

Conclusions from TrackML the HEP Tracking Machine Learning Challenge

Organization

Jean-Roch Vlimant (Caltech)

Vincenzo Innocente, Andreas Salzburger (CERN)

David Rousseau, Yetkin Yilmaz (LAL-Orsay)

Paolo Calafiura, Steven Farrell, Heather Gray (LBNL)

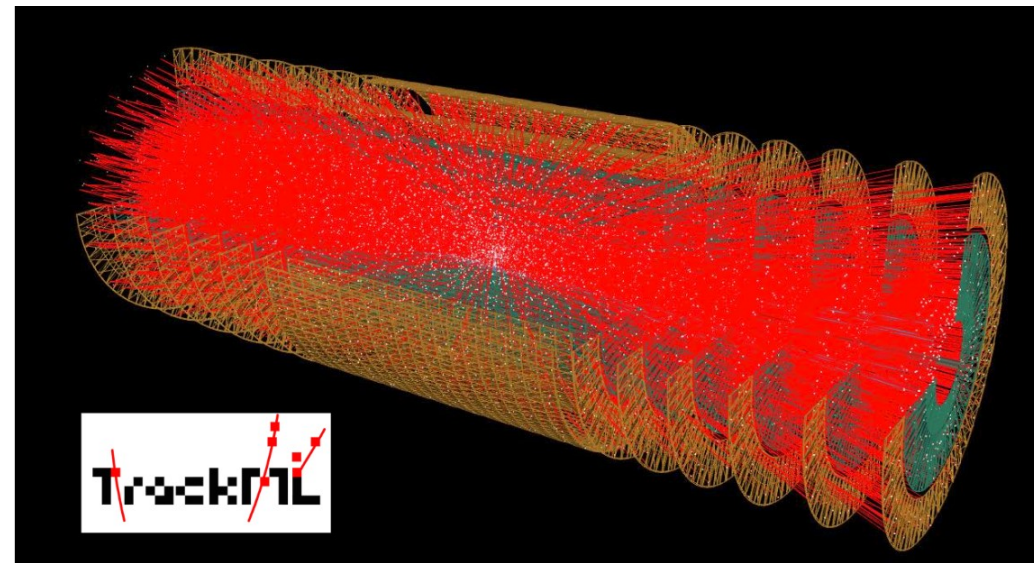
Vladimir Gligorov (LPNHE-Paris)

Sabrina Amrouche, Tobias Golling, Moritz Kiehn (Geneva University)

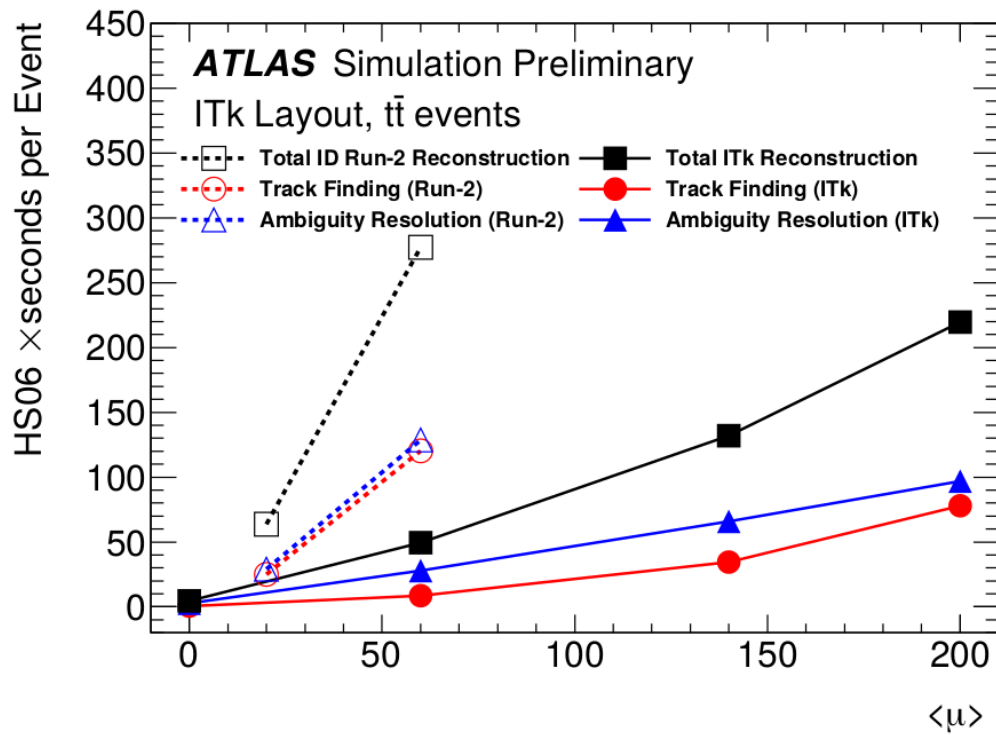
Laurent Basara, Cécile Germain, Isabelle Guyon, Victor Estrade (LRI-Orsay)

Edward Moyse (University of Massachusetts)

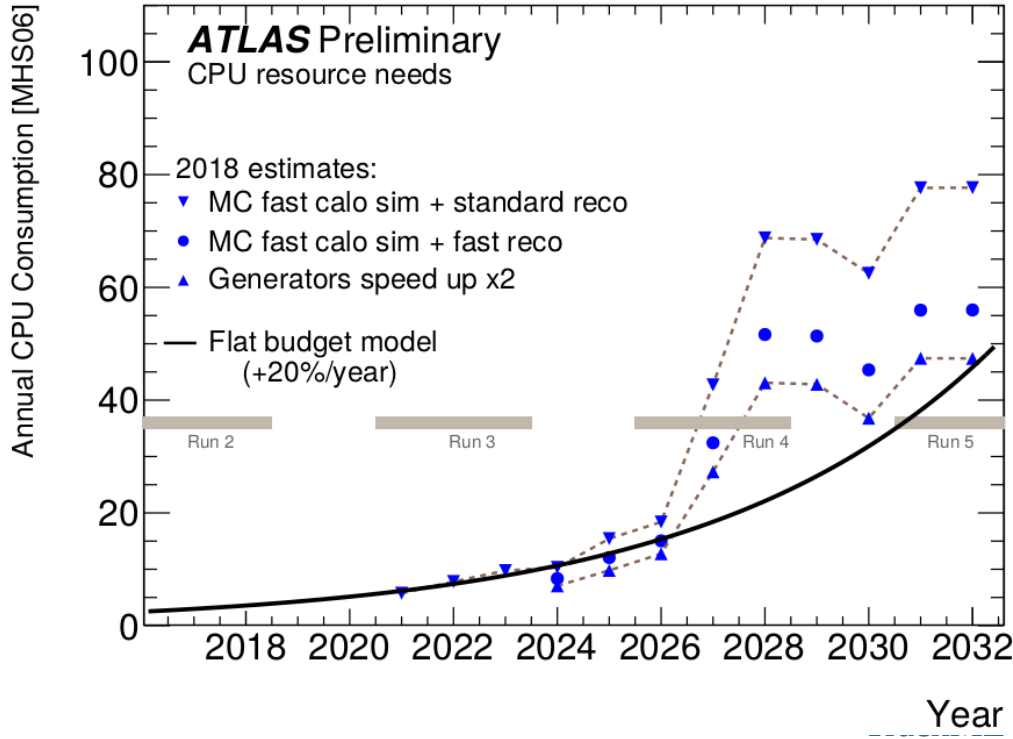
Mikhail Hushchyn, Andrey Ustyuzhanin (Yandex, HSE)



Tracking crisis



- Tracking dominates reconstruction CPU time
- At best quadratic
- HL-LHC (2025) : unmanageable
- Everything tried?
→ TrackML challenge



