

# Advanced Tracking Algorithms

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## Motivation

- BigBite tracking workable for  $G_E^n$ , but slow ( $\mathcal{O}(10 \text{ Hz})$ ), memory-hungry
- Transversity vs.  $G_N^n$ : 18 planes vs. 15. Higher rates. Reconstruction may become impossibly slow.
- (Maybe) plenty of time for development if 6 month shutdown this year . . .

## Types of Tracking Algorithms

Reference: R. Mankel, Rept. Prog. Phys. **67**, 553 (2004).

### Global Methods

- Treat all detector hits equally
- Classic “Pattern Recognition” task
- Examples: Template Matching, Neural Networks

### Local Methods

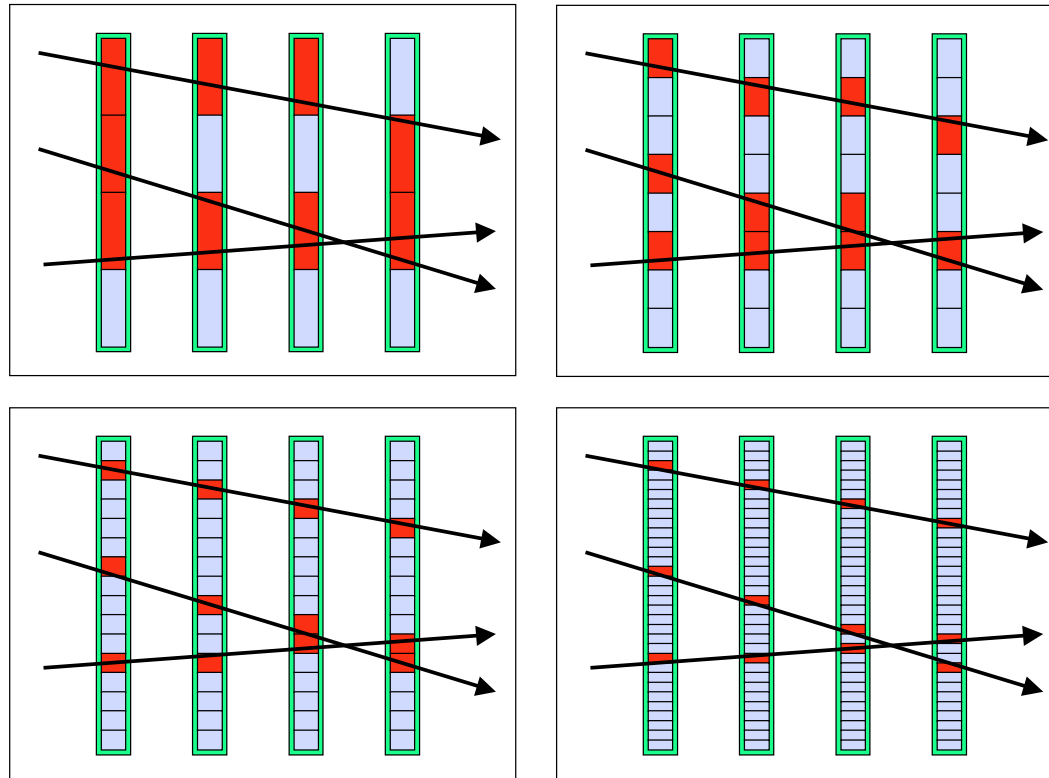
- Connect hits through detector step-by-step (“track following”)
- Choose starting points (“seeds”)
- Connect hits using a track model (“transport”)

## Global Methods: Tree Search (I)

### General

- Recursive template matching with increasing resolution
- Keeps template storage requirements reasonable even for high final resolution
- Fast and efficient
- Drawbacks: requires simple geometry, straight tracks

