintillating material+PMT and (b) an RPC detector

Examples of active material

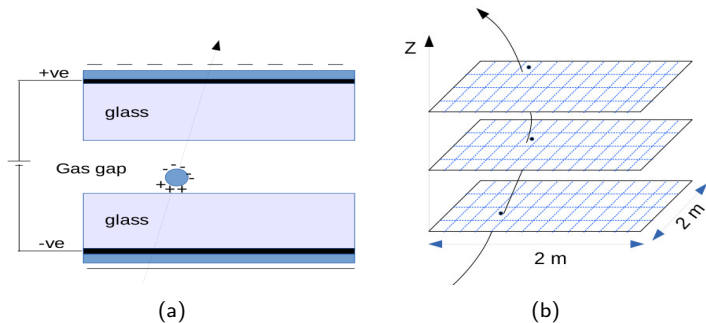


Figure : (a) and (b) (x,y) signal formation in RPC detector

How to build a detector

- In order to measure the properties of a particle
 - it must interact in the passive material and lose some energy.
 - transfer energy to the active material in some recognizable fashion.
- examples:
 - Charged particles: ionization, 
 - Photons: photo-electric/Compton effect, pair production

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From electronic signal to physics

- Whatever exotic physics we search for, the experimental data always come in the form of electronic signals.
- Measure the properties: charge, momenta, directions etc. from these signals.
- The physics depends on *how well* these measurements are.
- Issues: need to deal with background noise (process related and measurement related).

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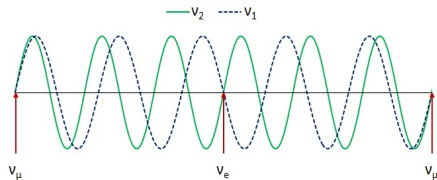
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Neutrino oscillation



$$\begin{pmatrix} \nu_e \\ \nu_\mu \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} \nu_1 \\ \nu_2 \end{pmatrix}$$

Oscillation probability - 2 flavors

Probability of transition from a state ν_i to a state ν_f :

$$P(\nu_i \rightarrow \nu_f) = \sin^2 2\theta \sin^2 \left(\frac{1.27 \Delta m^2 L}{E_f} \right)$$

where $\Delta m^2 = m_2^2 - m_1^2$

oscillation
parameters

m – mass (in eV)
 θ – mixing angle

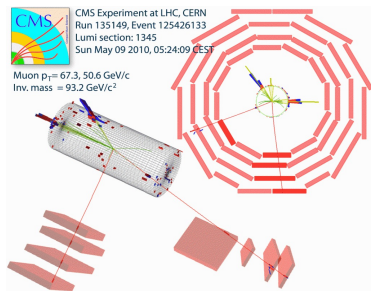
experimental conditions: E_f – neutrino energy (in GeV)

L – distance from a neutrino source to detector (km)

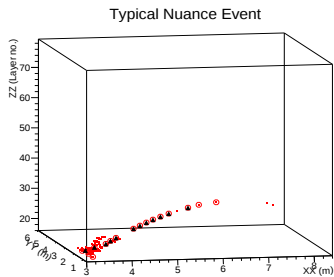
Oscillation length:

$$\sin \left(\frac{1.27 \Delta m^2 L}{E_f} \right) = \sin \pi \frac{L}{L_{\text{osc}}}$$

$$L_{\text{osc}} = \frac{2.5 E_f}{\Delta m^2}$$



(a)



(b)

Figure : (a) Typical event in CMS and (b) a neutrino event in ICAL

Typical challenges

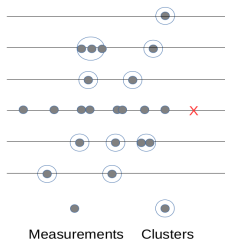
- Pattern recognition
- Track fitting

Typical challenges

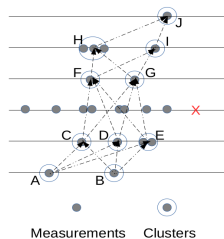
- Pattern recognition
- Track fitting

Standard method I

- Electronic signals at the active detector planes are used to form 'hits'
- Consider all hits in all active planes: form clusters.