TrackML

sponsors



kaggle



NVIDIA



UNIVERSITÉ DE GENÈVE



Paris-Saclay Center for Data Science

INSTITUTE DATAIA
Data Science, Intelligence & Society



iris hep

Institute for Research & Innovation in Software for High Energy Physics

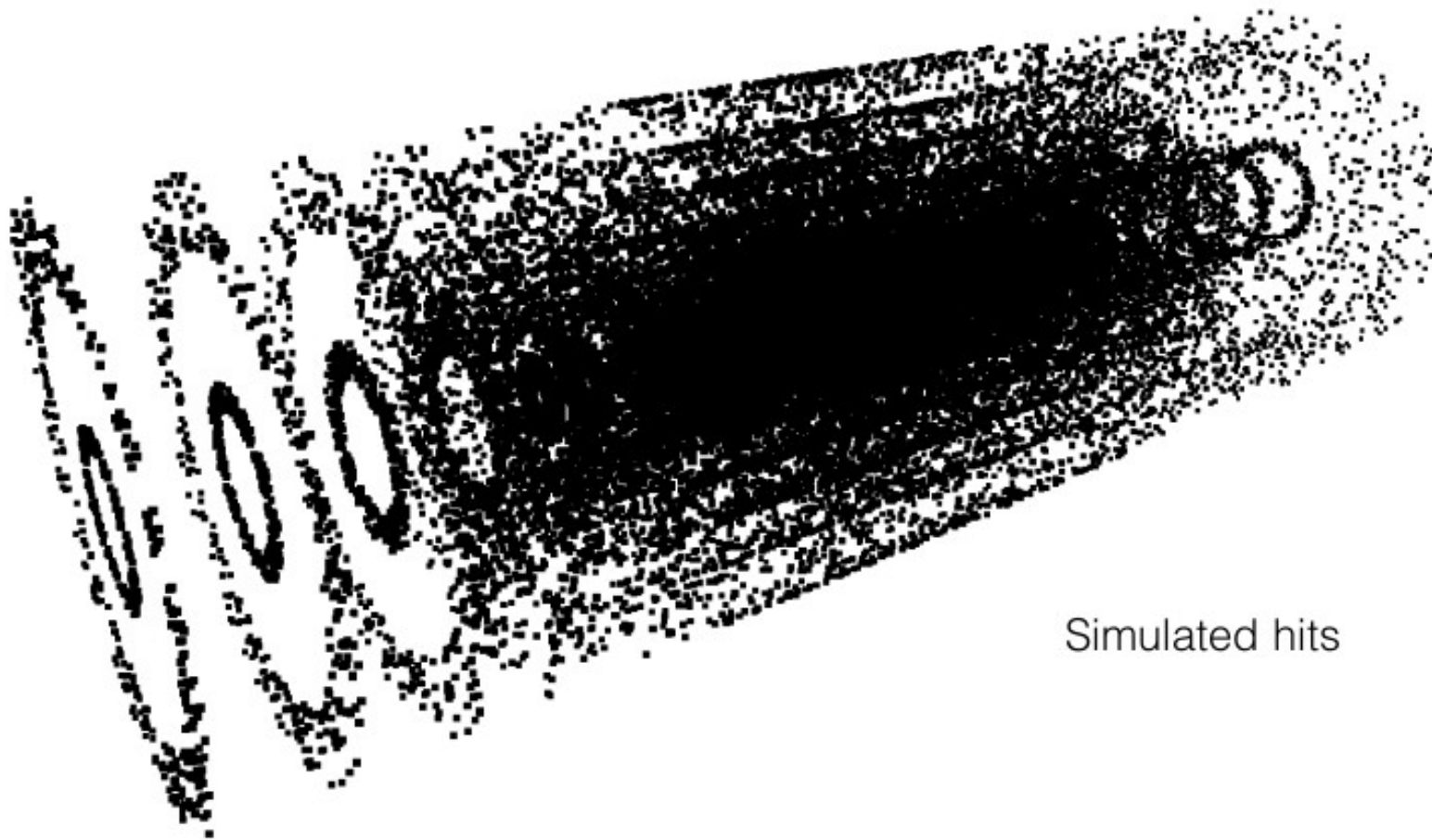


COMMON GROUND

inqnet

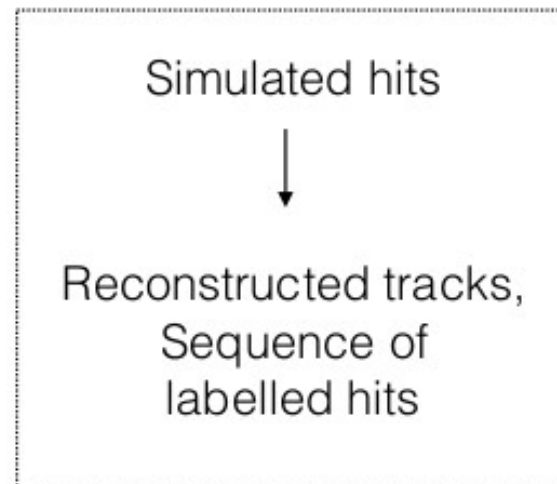
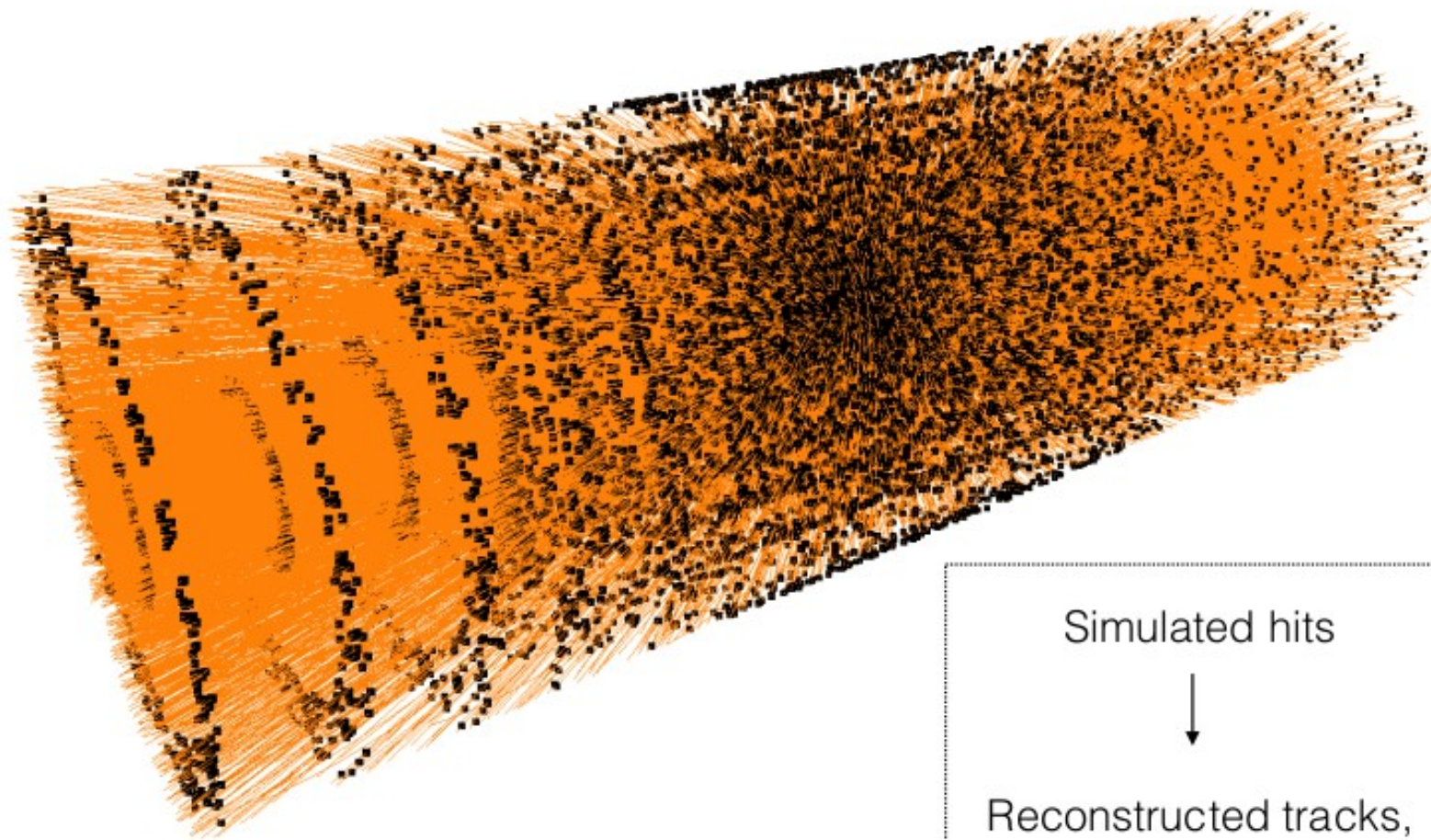
Alliance for Quantum Technologies





Simulated hits

See also in outreach session talk by D. Rousseau
« TrackML : the roller coaster of organizing a HEP challenge on Kaggle and Codalab »
<https://indico.cern.ch/event/577856/contributions/3423422/>




















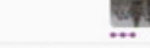


See also in outreach session talk by D. Rousseau
« TrackML : the roller coaster of organizing a HEP challenge on Kaggle and Codalab »
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First phase :
Accuracy
May – August 2018

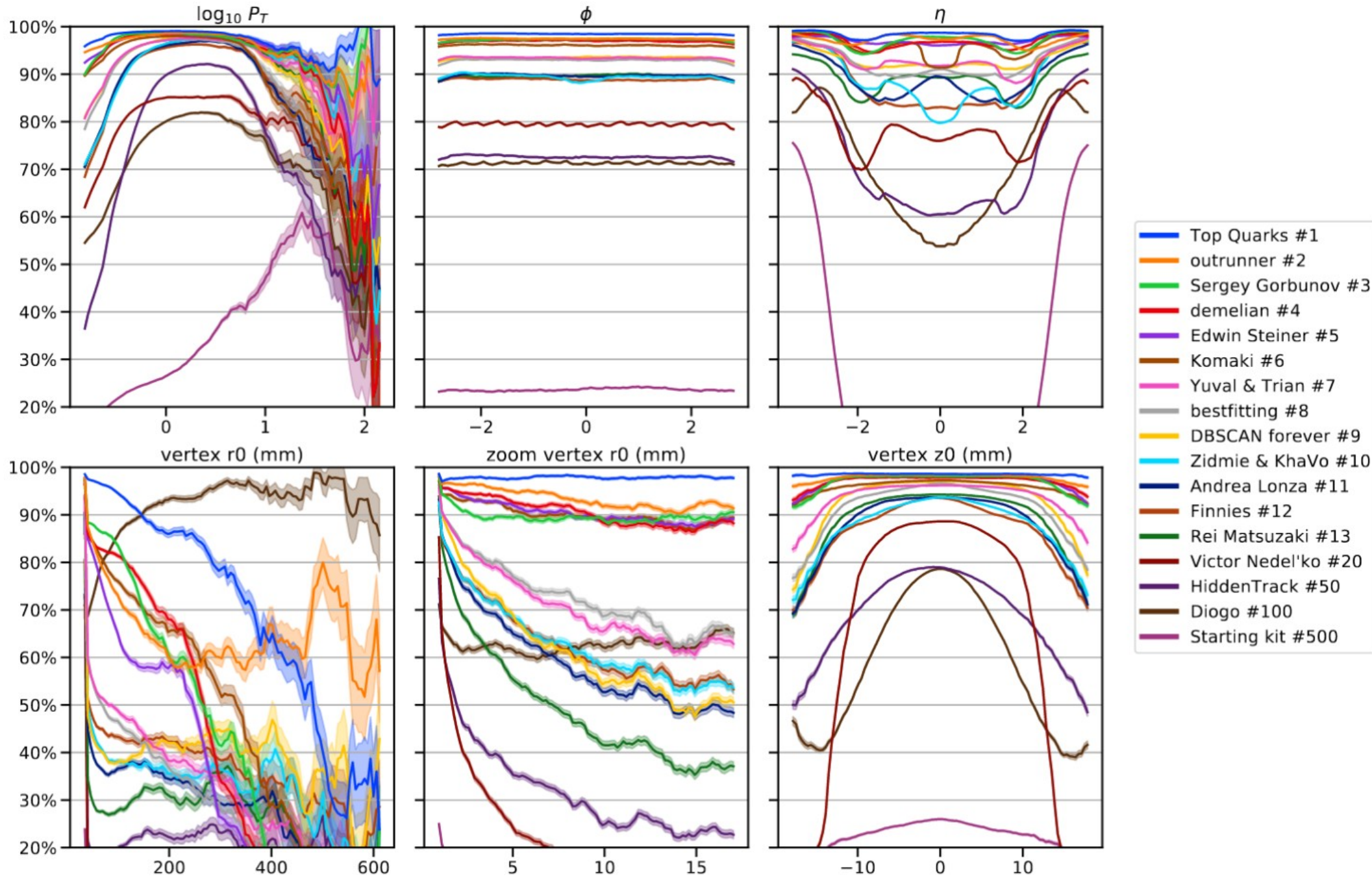
kaggle

Leaderboard scores

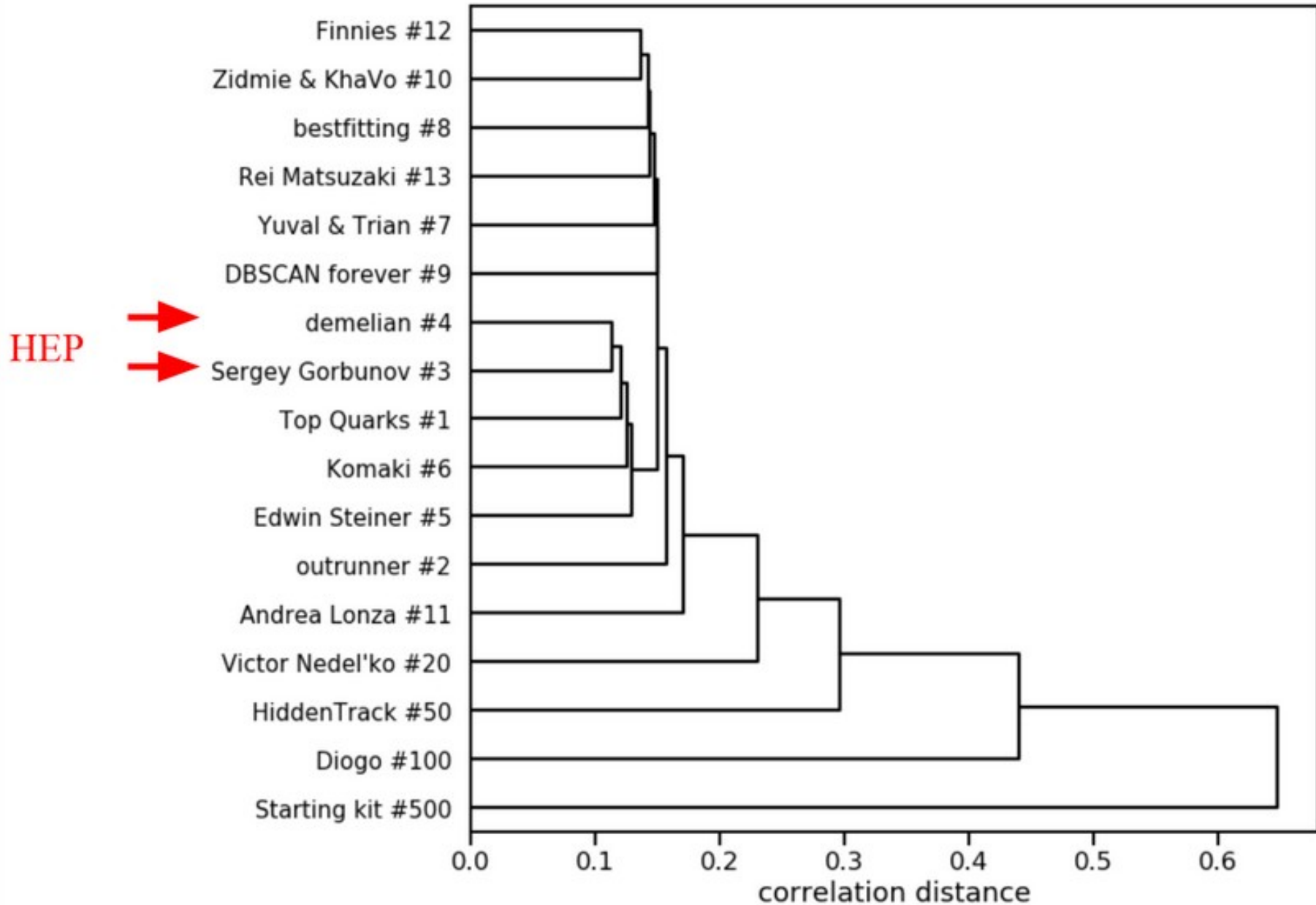
$$Score = \frac{1}{N} \sum_{test\ events} \sum_{good\ hits} weights$$

1	—	Top Quarks		0.92182	10	19d
2	—	outrunner		0.90302	9	18d
3	—	Sergey Gorbunov		0.89353	6	18d
4	—	demelian		0.87079	35	1mo
5	—	Edwin Steiner		0.86395	5	18d
6	—	Komaki		0.83127	22	18d
7	—	Yuval & Trian		0.80414	56	18d
8	—	bestfitting		0.80341	6	18d
9	—	DBSCAN forever		0.80114	23	18d
10	—	Zidmie & KhaVo		0.76320	26	18d
11	—	Andrea Lonza		0.75845	15	18d
12	—	Finnies		0.74827	56	18d
13	—	Rei Matsuzaki		0.74035	12	18d
14	—	Mickey		0.73217	10	2mo
15	—	Vicens Gaitan		0.70429	19	1mo
16	—	Robert		0.69955	3	21d
17	—	Yuval-CPMP tribute band		0.69364	20	20d
18	—	N. Hi. Bouzu		0.67573	9	22d
19	—	Steins;Gate		0.66763	12	19d
20	▲ 1	Victor Nedel'ko		0.66723	4	2mo