## Tree Search (II)



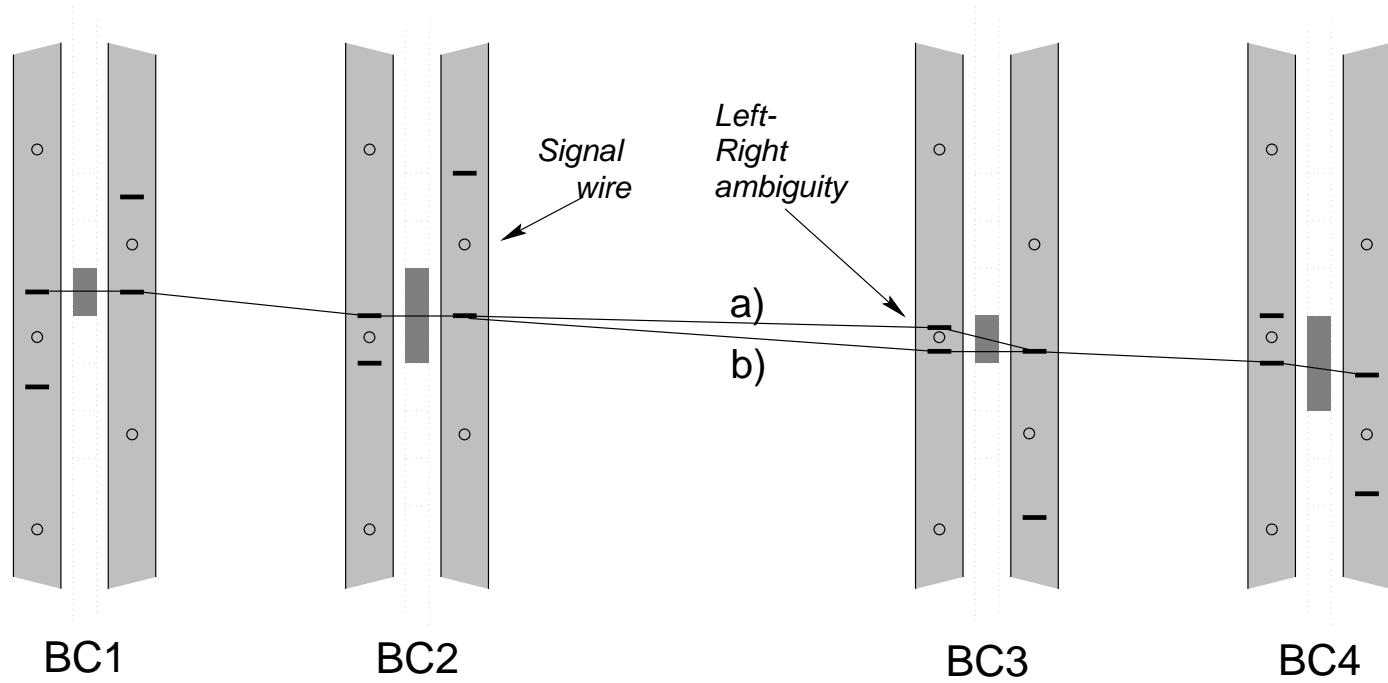
a.k.a. "track roads"

## Tree Search (III)

HERMES (ref.: W. Wander, PhD thesis)

- Back chambers: similar number of wires as BigBite; similar resolution  $\mathcal{O}(1 \text{ mm})$ ;  $u$ ,  $v$ , and  $x$  planes
- 126 million track patterns, reduced to 31.000 stored patterns by symmetry considerations
- Per road:  $\mathcal{O}(100)$  pattern comparisons, recursion depth of  $\mathcal{O}(10)$
- Least-squares fit within each road to find track.
- Operate on  $u$ ,  $v$ , and  $x$  projections separately, then combine in 3D

