

Summary of Event Reconstruction with Cellular Automaton, Kalman Filter and KF Particle

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Cellular Automaton as Track Finder

0. Hits (CBM)

1000 Hits

0. Hits

1. Segments

2. Counters

3. Track Candidates

4. Tracks

Detector layers

Hits

Cellular Automaton:

1. Build short track segments.
2. Connect according to the track model, estimate a possible position on a track.
3. Tree structures appear, collect segments into track candidates.
4. Select the best track candidates.

Cellular Automaton:

- local w.r.t. data
- intrinsically parallel
- extremely simple
- very fast

Perfect for many-core CPU/GPU !

4. Tracks (CBM)

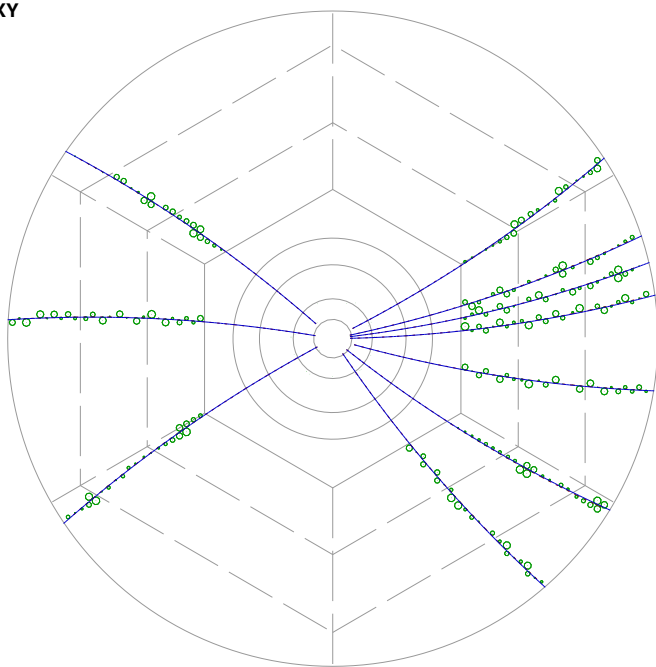
1000 Tracks

Useful for complicated event topologies with large combinatorics and for parallel hardware

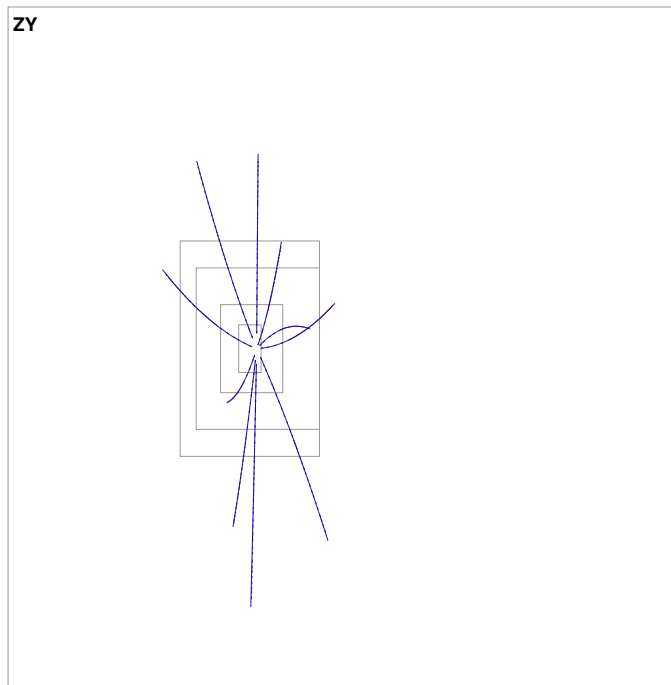
CA Track Finder in STT and MVD

10 primary tracks with $pt = 1 \text{ GeV}/c$

XY



ZY

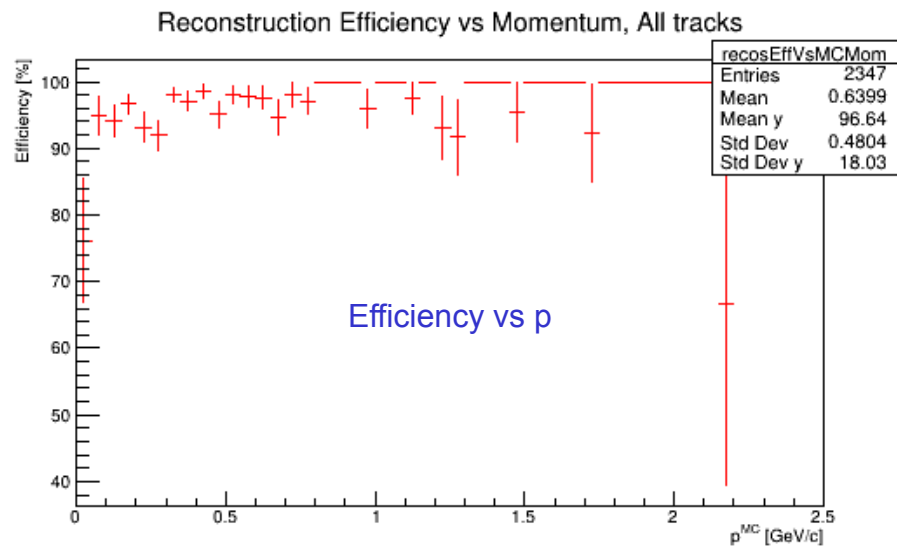
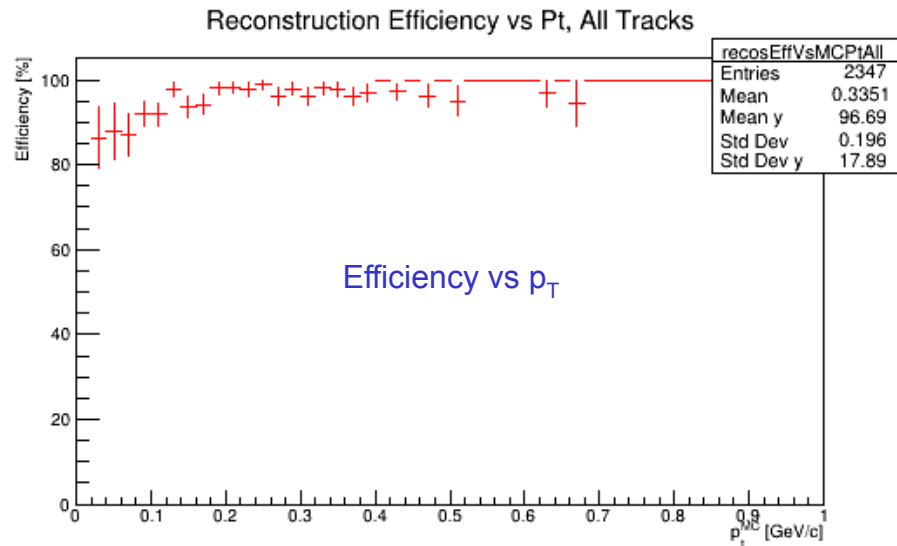