Optimizing score optimizes physics



Participants dendrogram



Phase 1 winner : Top Quarks

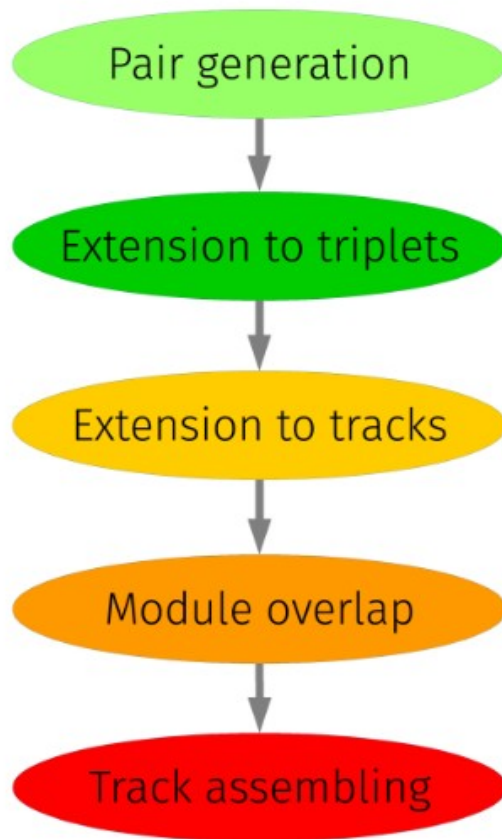


Illustration from J-R. Vlimant

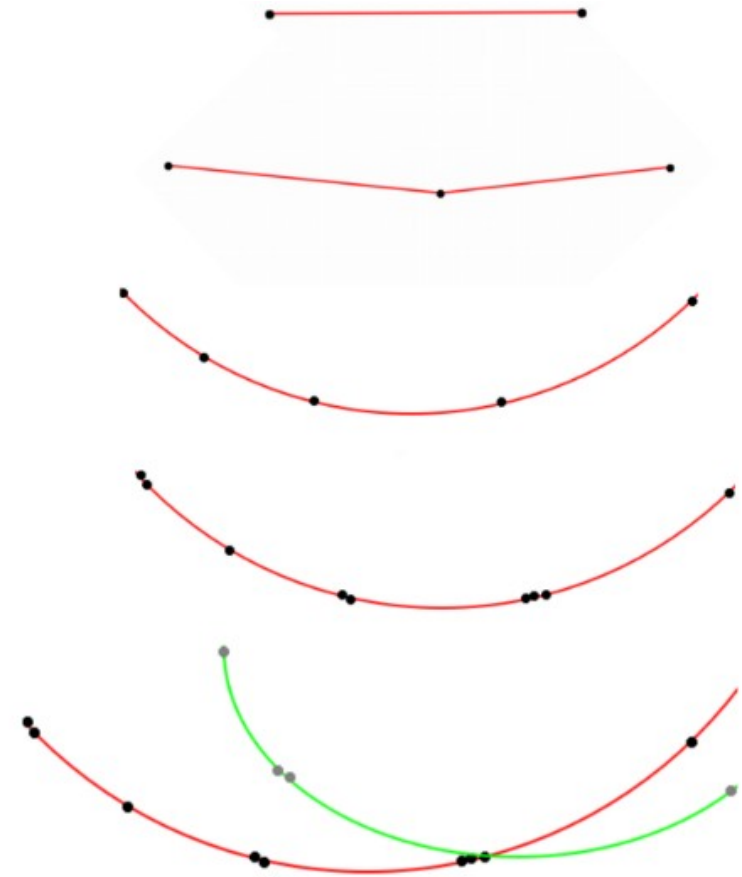
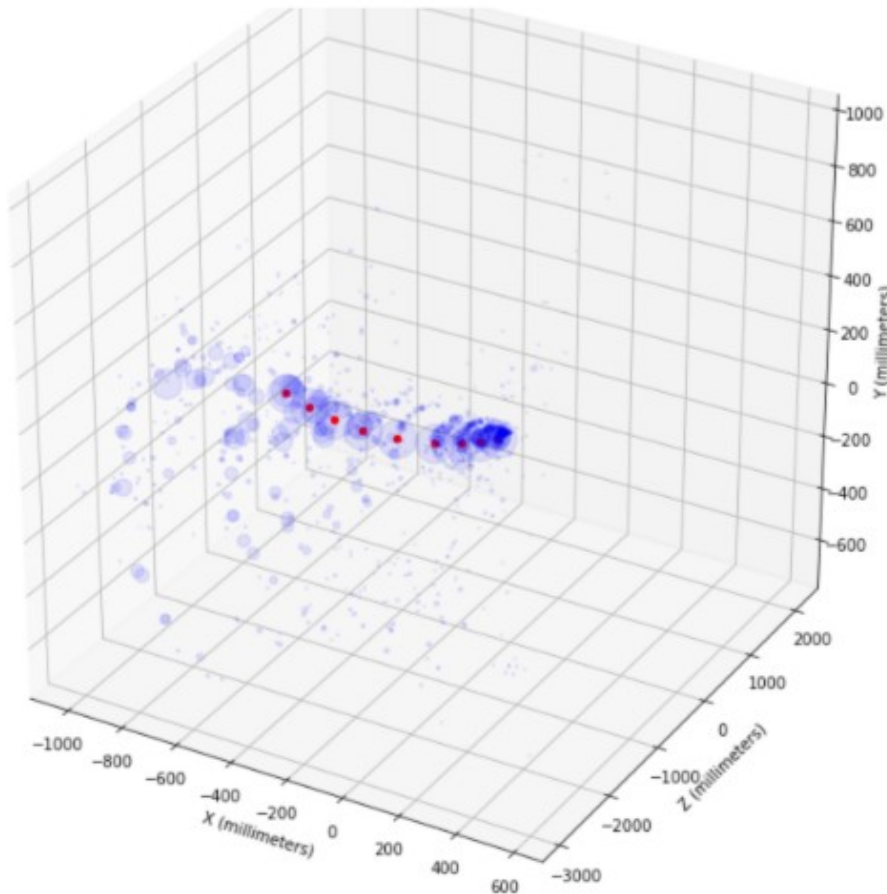


Illustration from J.S. Wind

Phase 1 #2 : outrunner



- Train DNN on hit pairs
 - 27 inputs (x,y,z,cells,...)
 - 4k-2k-2k-2k-1k hidden layers
- Compute full hit adjacency matrix:
 - probability $P(i,j)$ that 2 hits match
 - Pick high probability comb
- True Deep Learning Solution
 - No track following
 - No geometric modelling
- 1 Day / event

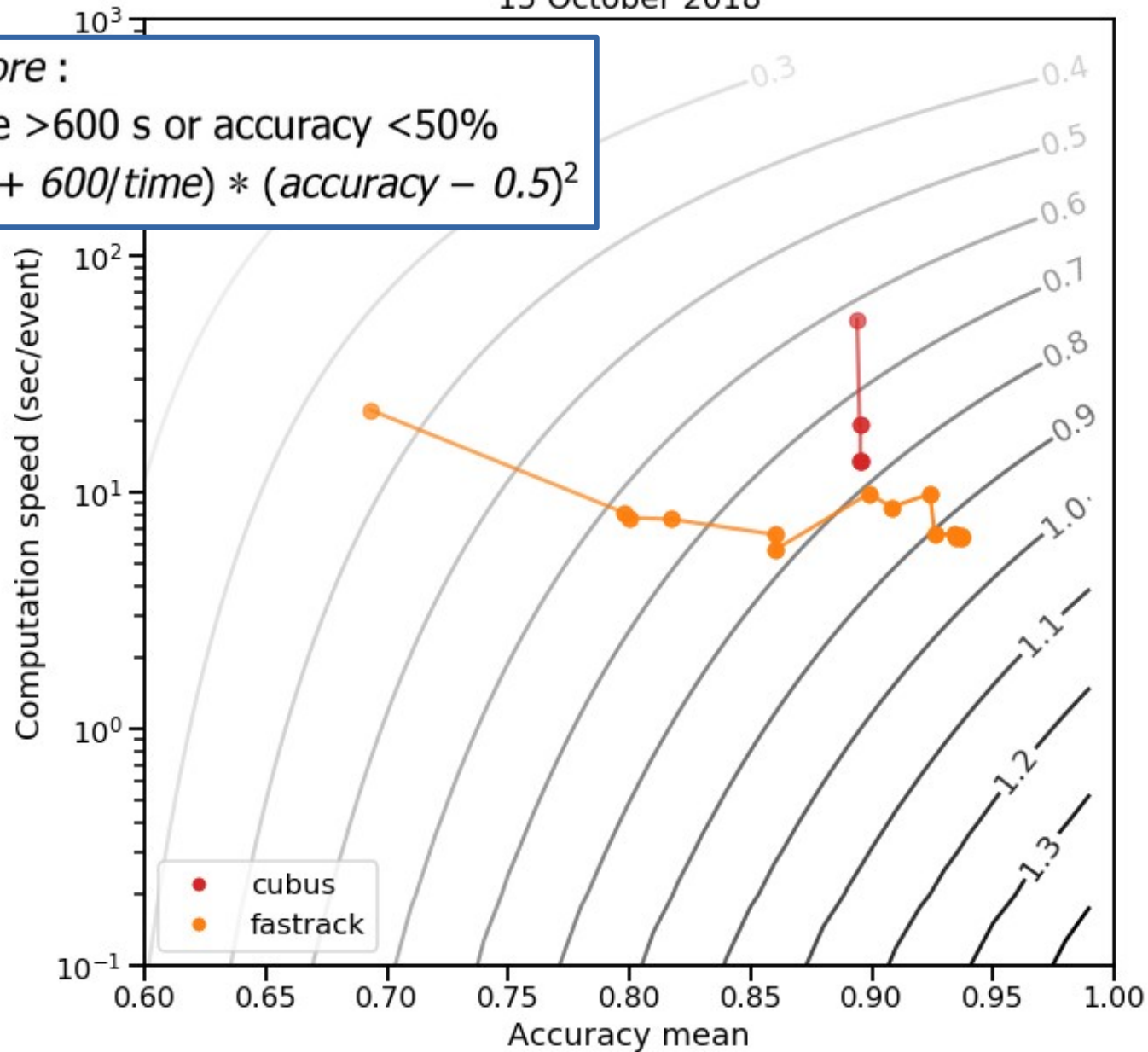
Second phase : Throughput Oct 2018 – March 2019



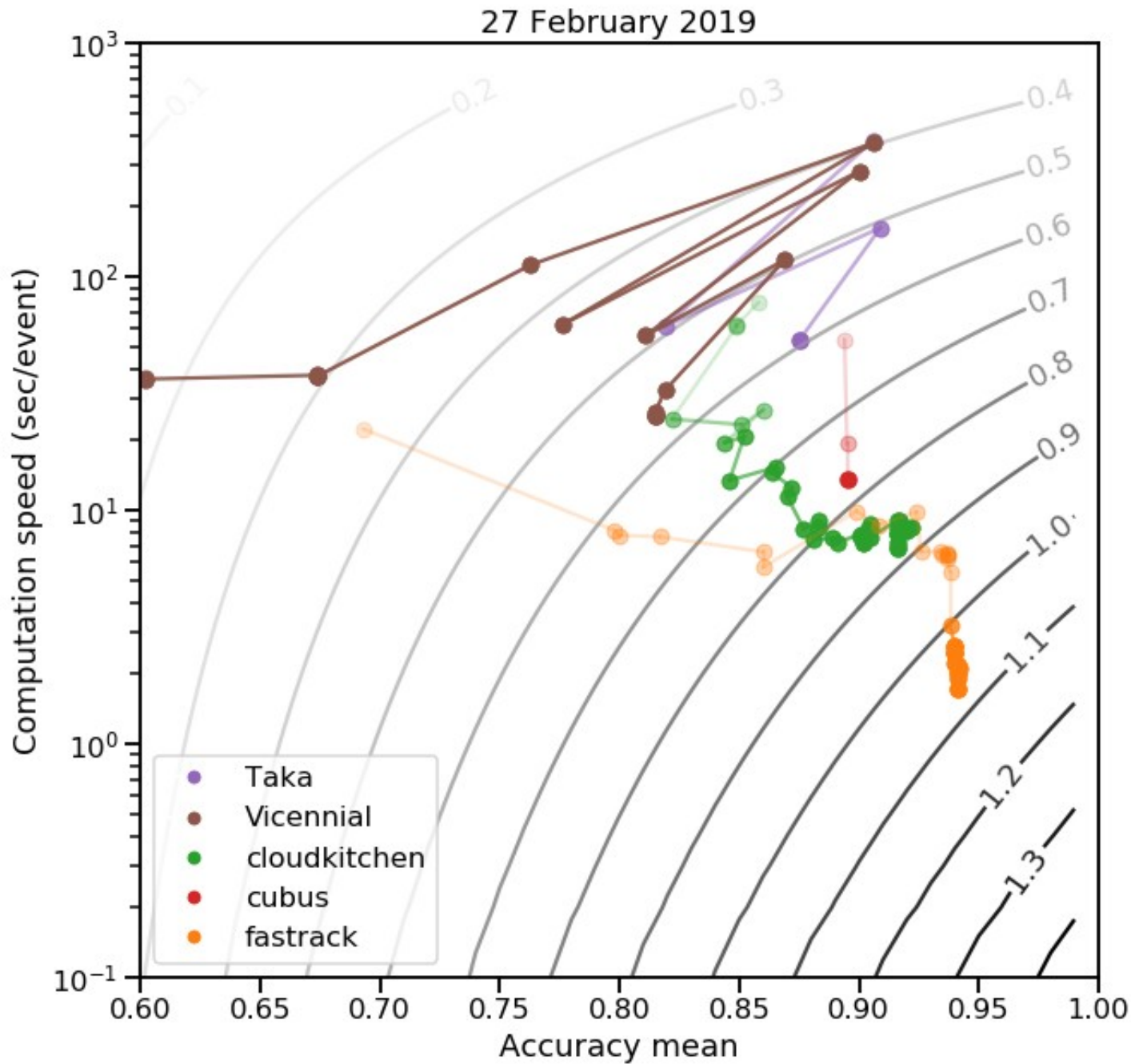
Leaderboard evolution

15 October 2018

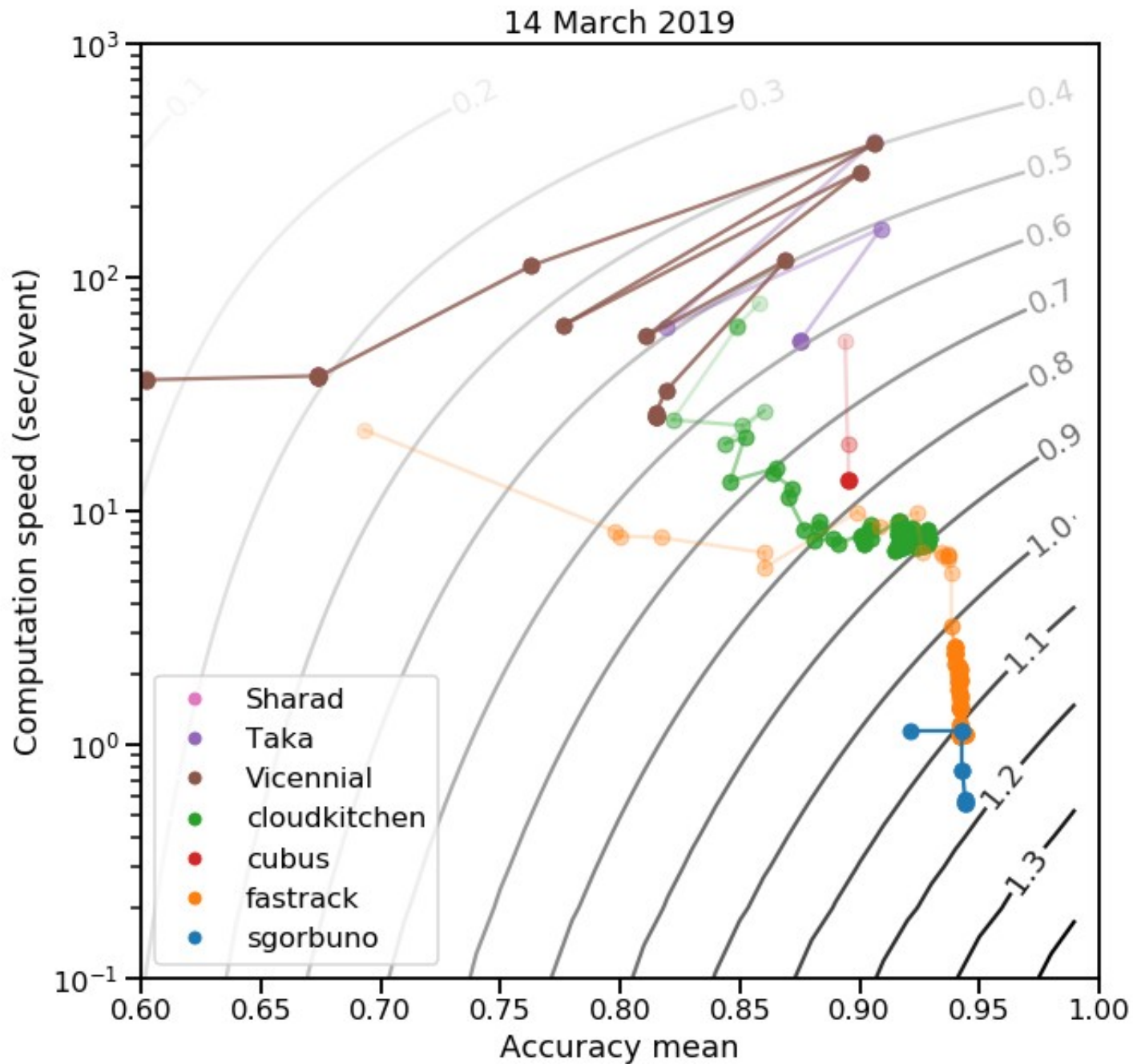
- Ranking score :
- 0 if time > 600 s or accuracy < 50%
 - $\sqrt{\log(1 + 600/\text{time})} * (\text{accuracy} - 0.5)^2$