# Tree Search (IV)

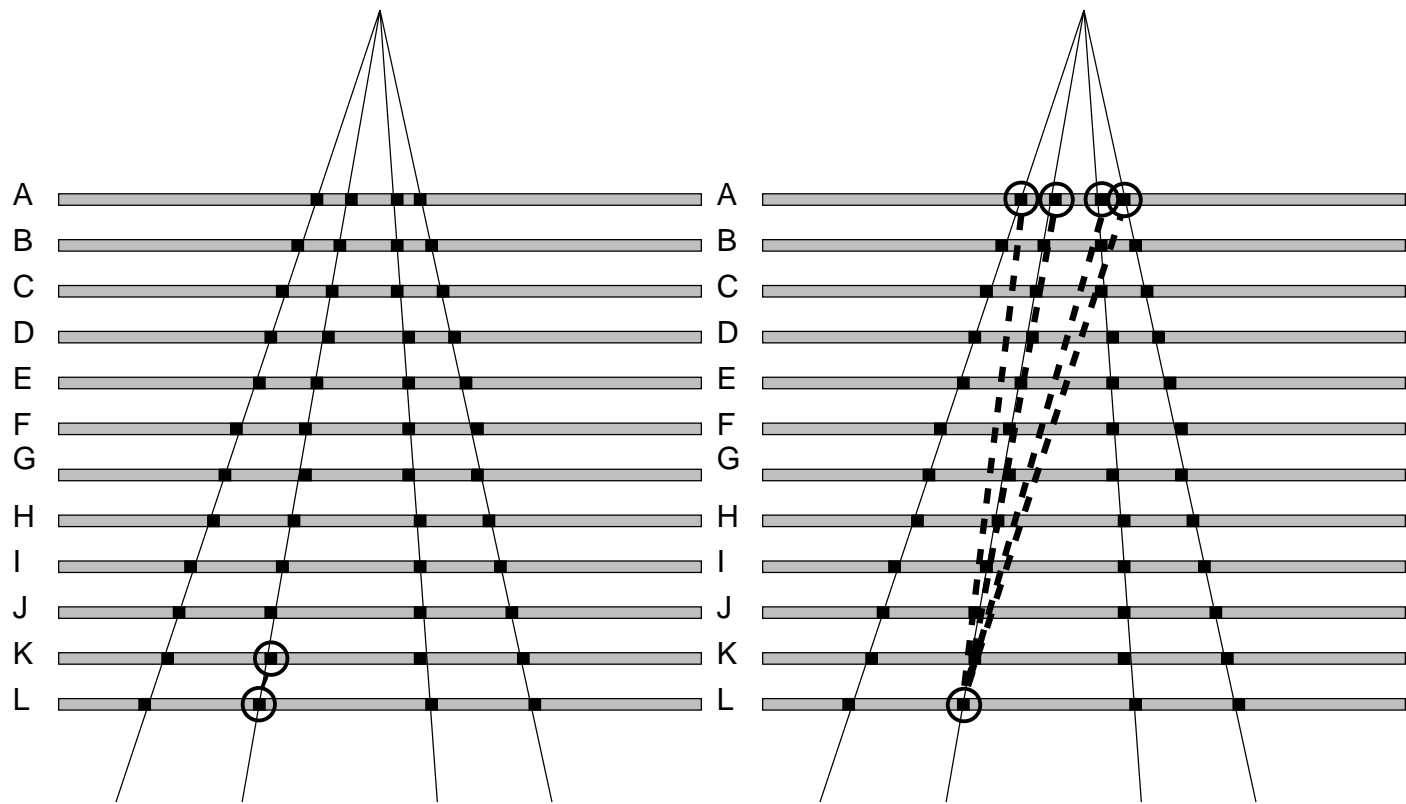


## Local Methods

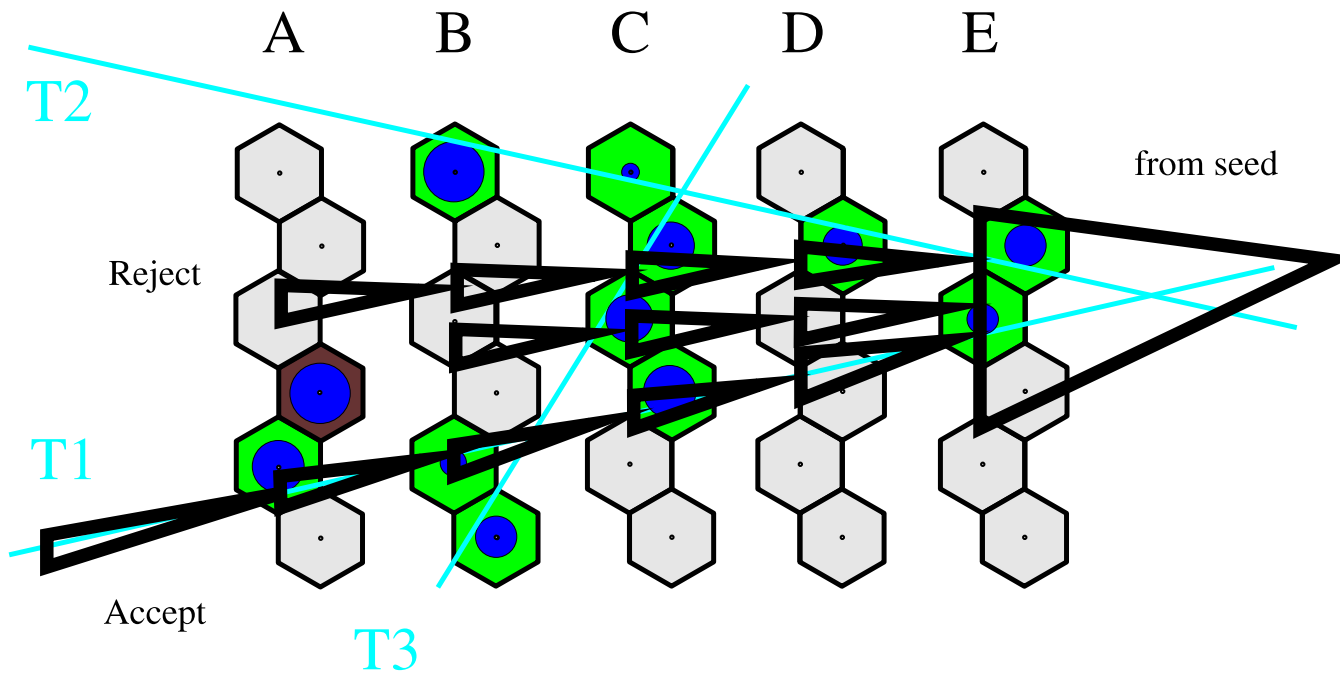
- Seeding required, may introduce bias
- Propagation: Kalman filter
- Naïve track following: Use best next match → one final track, but many ways to go wrong