	STT	STT+MVD
Efficiency	97.2	99.3
Clone	1.8	9.2
Ghost	2.5	2.5
Tracks/event	10	10
Time, ms/event	5	7

	STT+MVD
Efficiency	100.0
Clone	4.8
Ghost	1.5
Tracks/event	10
Time, ms/event	1

Tracking Efficiency

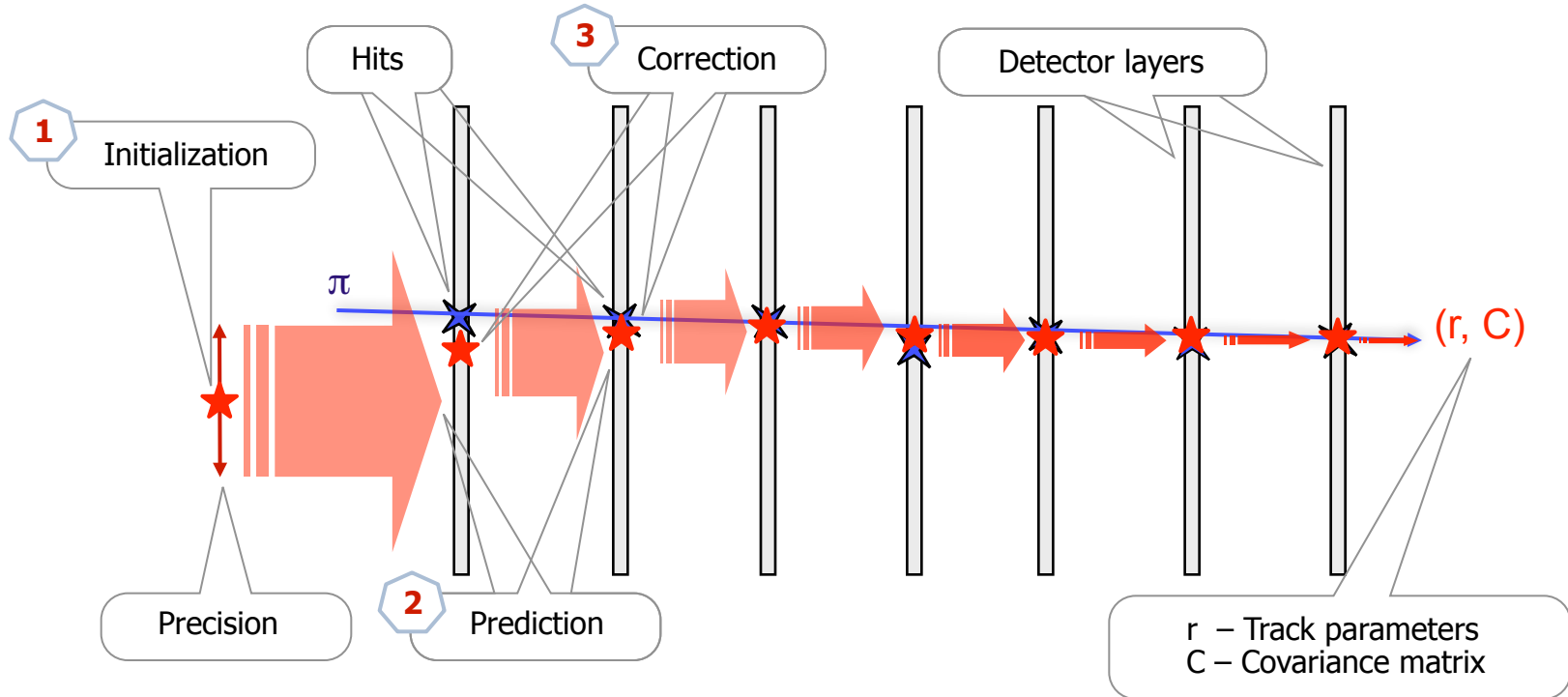
Efficiency, %	Eff. new
High-p primary	98.1
High-p secondary	93.3
High-p set	97.7
Low-p primary	97.7
Low-p secondary	93.0
Low-p set	96.2
All set	96.6



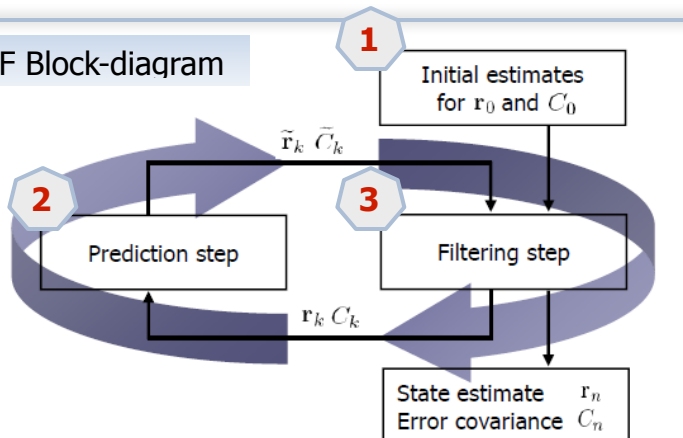
Irina Zivko, 07.06.2017

Kalman Filter based Track Fit

Track fit: Estimation of the track parameters at one or more hits along the track – Kalman Filter (KF)