(a)



(b)

Figure : (a) Formation of clusters from hits and (b) formation of triplets from clusters

Standard method II

- Form triplets from clusters
- Form chain of triplets (segments)

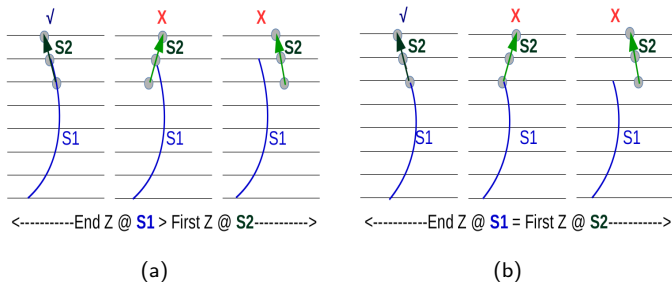
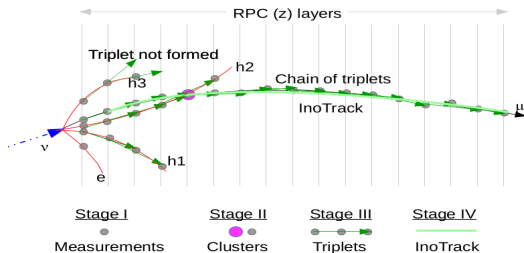


Figure : (a) and (b) Incrementing the length of track segments by appending new triplets

Standard method III



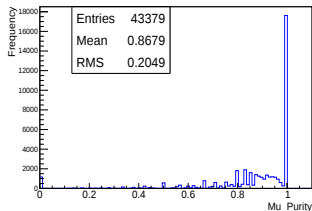
- Can work for any shape of tracking detector (barrel, planes etc.)
- Huge amount of combinations to try. Typically an event with 1000 hits/layer might take up to 100 sec

Performance in ICAL

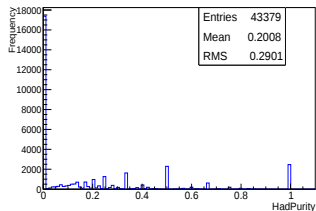
Table : NUANCE level Muon reconstruction performance

Program	Parameter Performance	
Finder	Directionality (up/down)	~ 96%
Finder	Correct identification of μ hits	~ 86%
,,	Correct identification of hadron hits	~ 80%
,,	Background rate of NC events	~ 1.7%

μ -hadron separation performance



(a) Degree of correct identification of muon hits

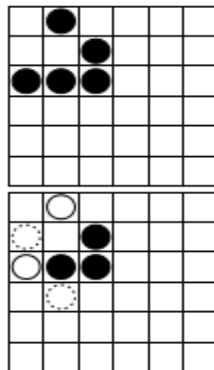


(b) Rate of inclusion of muon hits into hadron shower

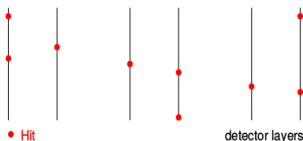
Figure : (a) Genuine μ fraction in hits tagged as 'muon hits' (b) genuine μ contamination in hits tagged as 'hadron hits'

Cellular automaton I

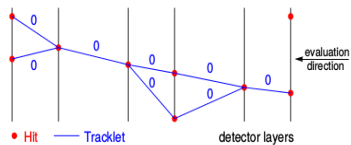
- Cell (*dead* or *alive*)
- Neighborhood
- Rule of evolution
- Discrete time step
- State of a cell depends on the state of its neighbors in the previous step
- Parallel track reconstruction in CMS, HERA-B using CA