- Combinatorial track following: Follow all possible next hits, then eliminate candidates via quality criterion  
→ efficient, but expensive, not scalable
- Arbitrated Track Following: Allow moderate concurrency of track candidates



# Seeding



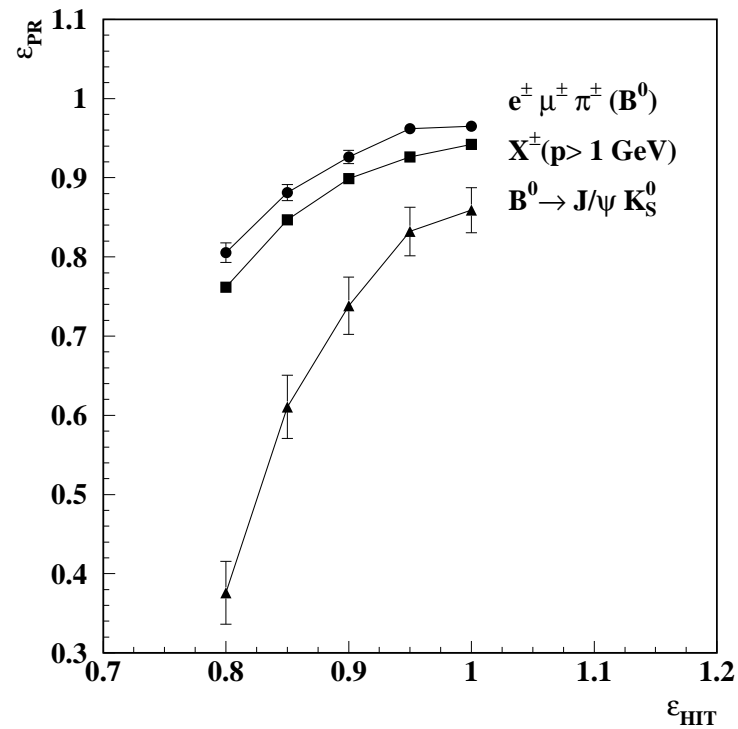
## Concurrent Track Evolution (I)