KF Block-diagram



KF as a recursive least squares method

State vector

Position, direction and momentum

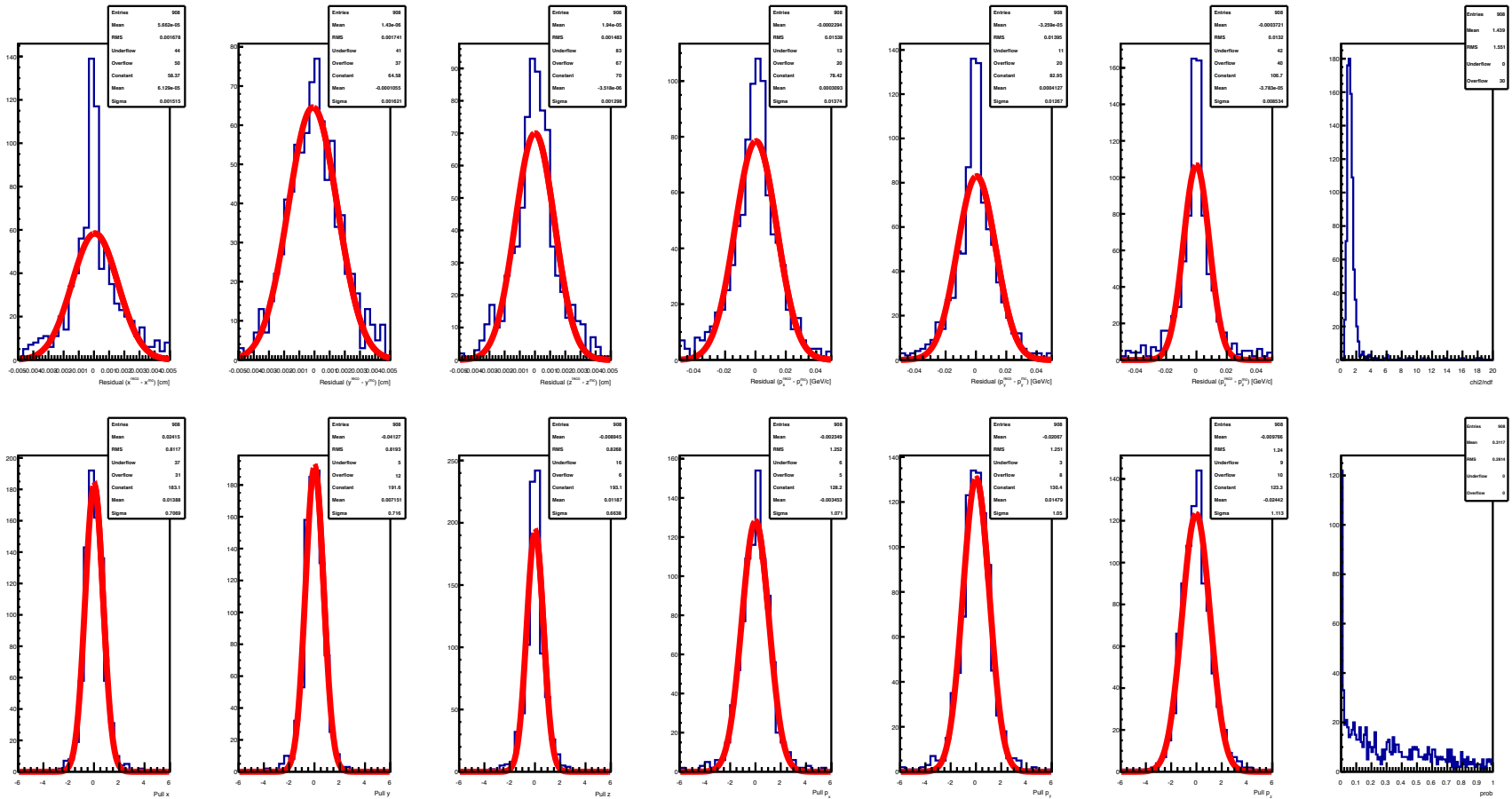
$$r = \{x, y, z, p_x, p_y, p_z\}$$

Kalman Filter:

1. Start with an arbitrary initialization.
2. Add one hit after another.
3. Improve the state vector.
4. Get the optimal parameters after the last hit.