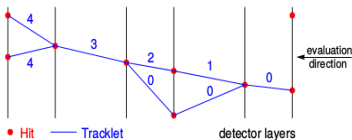
Cellular automaton II



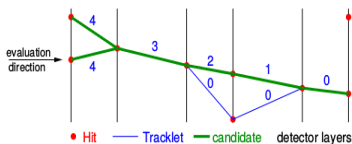
(a)



(b)



(c)



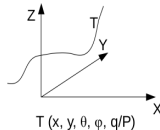
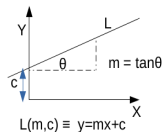
(d)

Typical challenges

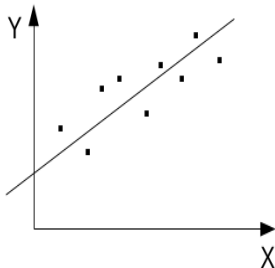
- Pattern recognition
- Track fitting

The problem

- Fit the tracks of selected track candidates
- Parameters of these tracks are q , P , positions and angles
- The state of the track evolves
- Use quadratic fit (Gluckstern), Kalman fit etc.



For straight line



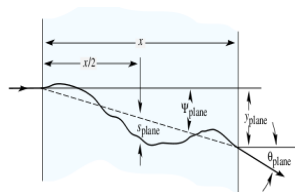
$$R^2 = \sum [y_i - f(a_1, a_2, \dots, a_n)]^2$$

$$\frac{\partial}{\partial a_i} (R^2) = 0$$

Use of Kalman filter

- Prediction

- $\vec{F} = q\vec{v} \times \vec{B} + \hat{u} \frac{dE}{ds}$
- $m \frac{d\vec{v}}{dt} = \vec{F}$
- Solve for $(x, y, t_x, t_y, q/p)$
- Account for the random multiple scattering



- Measurement

- Hits in a track
- Detector resolution, electronic noise and alignment errors

Use of Kalman filter

Figure : Kalman filtering and smoothing

- Start with a seed value of the state vector
 $\mathbf{x} = (x, y, t_x, t_y, q/p)$.
- Next plane is **predicted**.
- The measurement is considered.
- Hence, the state is **filtered**.
- **Smooth** the filtered states.

Least square minimization

- Filtered estimate is a weighted average of measurement and prediction:

$$x_f = \frac{\frac{1}{\sigma_p^2} x_p + \frac{1}{\sigma_m^2} x_m}{\frac{1}{\sigma_p^2} + \frac{1}{\sigma_m^2}}$$

$$= \frac{\sigma_m^2}{\sigma_p^2 + \sigma_m^2} x_p + \frac{\sigma_p^2}{\sigma_p^2 + \sigma_m^2} x_m$$

- $\sigma_p \rightarrow \infty$ implies $x_f \rightarrow x_m$
- $\sigma_m \rightarrow \infty$ implies $x_f \rightarrow x_p$

- Filtered error is given by:

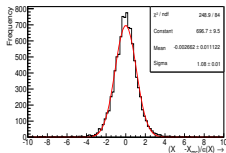
$$\frac{1}{\sigma_f^2} = \frac{1}{\sigma_m^2} + \frac{1}{\sigma_p^2}$$

- $\sigma_f^2 \leq \min(\sigma_m^2, \sigma_p^2)$. Filtering shrinks down the error.
- It can be written as:

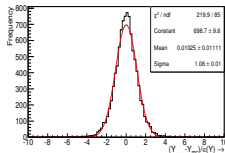
$$\sigma_f^2 = \left(1 - \frac{\sigma_p^2}{\sigma_m^2 + \sigma_p^2}\right) \sigma_p^2$$