Used e.g. in HERA-B main tracker

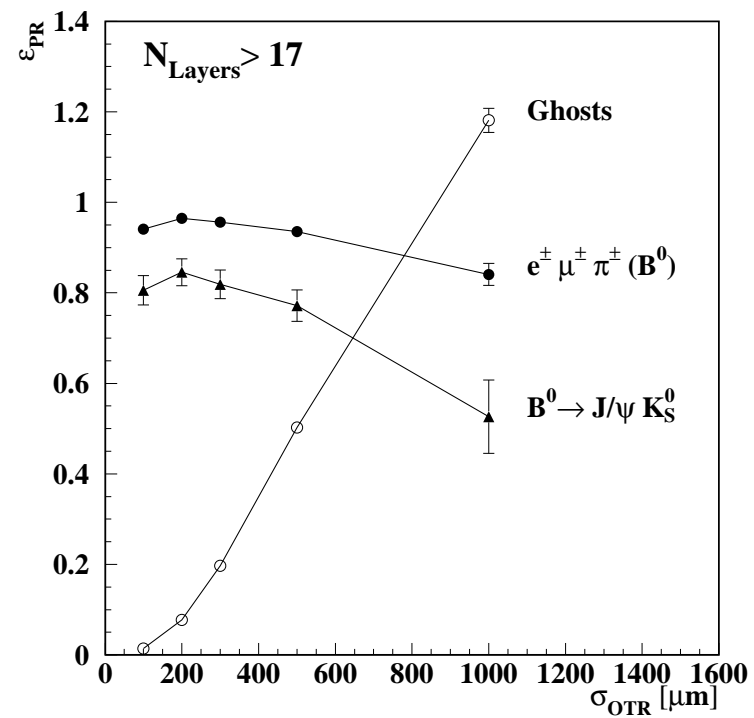
## Concurrent Track Evolution (II)

Tracking efficiency vs. hit efficiency. HERA-B simulation,  $N_{planes} = 24$ .



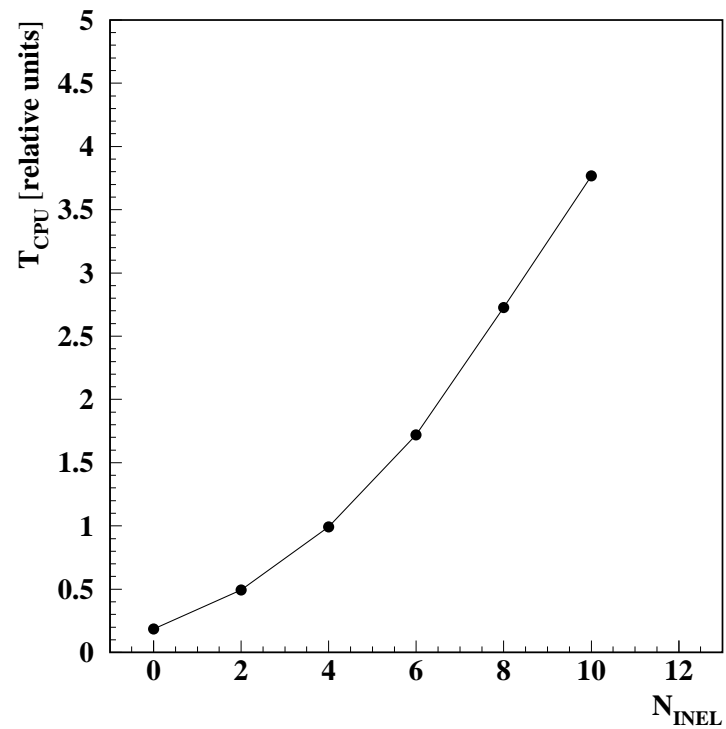
## Concurrent Track Evolution (III)

Tracking efficiency vs. resolution



## Concurrent Track Evolution (IV)

Mean computing time per event vs. number of tracks per event



## The Kalman Filter

- Progressive least-squares fit
- Two steps: prediction and filter
- Prediction: rejects noise efficiently
- Filter: Fast; much smaller matrix to be inverted than in global fit
- Elegantly accomodates “process noise”, e.g. multiple scattering, without introducing correlated errors
- See e.g. R. Frühwirth, NIM **A262**, 444 (1987)

## Summary

- Many advanced track reconstruction methods available in the literature
- Described two examples: a global tree search method and a local arbitrated track following method
- Both methods successfully used in trackers of similar or higher complexity than BigBite MWDCs & seem suitable for BigBite
- Work in progress: Experimental tree search code.