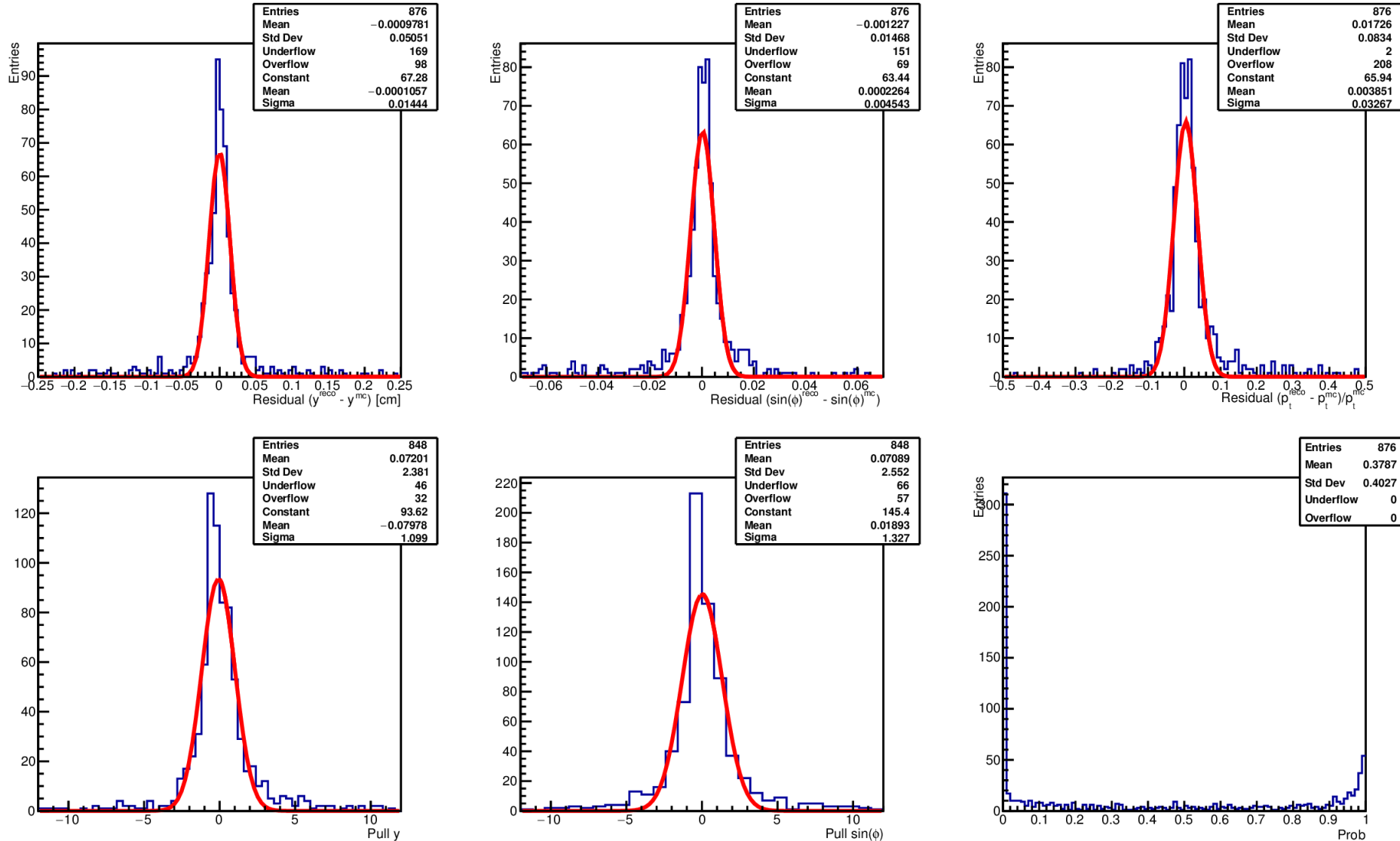
Nowadays the Kalman Filter is used in almost all HEP experiments

KF Fit Quality (STT+MVD)



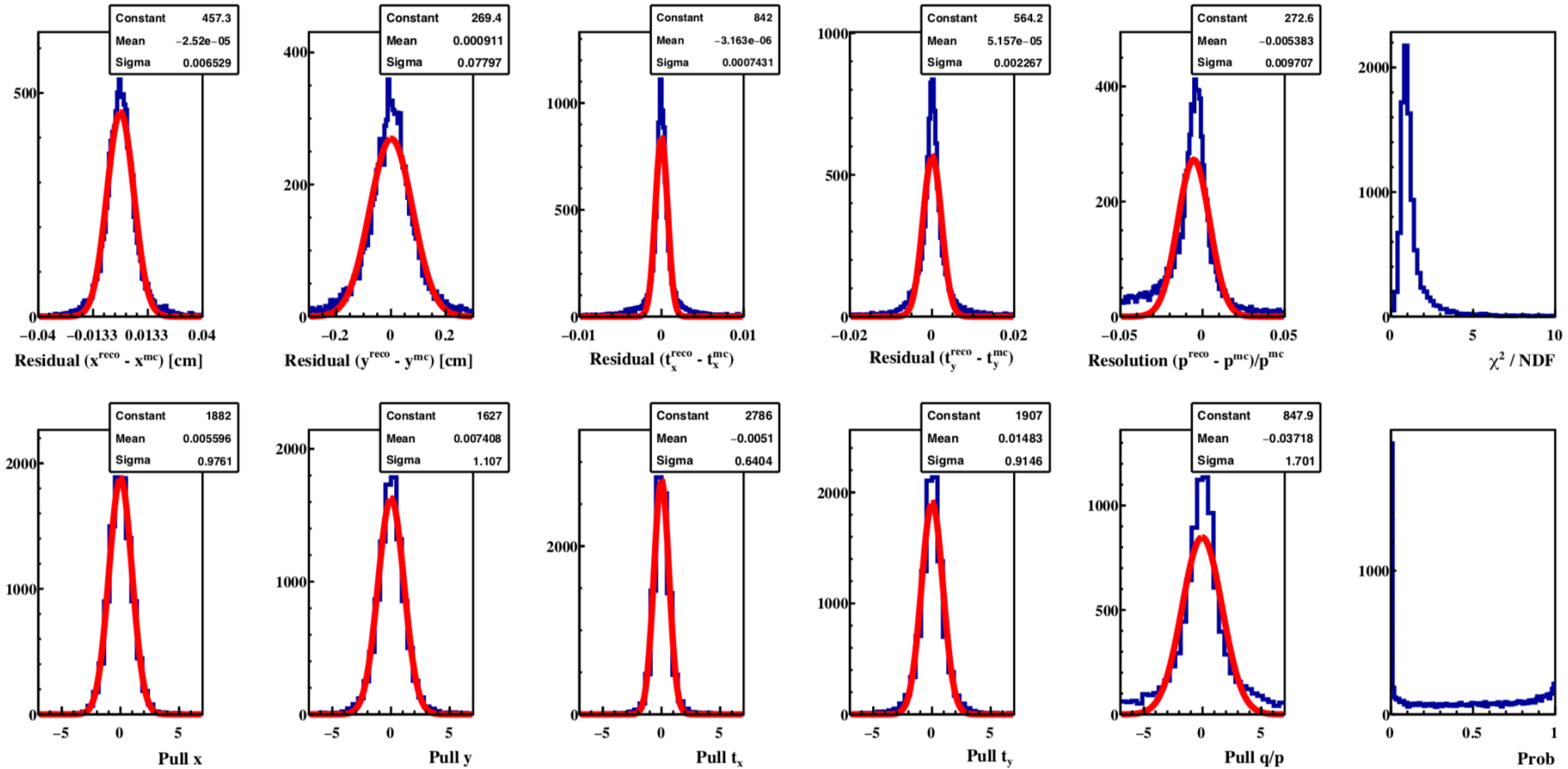
	x, μm	y, μm	z, μm	p_x , MeV/c	p_y , MeV/c	p_z , MeV/c
Residual	15	16	13	14	13	9
Pull	0,7	0,7	0,7	1,1	1,1	1,1