Leaderboard evolution



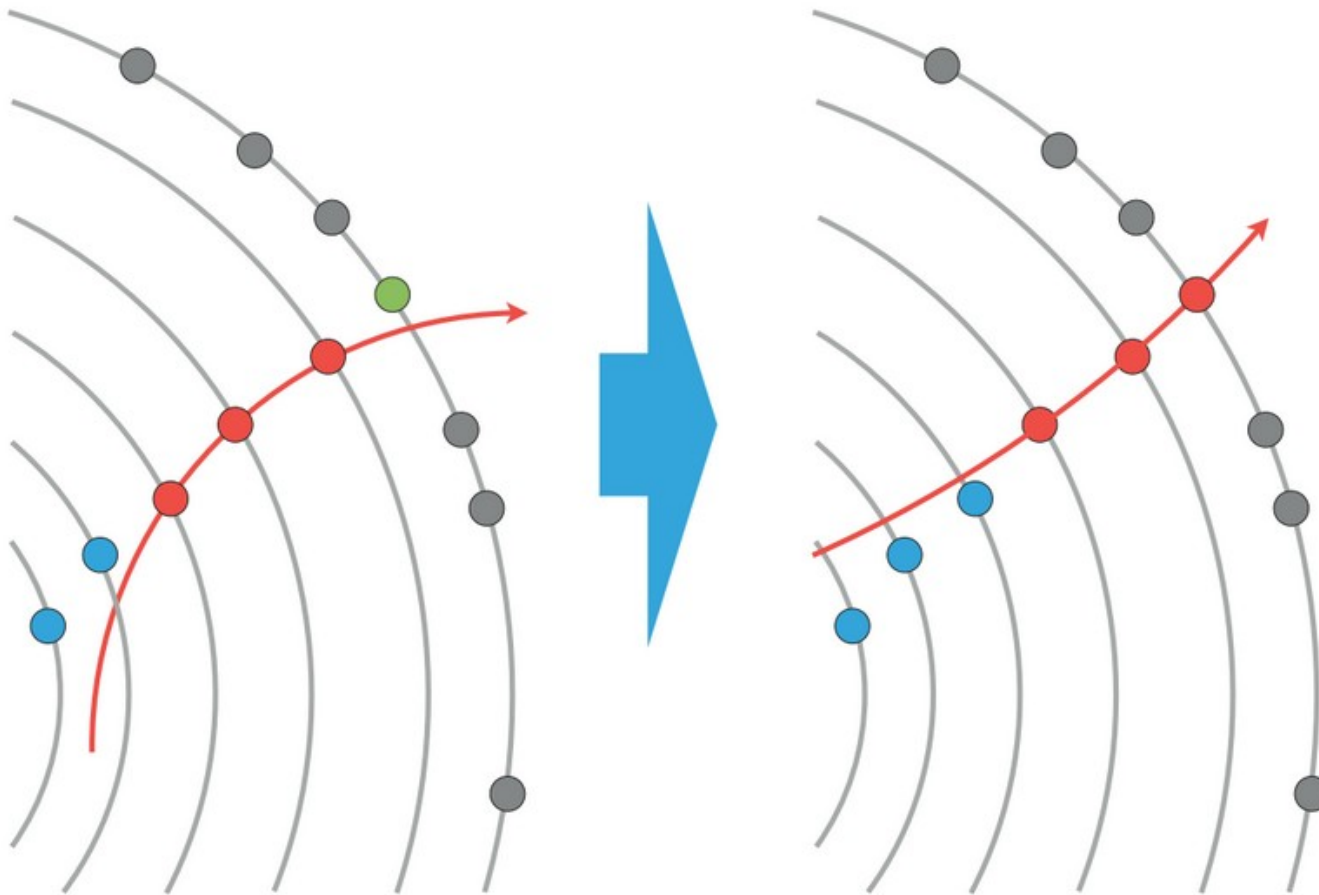
Leaderboard evolution



Final leaderboard

RESULTS									
#	User	Entries	Date of Last Entry	score ▲	accuracy_mean ▲	accuracy_std ▲	computation time (sec) ▲	computation speed (sec/event) ▲	Duration ▲
1	sgorbuno HEP	9	03/12/19	1.1727 (1)	0.944 (2)	0.00 (14)	28.06 (1)	0.56 (1)	64.00 (1)
2	fastrack HEP	53	03/12/19	1.1145 (2)	0.944 (1)	0.00 (15)	55.51 (16)	1.11 (16)	91.00 (6)
3	cloudkitchen	73	03/12/19	0.9007 (3)	0.928 (3)	0.00 (13)	364.00 (18)	7.28 (18)	407.00 (8)
4	cubus	8	09/13/18	0.7719 (4)	0.895 (4)	0.01 (9)	675.35 (19)	13.51 (19)	724.00 (9)
5	Taka	11	01/13/19	0.5930 (5)	0.875 (5)	0.01 (12)	2668.50 (23)	53.37 (23)	2758.00 (13)
6	Vicennial	27	02/24/19	0.5634 (6)	0.815 (6)	0.01 (10)	1270.73 (20)	25.41 (20)	1339.00 (10)
7	Sharad	57	03/10/19	0.2918 (7)	0.674 (7)	0.02 (4)	1902.20 (22)	38.04 (22)	1986.00 (12)
8	WeizmannAI	5	03/12/19	0.0000 (8)	0.133 (11)	0.01 (11)	88.08 (17)	1.76 (17)	124.00 (7)

#1 S. Gorbunov: « fast combinatorial »



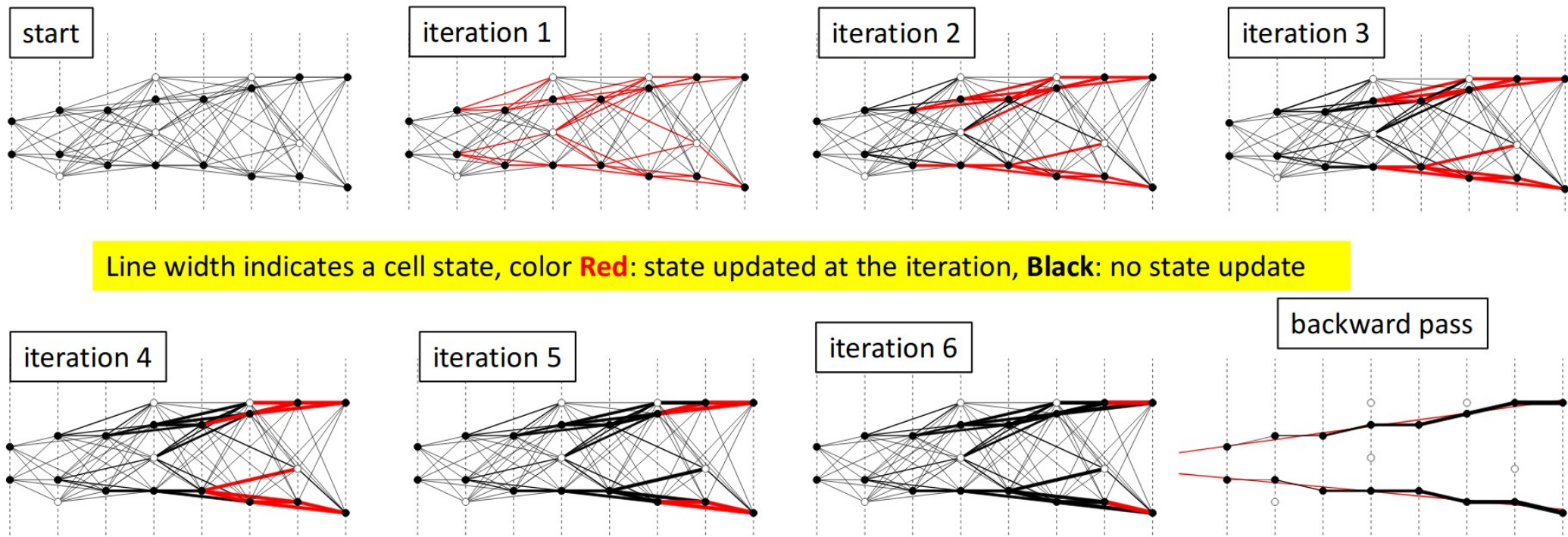
TRACK MODEL

- ▶ local 3-hit helix

TRACK PROLONGATION

- ▶ cross the next layer
- ▶ pick the best hit
- ▶ refit with the new hit

#2 FASTrack: Graph of neighbours, cellular automata and Kalman filter



https://indico.cern.ch/event/813759/contributions/3479706/attachments/1870758/3078234/TheTrackML_workshop_talk.pdf

Phase 2 cloudkitchen

Author: Marcel Kunze

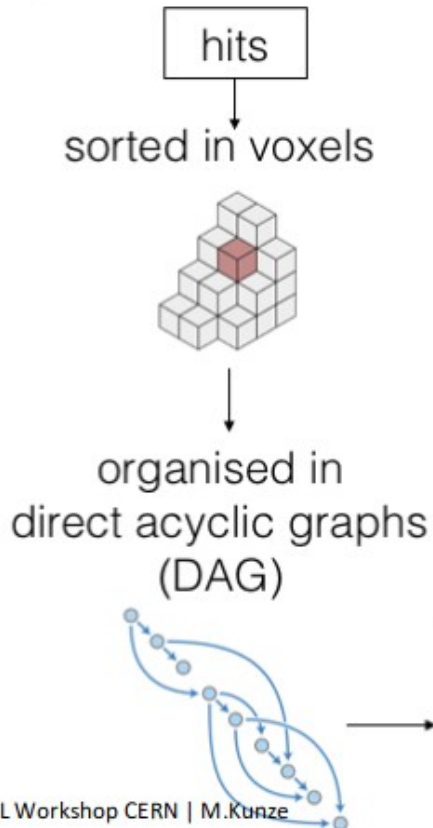


Accuracy: 0.93
Time/event: ~7 sec
Memory: 0.7 Gb

partly based on top quarks Phase 1 solution

1 - Top Quarks 0.92182 10 2mo

Algorithm outline



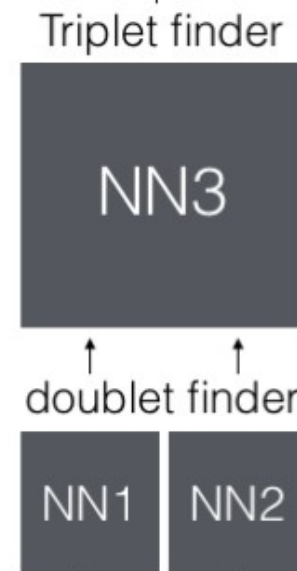
Main steps

- Select promising pairs
 - 7 million / 0.99
- Extend pairs to triples
 - 12 million / 0.97
- Extend triples to tracks
 - 12 million / 0.95
- Add duplicate hits to tracks
 - 12 million / 0.96
- Assign hits to tracks
 - 90% of hits / 0.92

DAGs are pre-trained on ~25 events ground truth

DAGs are used to fast navigate through voxel space

- Disc section
- Tube section



ca. 300k
97.2%

ca. 500k
99.4%

ca. 2 Mio.