$$= (1 - K) \sigma_p^2$$

Pull parameters



(a) Pull of X



(b) Pull of Y

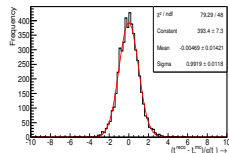
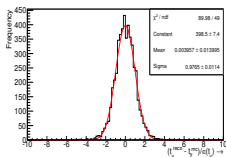
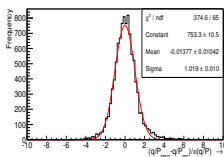
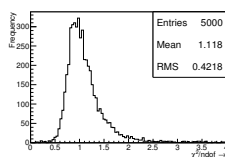
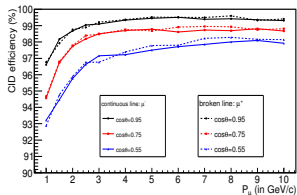
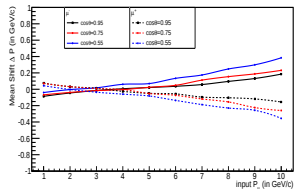
(c) Pull of t_x (d) Pull of t_y (e) Pull of q/P (f) Reduced χ^2

Figure : GoF plots for GEANT4 MC events $P_{\mu}^{\text{Gen}} = 6 \text{ GeV}/c$ at $\cos \theta_{\mu}^{\text{Gen}} = 0.95$

Performance



(a) Charge id. efficiency

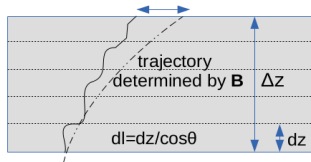


(b) Momentum mean shift

Current projects

- Generalising extrapolation formulae for (x, y, t_x, t_y) at low tracking momenta.
- $\frac{\partial(\frac{q}{p})_{l+dl}}{\partial(\frac{q}{p})_l} = \left(1 - \frac{2}{\beta P} \langle \frac{dE}{ds} \rangle dl\right) + O(B_i^{(2)})$
- $Q(l+dl) = Q(l) + Q'(l)dl = F(Ddl)Q(l)F(Ddl)^T + \delta Qdl$, Mankel¹

$$\begin{pmatrix} c(t_x t_x) \frac{\ell^2}{3} & c(t_x t_y) \frac{\ell^2}{3} & c(t_x t_x) \frac{\ell^2}{2} D & c(t_x t_y) \frac{\ell^2}{2} D \\ \dots & c(t_y t_y) \frac{\ell^2}{3} & c(t_x t_y) \frac{\ell^2}{2} D & c(t_y t_y) \frac{\ell^2}{2} D \\ \dots & \dots & c(t_x t_x) \ell & c(t_x t_y) \ell \\ \dots & \dots & \dots & c(t_y t_y) \ell \end{pmatrix}$$



$$c(t_x, t_x) = (1 + t_x^2)(1 + t_x^2 + t_y^2) \langle \theta_0^2 \rangle$$

$$\langle \theta_0^2 \rangle = \left(\frac{0.0136}{\beta p} \right)^2 \left(\frac{l}{l_{rad}} \right) \left[1 + 0.038 \ln \left(\frac{l}{l_{rad}} \right) \right]$$

²Ranger, A pattern recognition algorithm (Hera-B), R Mankel

Concluding remarks

- Methods are optimal but never fullproof
- Search for new methods are always on
- In ICAL, the standard approaches have been applied
- Other groups are attempting newer algorithms...
- Hope to have a better track reconstruction which will lead to better sensitivity of HEP experiments.

- I would like to thank the INO-ICAL collaboration for granting me the opportunity to spend time with the development of the reconstruction code for ICAL. The ample supports from Prof. N K Mondal and Dr. Wittek are appreciated.

Thank You

Bremsstrahlung and Cherenkov radiation

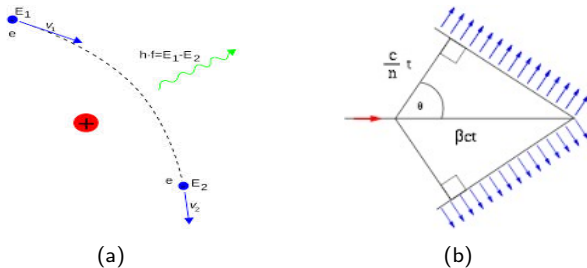


Figure : (a) Bremsstrahlung (b) Cherenkov radiation

Back to

BremssCheren