KF Residuals and Pulls (STT+MVD)

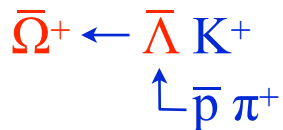
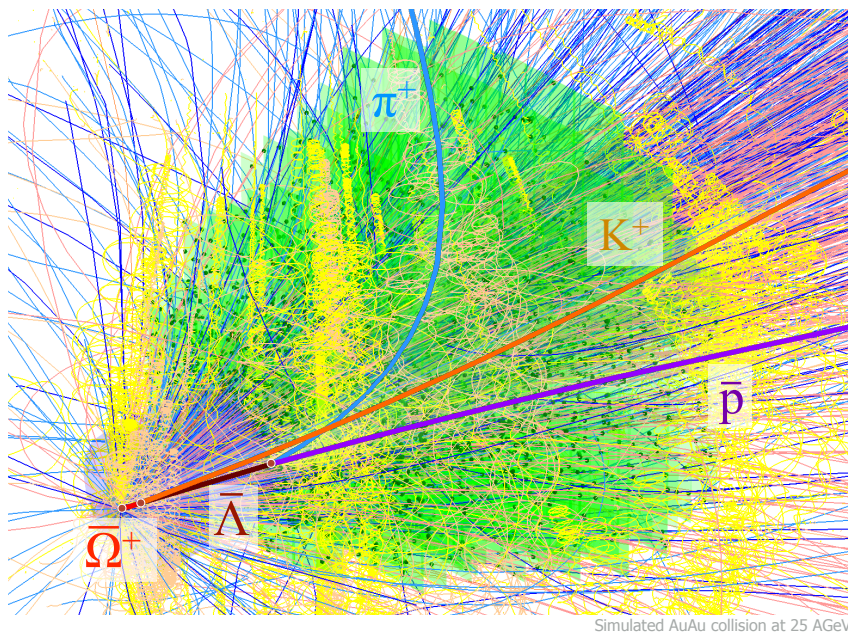


Irina Zivko, 07.06.2017

KF Residuals and Pulls (FTS)



KF Particle: Reconstruction of short-lived Particles



```

KFParticle Lambda(P, Pi);           // construct anti Lambda
Lambda.SetMassConstraint(1.1157);    // improve momentum and mass
KFParticle Omega(K, Lambda);        // construct anti Omega
PV -= (P; Pi; K);                   // clean the primary vertex
PV += Omega;                         // add Omega to the primary vertex
Omega.SetProductionVertex(PV);       // Omega is fully fitted
(K; Lambda).SetProductionVertex(Omega); // K, Lambda are fully fitted
(P; Pi).SetProductionVertex(Lambda); // p, pi are fully fitted
    
```

$$\mathbf{r} = \{ x, y, z, p_x, p_y, p_z, E \}$$

State vector

$$\mathbf{C} = \langle \mathbf{r} \mathbf{r}^T \rangle =$$

Covariance matrix

$$\begin{bmatrix} \sigma_x^2 & C_{xy} & C_{xz} & C_{xp_x} & C_{xp_y} & C_{xp_z} & C_{xE} \\ C_{xy} & \sigma_y^2 & C_{yz} & C_{yp_x} & C_{yp_y} & C_{yp_z} & C_{yE} \\ C_{xz} & C_{yz} & \sigma_z^2 & C_{zp_x} & C_{zp_y} & C_{zp_z} & C_{zE} \\ C_{xp_x} & C_{yp_x} & C_{zp_x} & \sigma_{p_x}^2 & C_{p_x p_y} & C_{p_x p_z} & C_{p_x E} \\ C_{xp_y} & C_{yp_y} & C_{zp_y} & C_{p_x p_y} & \sigma_{p_y}^2 & C_{p_y p_z} & C_{p_y E} \\ C_{xp_z} & C_{yp_z} & C_{zp_z} & C_{p_x p_z} & C_{p_y p_z} & \sigma_{p_z}^2 & C_{p_z E} \\ C_{xE} & C_{yE} & C_{zE} & C_{p_x E} & C_{p_y E} & C_{p_z E} & \sigma_E^2 \end{bmatrix}$$

Features:

- KF Particle class describes particles by the **state vector** and the **covariance matrix**.
- Covariance matrix contains essential information about tracking and **detector** performance.
- The method for **mathematically correct** usage of covariance matrices is provided by the KF Particle package based on the **Kalman filter** (KF).
- Heavy mathematics of KF requires **fast** and **vectorised** algorithms.
- **Mother** and **daughter** particles are treated in the same way.
- The **natural** and **simple interface** allows two reconstruct easily complicated decay chains.
- The package is geometrically independent and can be adapted to **different experiments** (CBM, ALICE, STAR).