Conclusions

- Open tracking competition organised to reach out to CS and ML communities
- Winner and runner-up HEP tracking experts...
- Retained solution will be blend from HEP expertise and new ideas
- Dataset released on CERN Open Data Portal to serve as benchmark
- Ongoing work

Contacts

- Contact : trackml.contact@gmail.com
- <https://sites.google.com/site/trackmlparticle>
- Twitter : @trackmlhc
- Accuracy phase @ Kaggle:
<https://www.kaggle.com/c/trackml-particle-identification>
 - Chapter in the NeurIPS2018 Competition book arXiv:1904.06778
- Throughput phase @ Codalab:
<https://competitions.codalab.org/competitions/20112>
 - Write-up to be finalised
- TrackML challenge Grand Finale: <https://indico.cern.ch/event/813759/>

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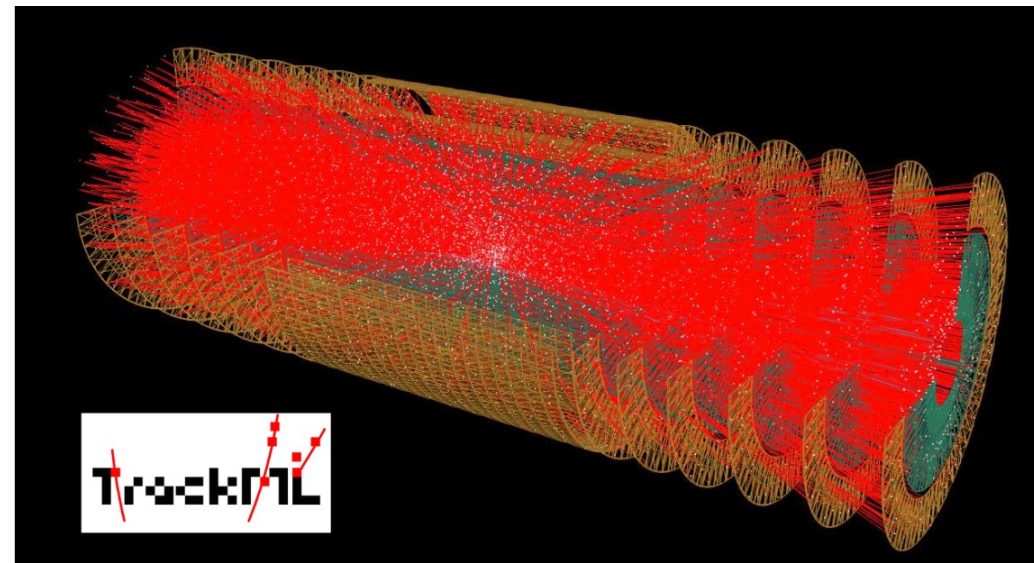
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Phase 2 Mikado



Author: Sergey Gorbunov



Accuracy: 0.944

Time/event: 0.56 sec

Memory: 0.1/0.178 Gb (1core/2 cores)

third in Phase-1

Based on Phase-1 algorithm

- runs iteratively in **80 passes**

& **hit removal** from high to low pT

modifications with respect to Phase 1

search branches enabled

every pass has optimised parameters

results in $O(10^4)$ parameters to be tuned,

tuning done semi-automated

Phase 1 Sergey Gorbunov

Execution time
1.2 min on single core 2.6 GHz CPU

Summary

- A combinatorial algorithm, based on the track following method
- No search branches
- Simple track model: local 3-hit helix
- Fast data access

Primary tracklets

Regular grid with overlaps

Prolongation of tracklets

XY



Phase 2 FASTrack

Author: Dmitry Emelianov



Accuracy: 0.944

Time/event: 1.11 sec → 0.8 sec

Memory: 0.6 Gb recently down to

first runner-up to podium in Phase-1

4	-	demelian		0.87079	35	2mo
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Algorithm outline

Phase-1 w/o measurement shapes

- using measurement shapes to predict intervals of track inclination
- segment based track following network with embedded Kalman Filter
 - **connection graph** pre-build (&compiled) from `Detector.csv` file
 - run with a **Cellular Automaton (CA)**, **parallelised** with **OpenMP**
 - **candidate building**: graph traversal with applied simplified KF
- combinatorial track following for track completion
 - fast **combinatorial** Kalman Filter using **3rd order RK** & **simplified field** includes **clone identification** & **track merging**

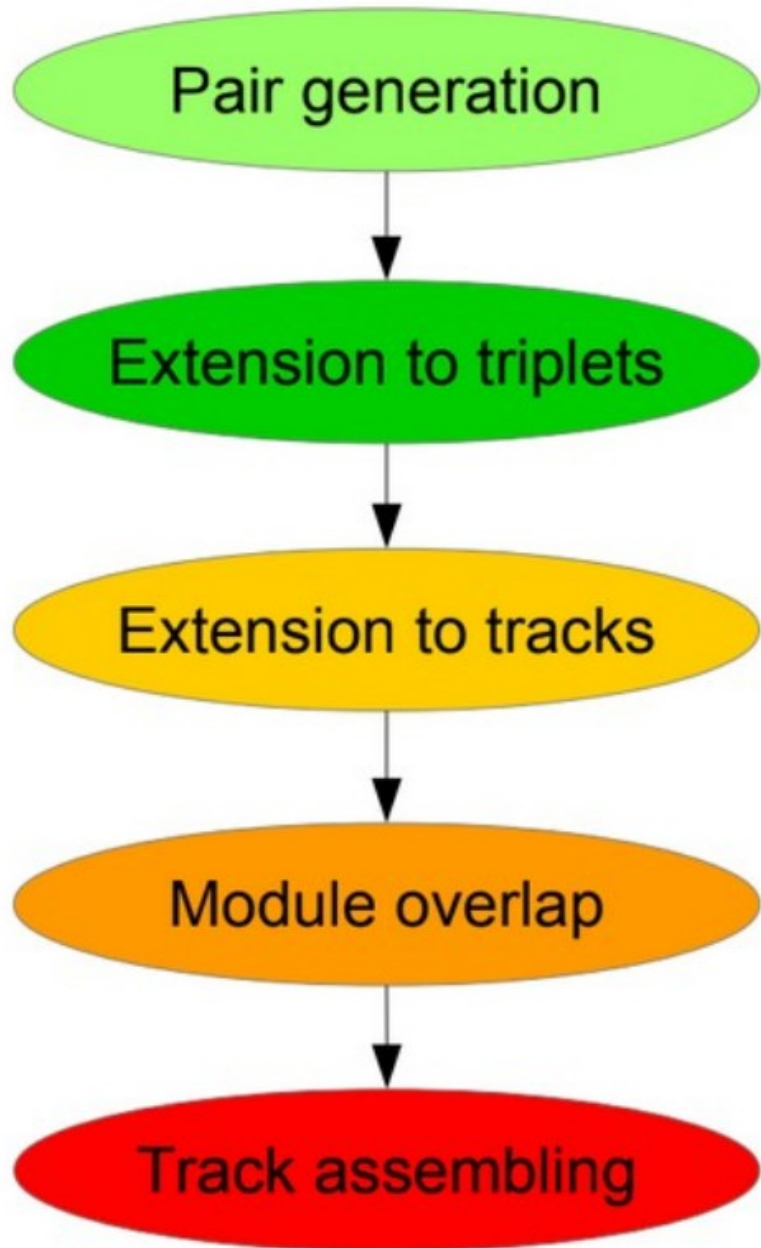
3 passes (hit removal):

- high momentum
- low momentum
- rest

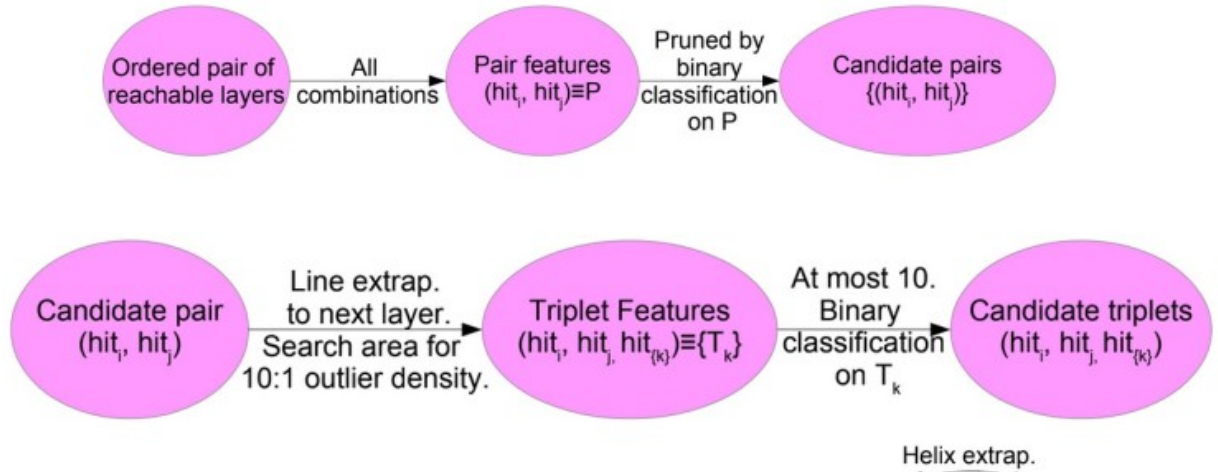
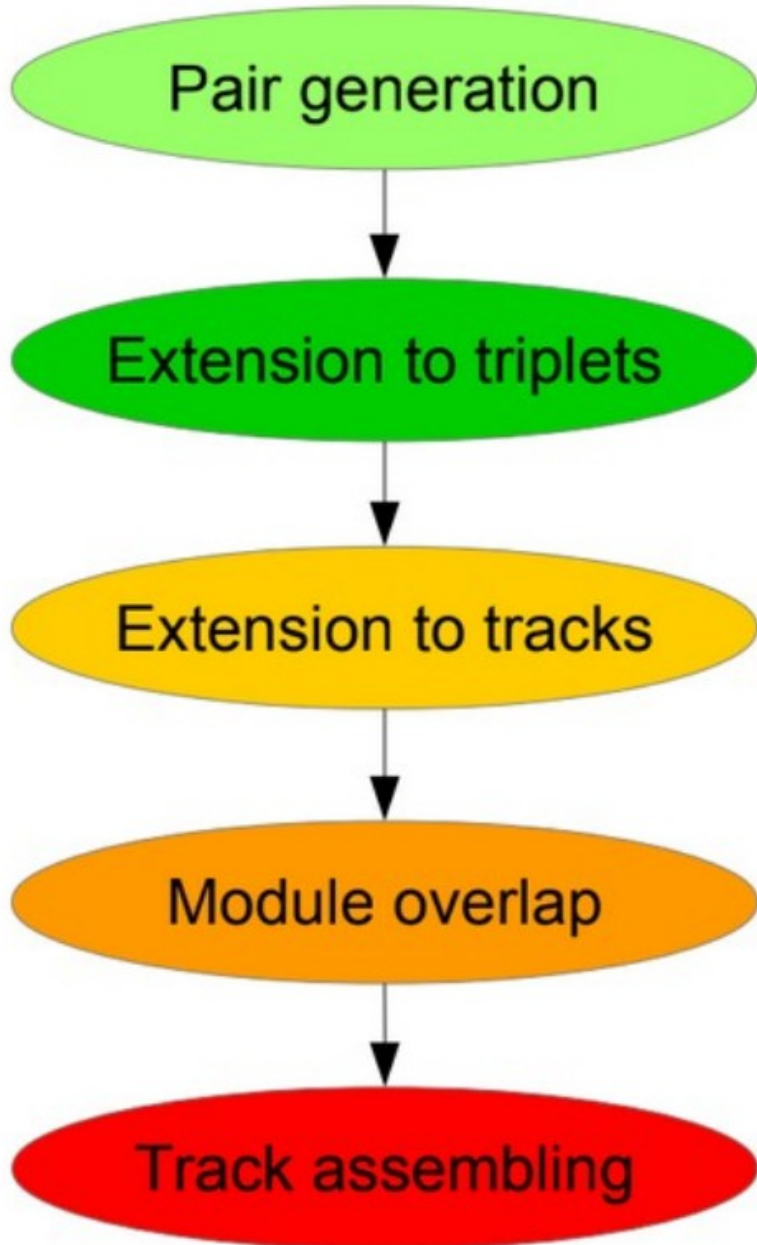
TrackML, David Rousseau, IML, 16th April 2019