KF Particle provides a simple and very efficient approach to physics analysis

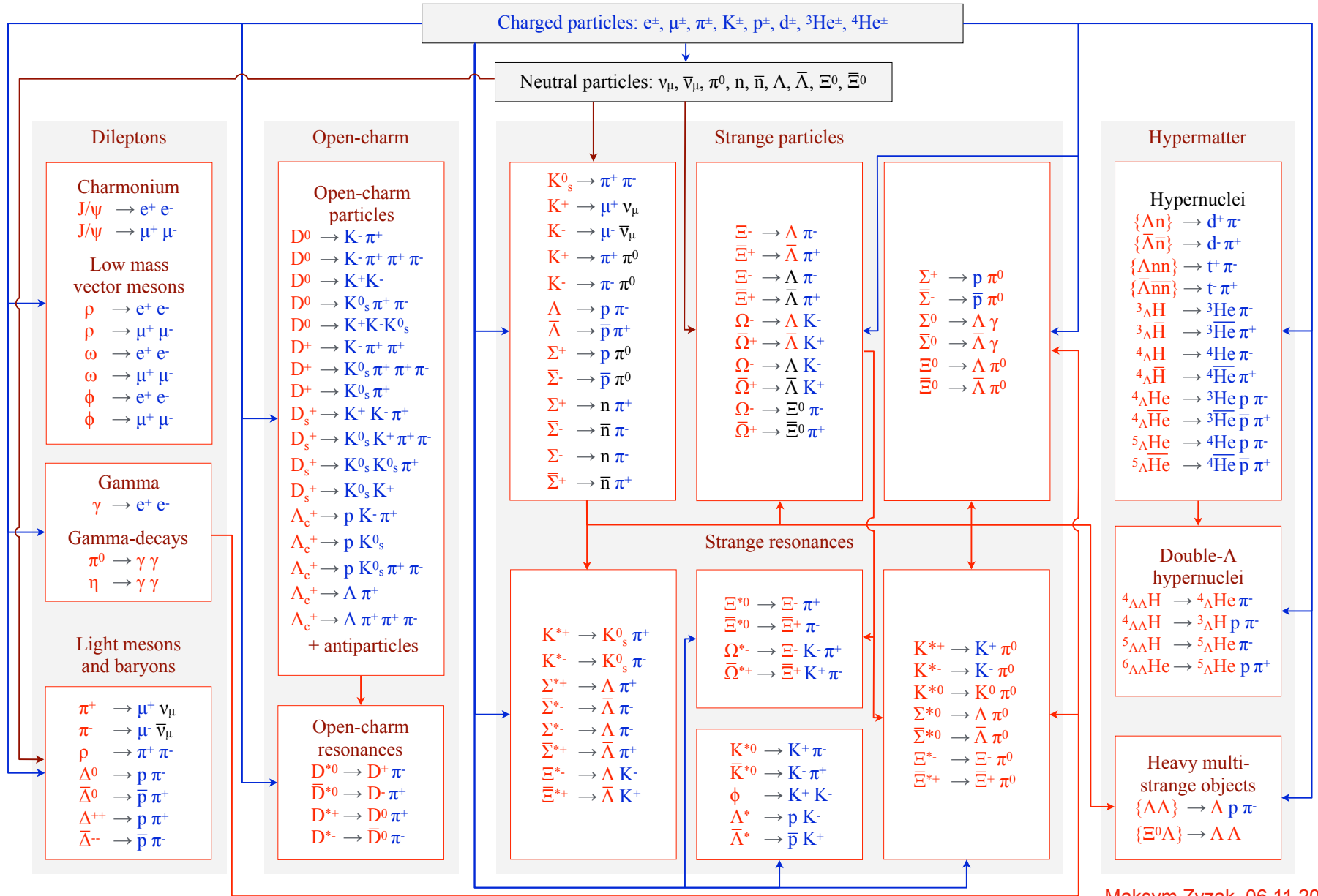
KF Particle: Functionality

Functions	CBM	PANDA	ALICE	STAR
Construction of mother particles	+	+	+	+
Addition and subtraction of the daughter particle to (from) the mother particle	+	+	+	+
$+=$ and $-=$ operators	+	+	+	+
Accessors to the physical parameters (mass, momentum, decay length, lifetime, rapidity, etc)	+	+	+	+
Transport: to an arbitrary point, to the decay and production points, to another particle, to a vertex, on the certain distance	+	+	+	+
Calculation of a distance: to a point, to a particle, to a vertex	+	+	+	+
Calculation of a deviation: from a point, from a particle, from a vertex	+	+	+	+
Calculation of the angle between particles	+	+	+	+
Constraints: on mass, on a production point, on a decay length	+	+	+	+
KF Particle Finder	+	+	+	+

Exactly the same package in all four experiments: CBM, PANDA, ALICE and STAR

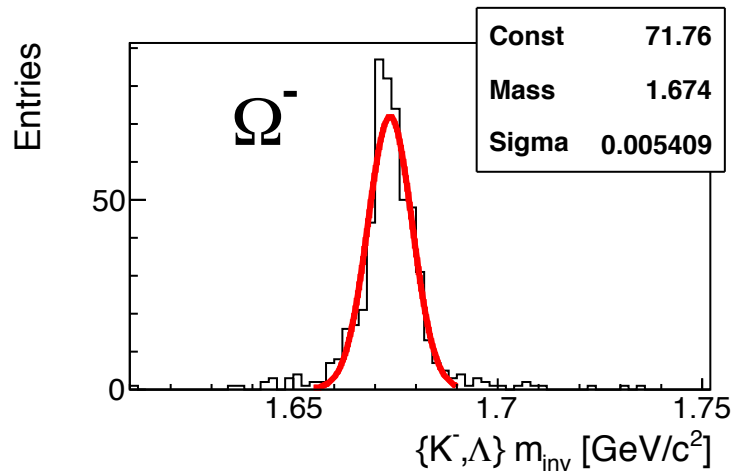
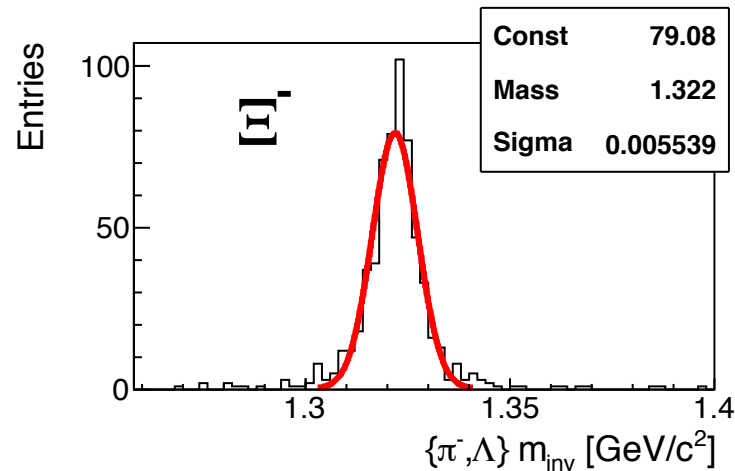
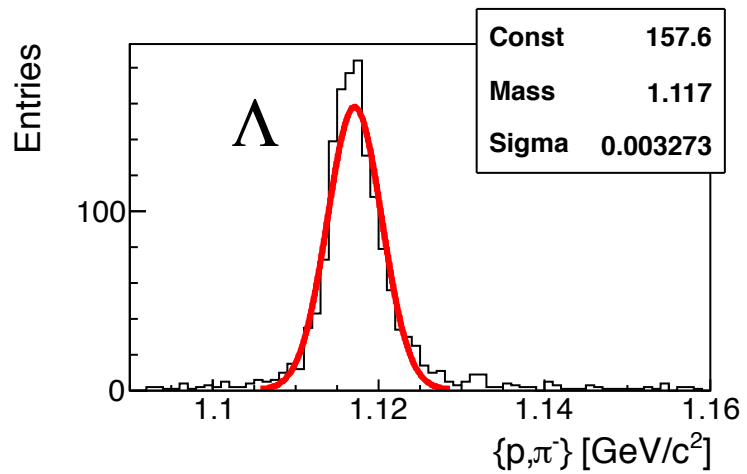
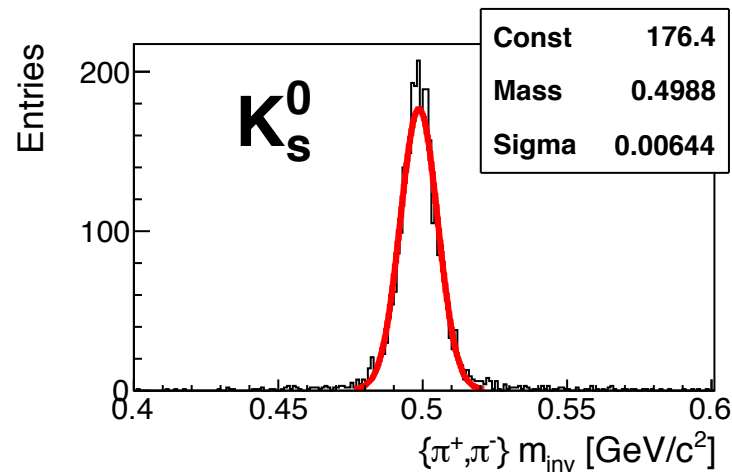
Maksym Zyzak, 09.06.2015

KF Particle Finder for Physics Analysis and Selection



Maksym Zyzak, 06.11.2018

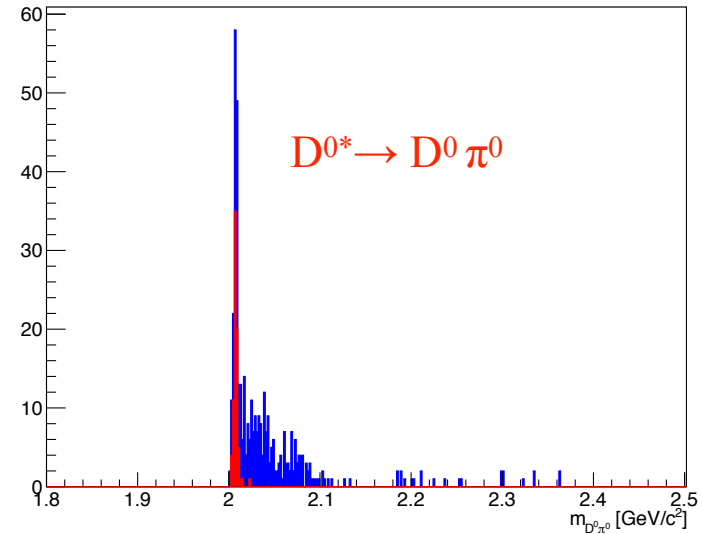
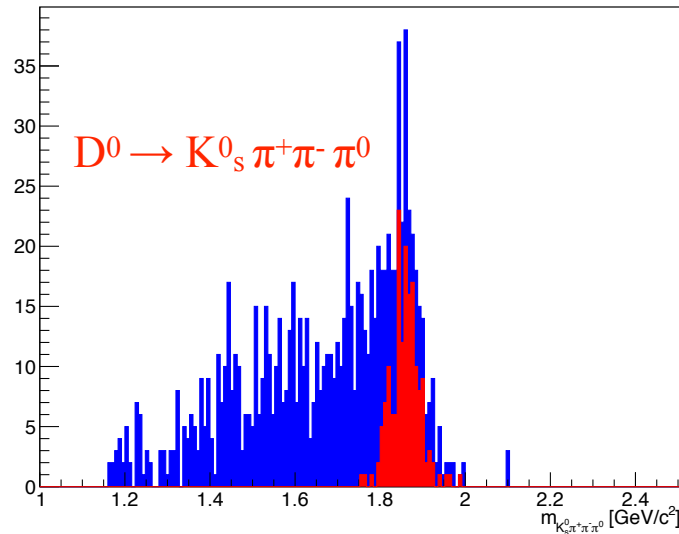
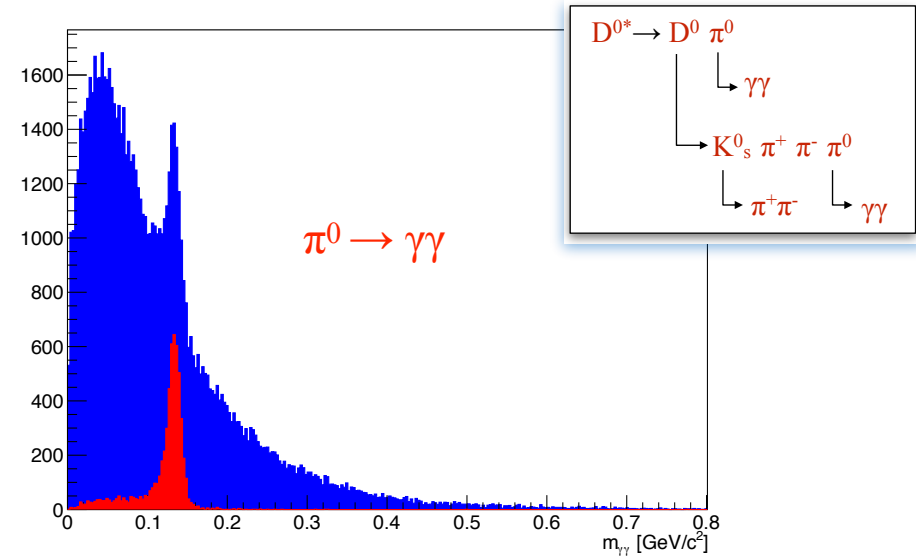
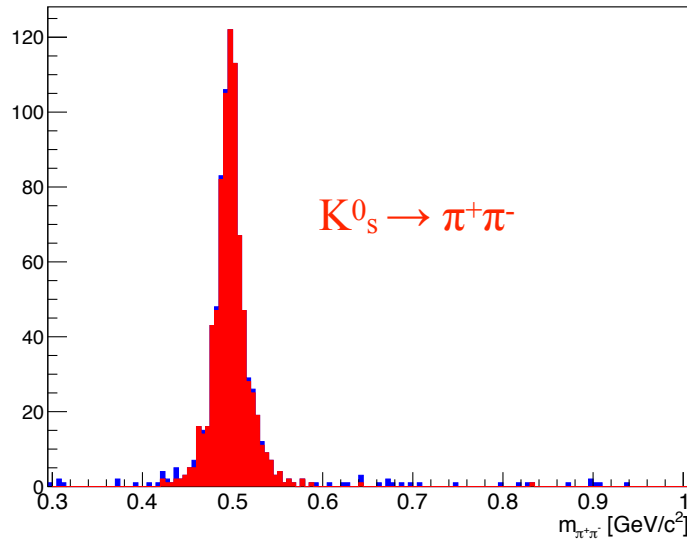
PANDA: Reconstruction of Strange Particles