TrackML conclusion

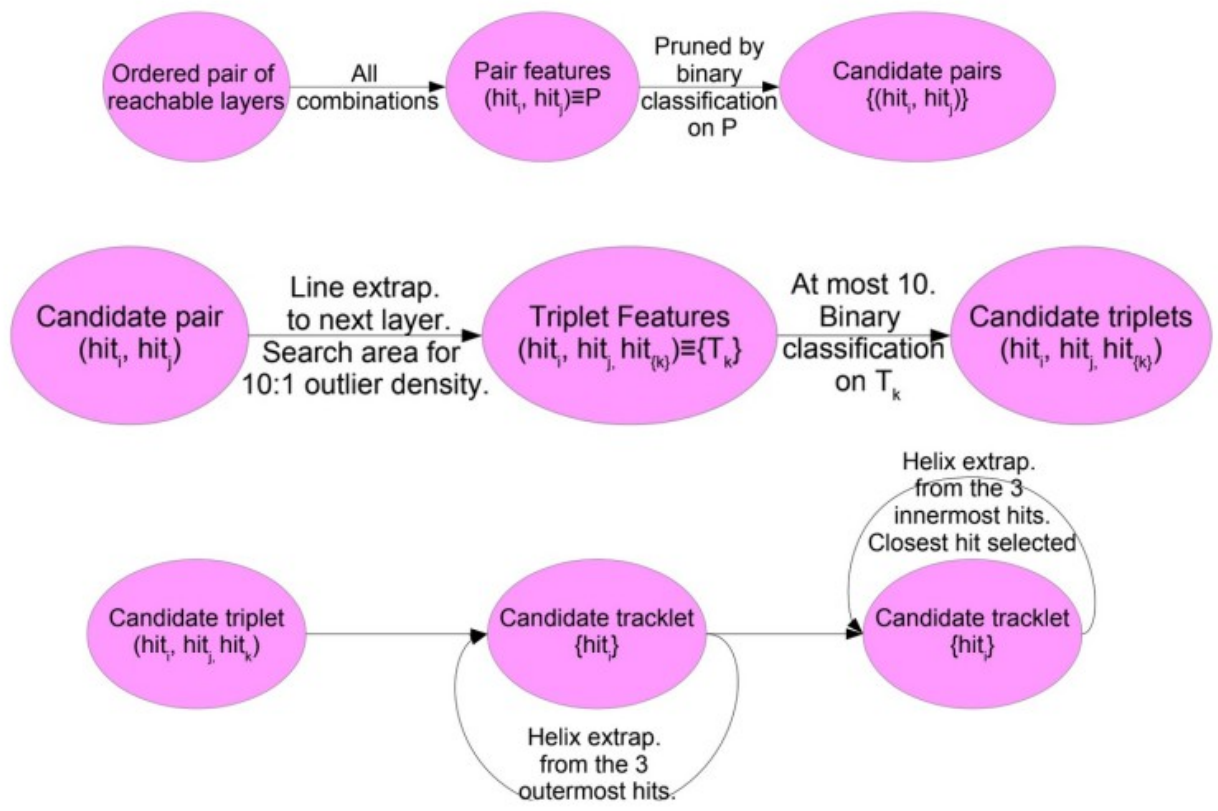
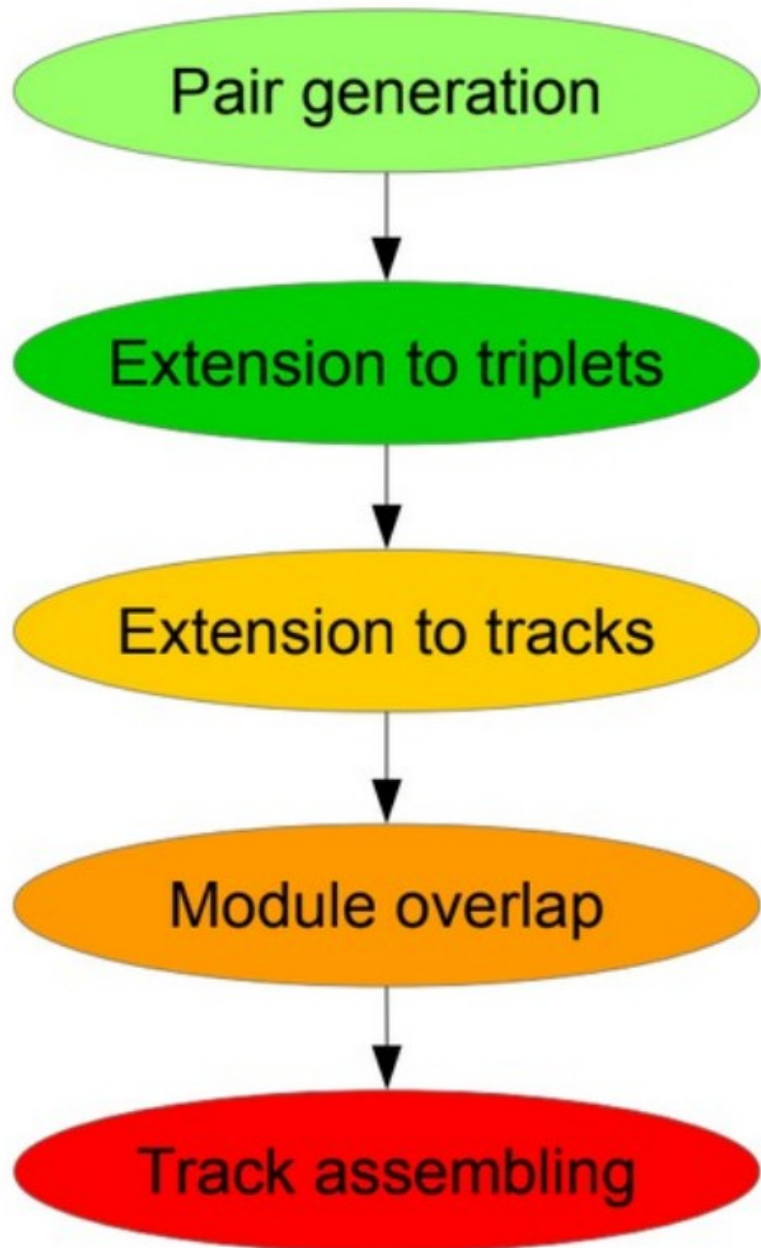
Winning solution → not ML



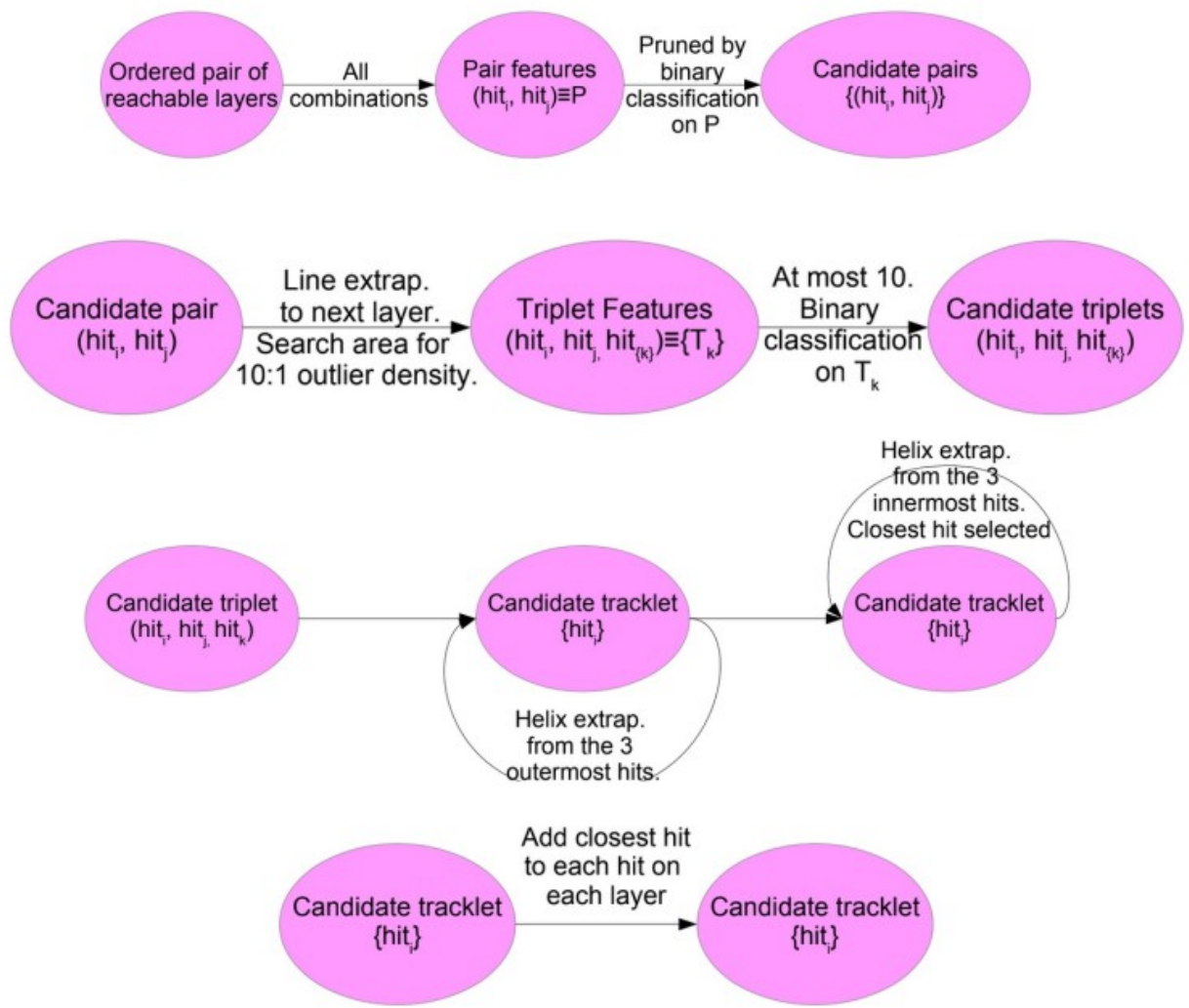
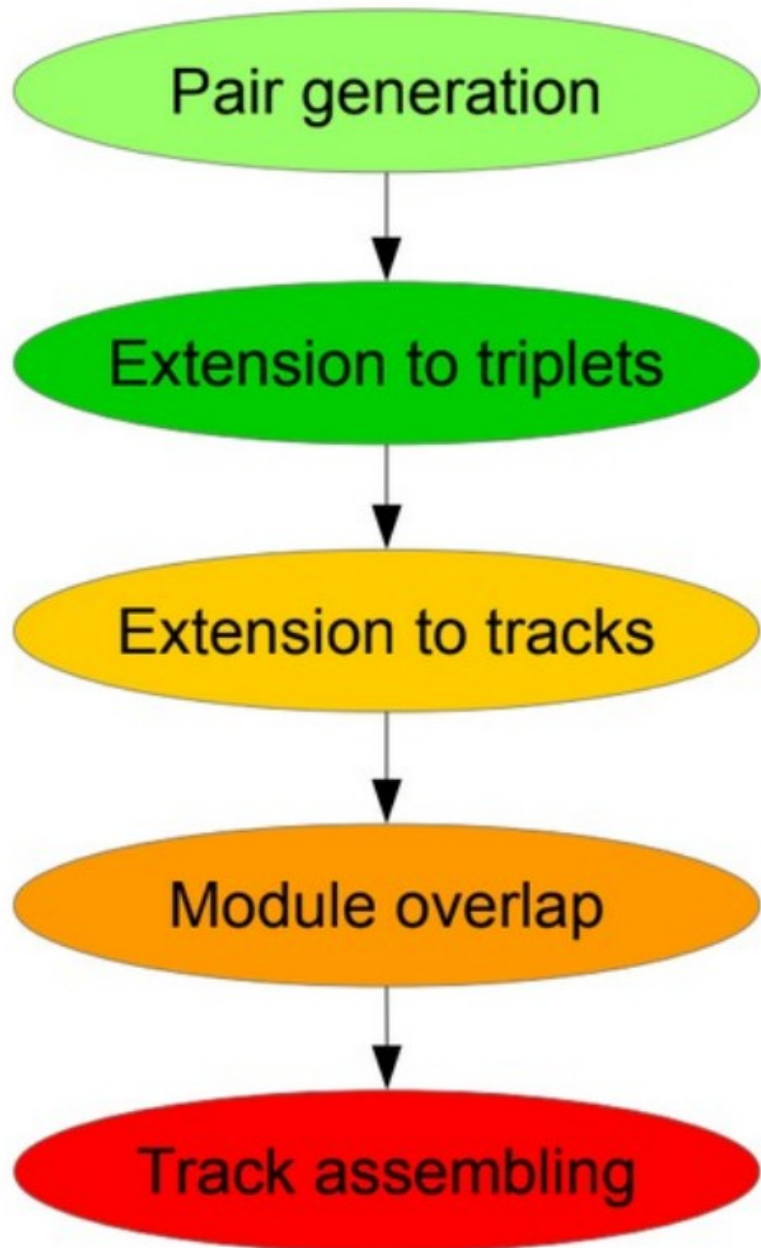
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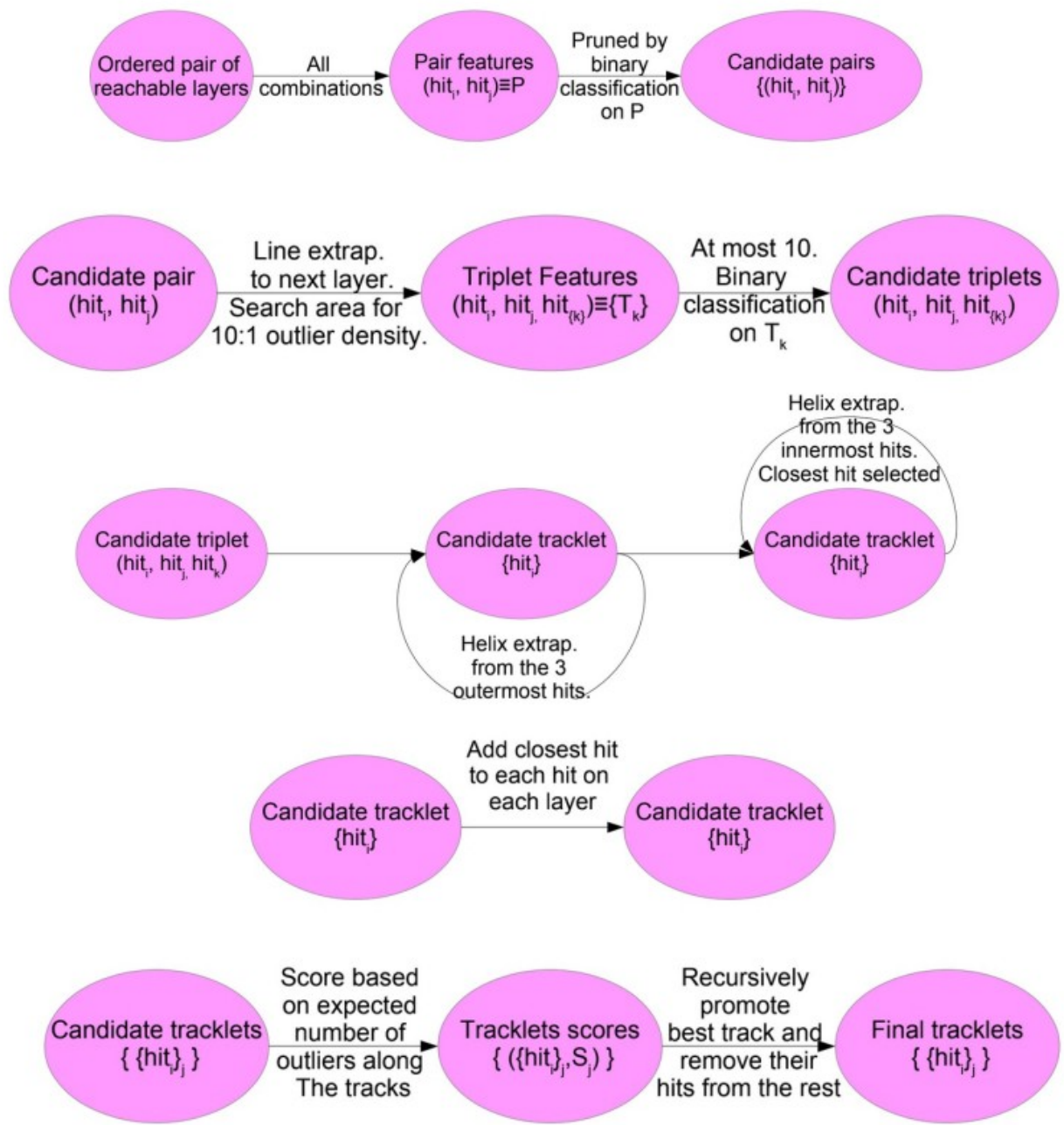
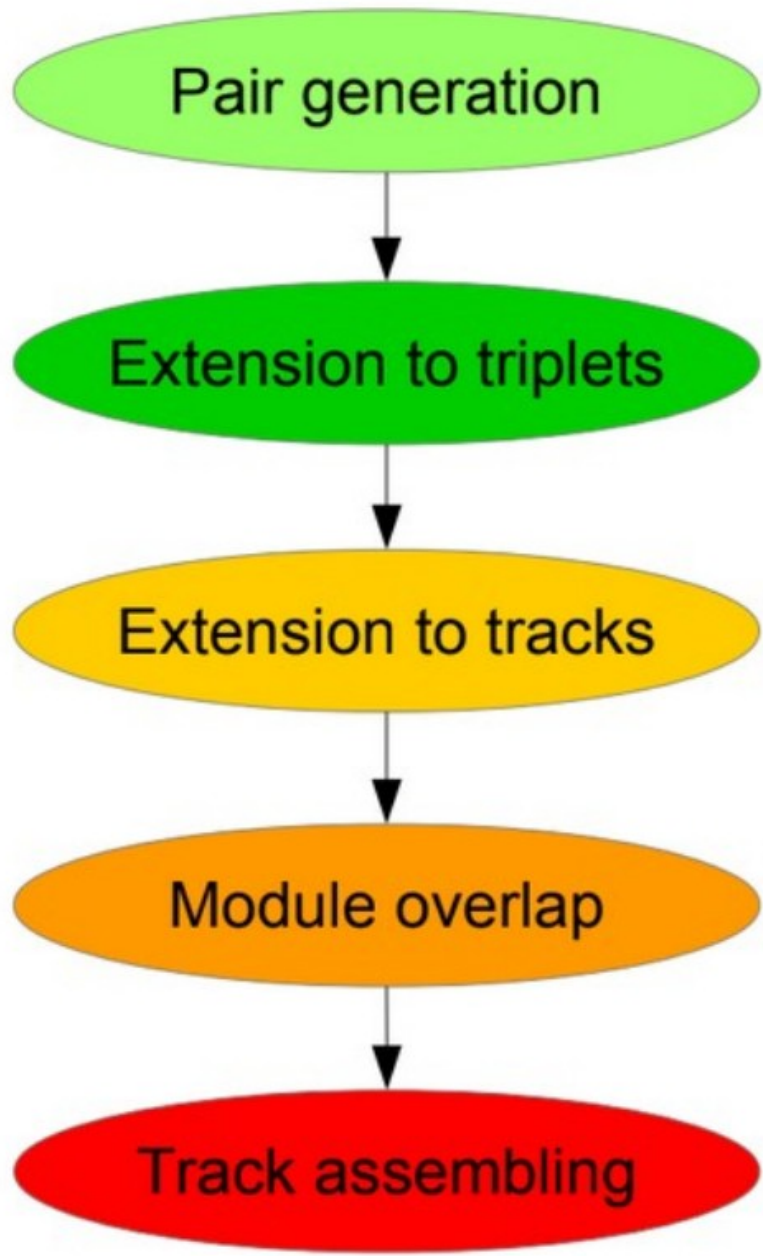
Winning solution → not ML



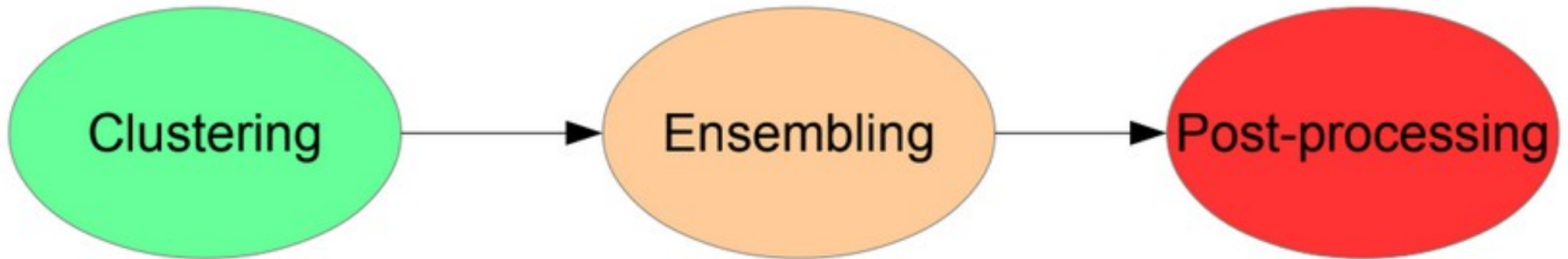
Winning solution → not ML

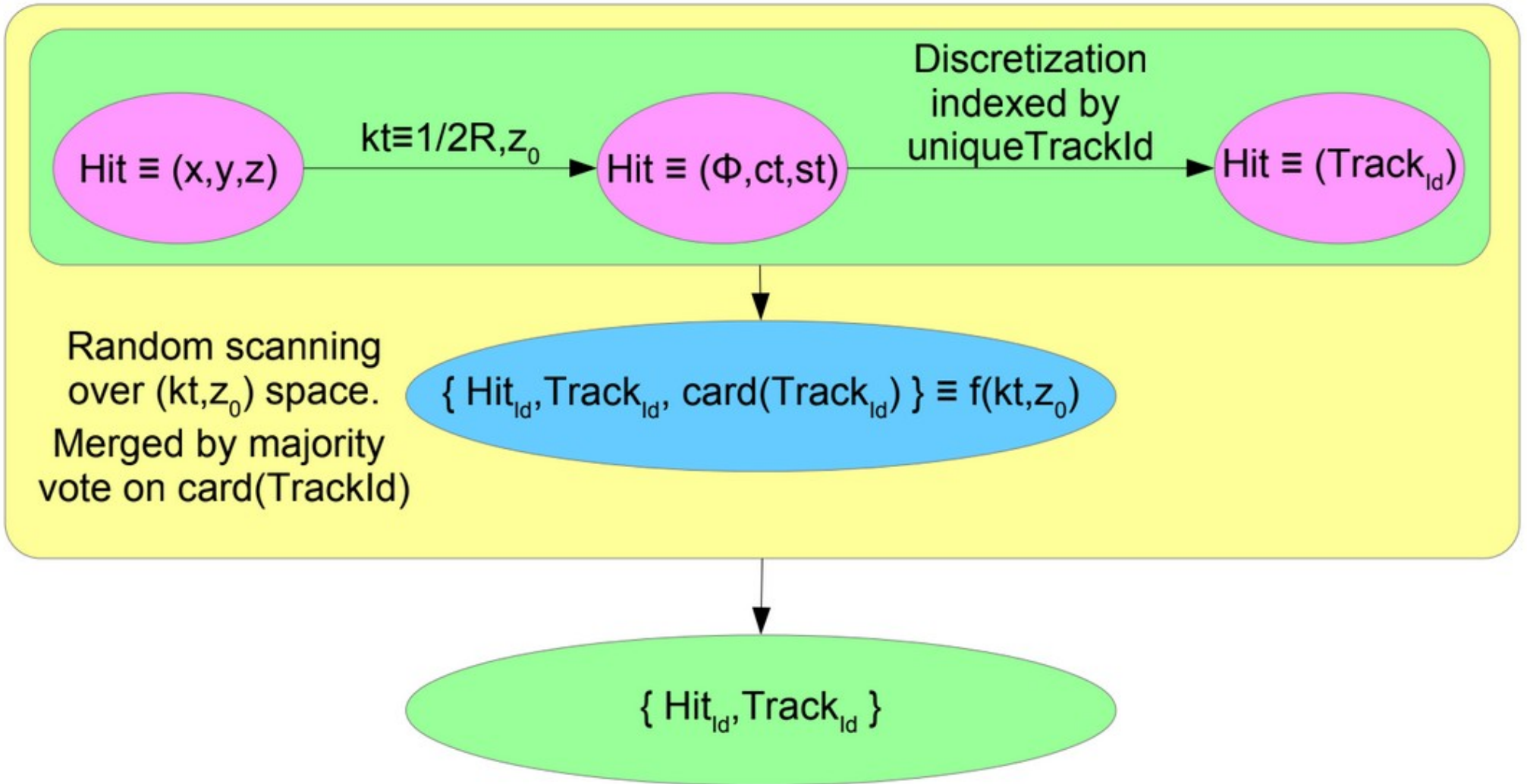


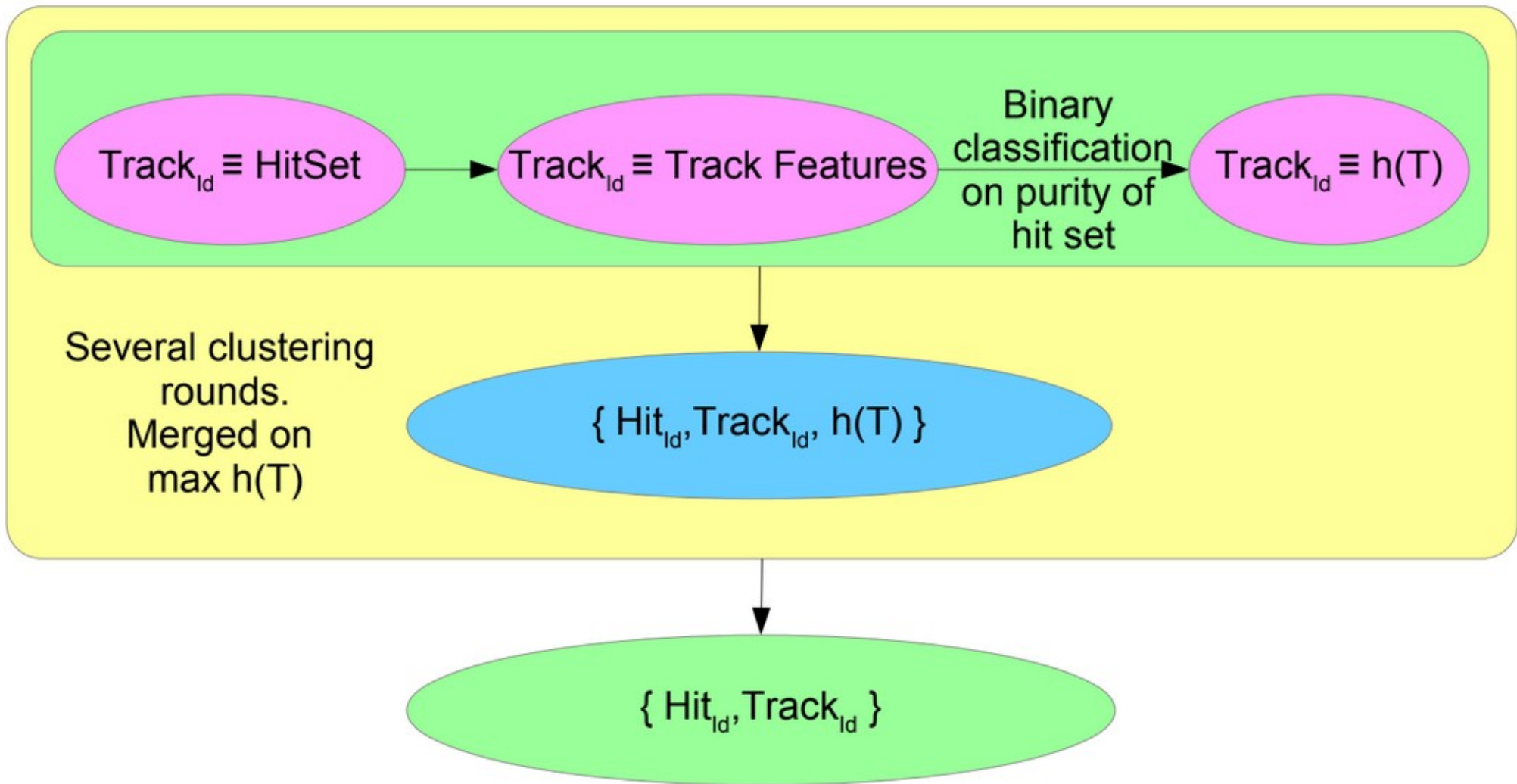
Winning solution → not ML

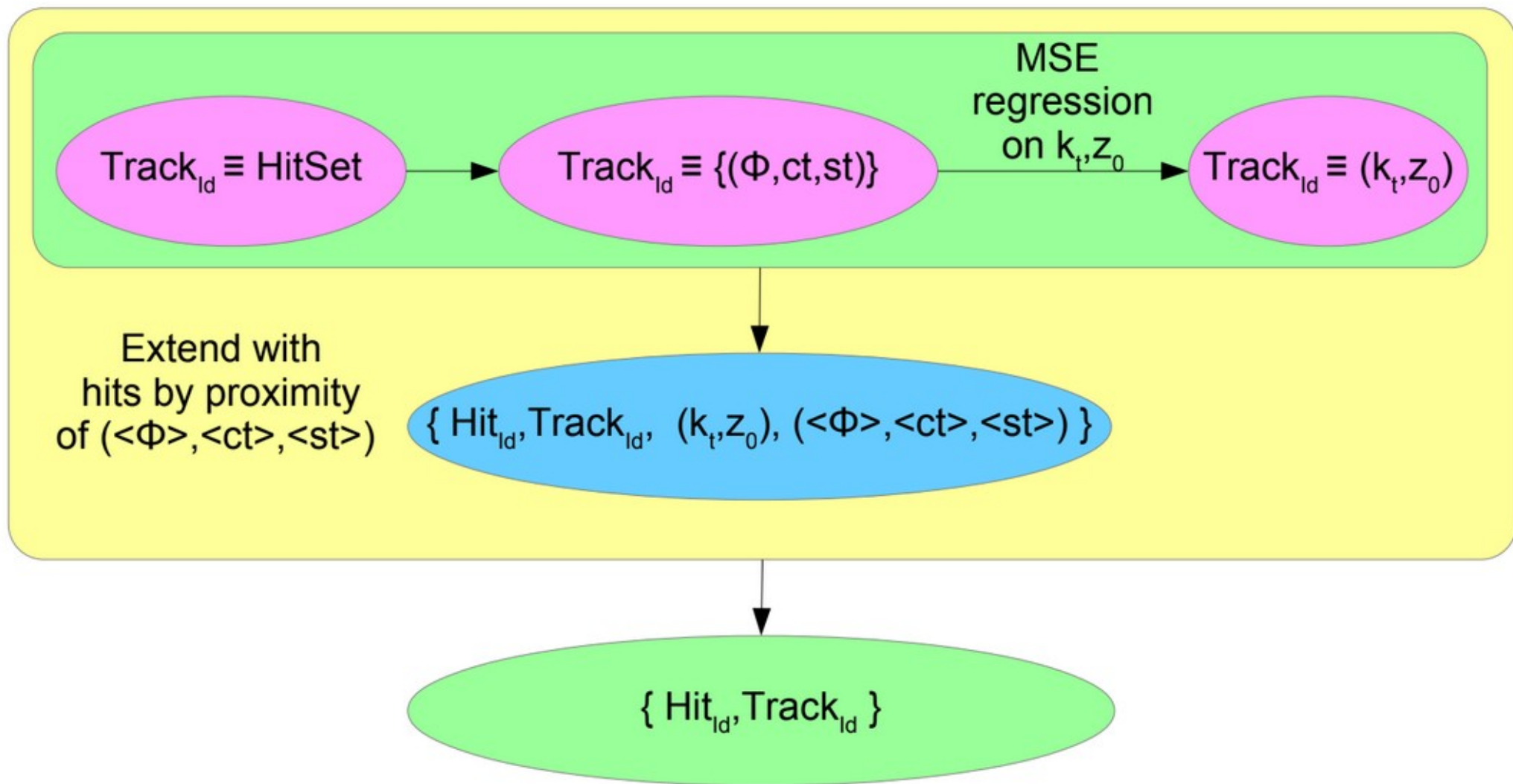


Yuval & Trian (#7)





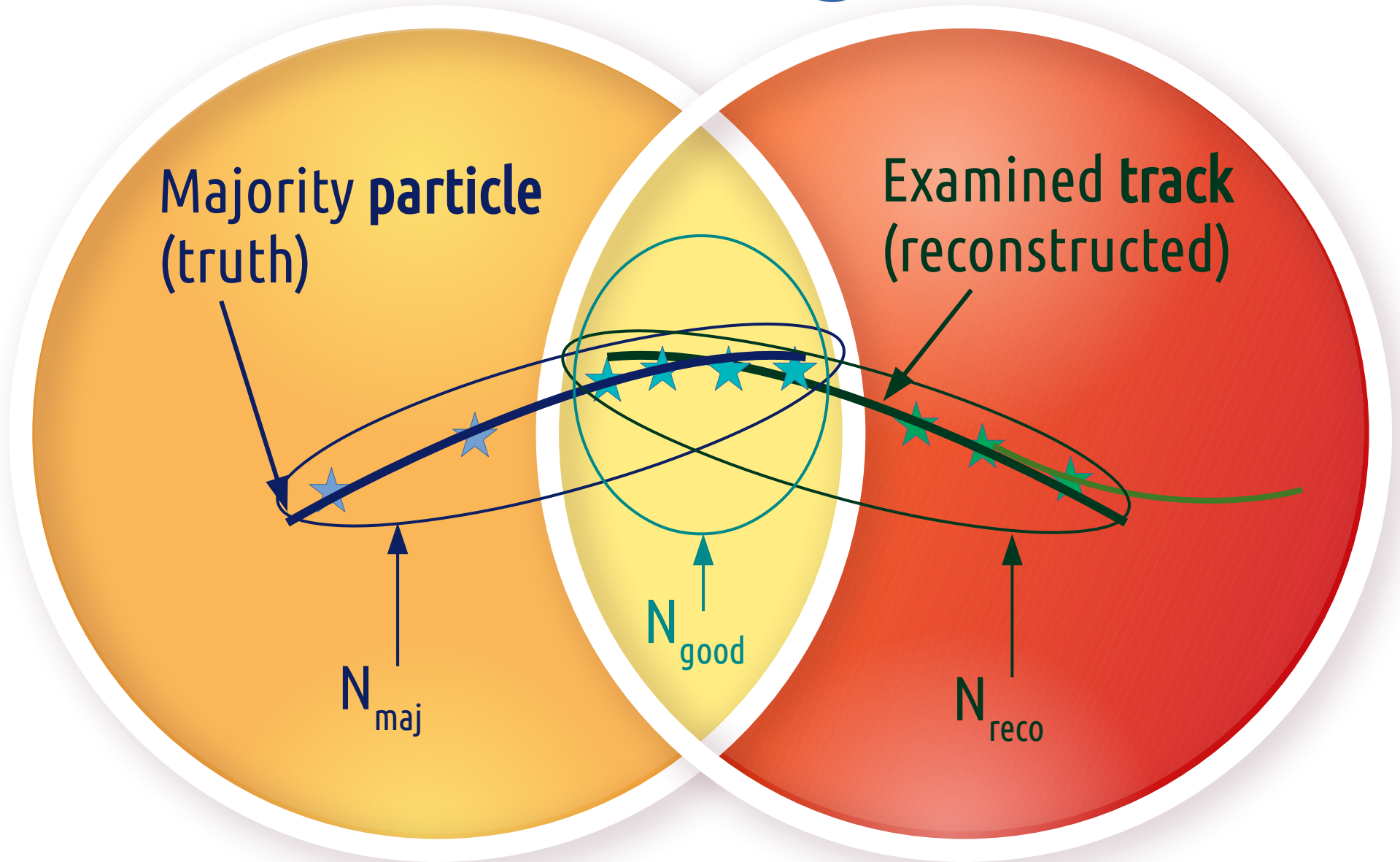




Other contestants

- #7 : Yuval & Trian
 - « Binned randomized Hough transform » for clustering
 - ML (LightGBM) to merge tracks
- # 9 : CPMP - « DBSCAN Forever »
 - DBSCAN on transformed space including deviation from helix