Particle	p_t , GeV/c	Efficiency	Particle	p_t , GeV/c	Efficiency
K_s^0	1	56,2 %	Λ	1	32 %
Ξ^-	2	14,1 %	Ω^-	2	17,1 %

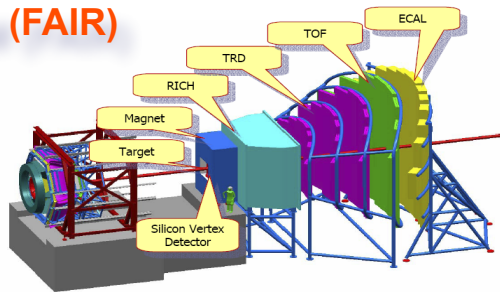
10000 signal events, Ideal track finder, MC primary vertex
Maksym Zyzak, 11.03.2014

KF Particle: Reconstruction of Decay Chains

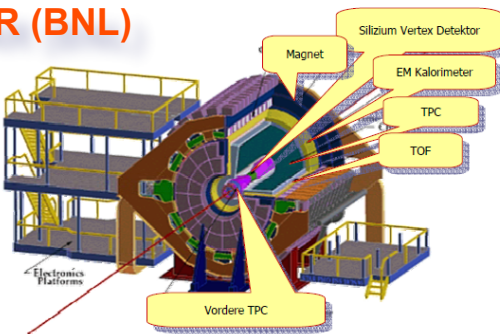


Concluding Remarks

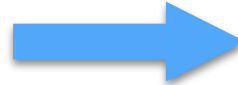
CBM (FAIR)



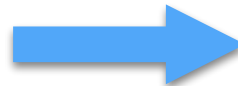
STAR (BNL)



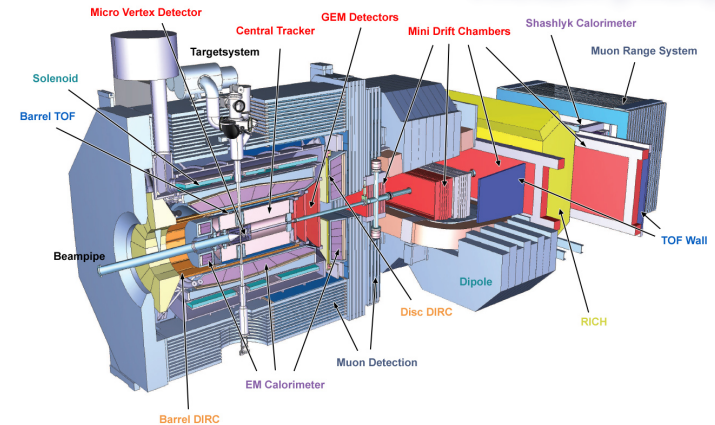
STS CA Track Finder
KF Track Fitter
KF Particle Finder



HFT CA Track Finder
KF Particle Finder



PANDA (FAIR)



1. The idea was to **share experience** between different experiments and thus to develop a **common reconstruction package** for detectors with **different technologies and geometries**.
2. The **Cellular Automaton** for searching for particle trajectories and the **Kalman Filter** to estimate their parameters have a **high level of intrinsic parallelism** for their efficient implementation on modern and future **many-core HPC architectures**.
3. The event reconstruction in real time requires to **redesign offline algorithms** for their **fast and reliable online operation**, and for the maximum utilisation of the online HPC farm. With the CA track finder we have reached the reconstruction speed of **1 ms/event/core**.
4. The **KF Particle Finder** package with more than 100 decay channels implemented (**100 μ s/core/decay**) is a common platform, both for the offline physics analysis and for the real-time express analysis.