 - On each iteration clusters = new candidate tracks, merged
- #12 : The Finnies
 - DBSCAN variants → 5 hit track seeding
 - LSTM → estimate 10 hits
 - KNN for Track fitting

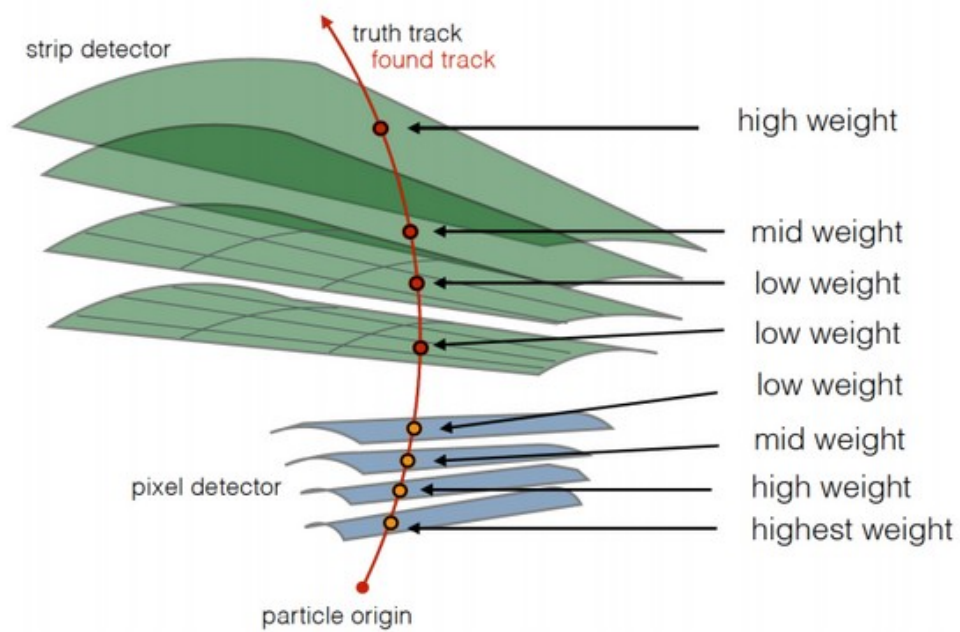
Scoring



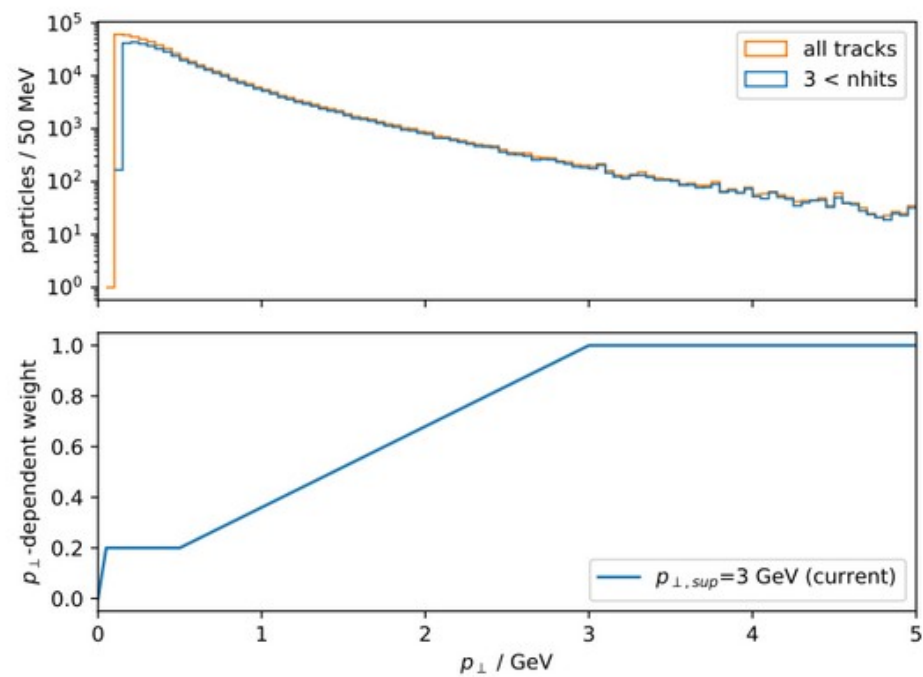
$$\text{Purity}_{maj} = N_{good} / N_{maj}$$

$$\text{Purity}_{reco} = N_{good} / N_{reco}$$

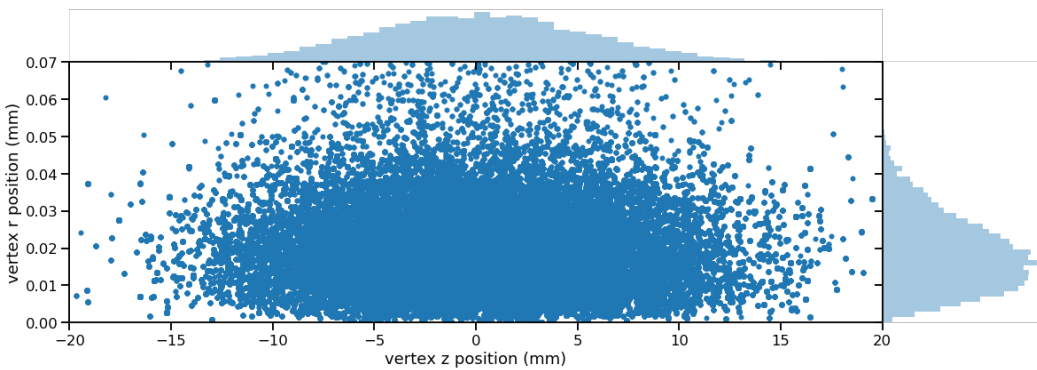
Hit order



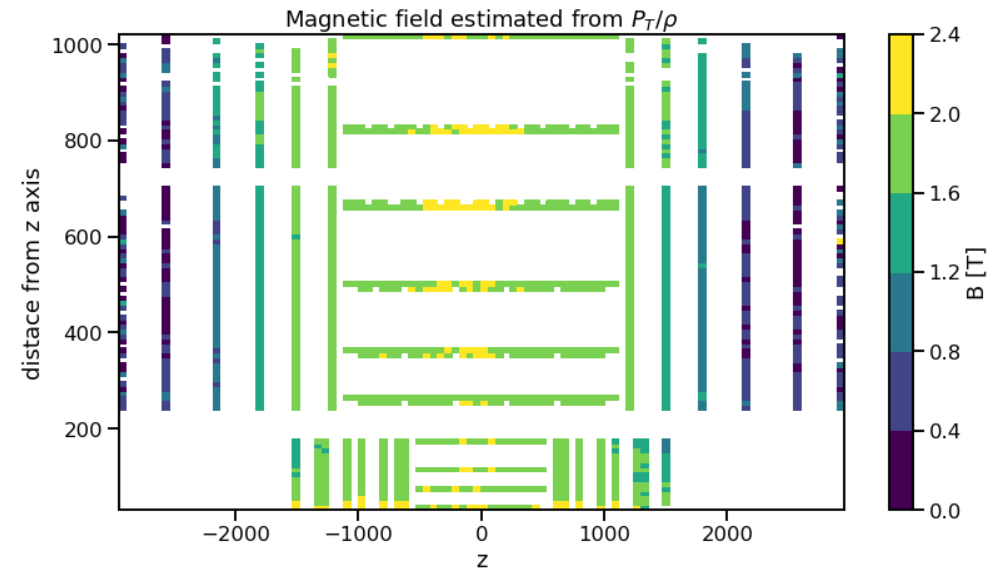
Particle p_{\perp}



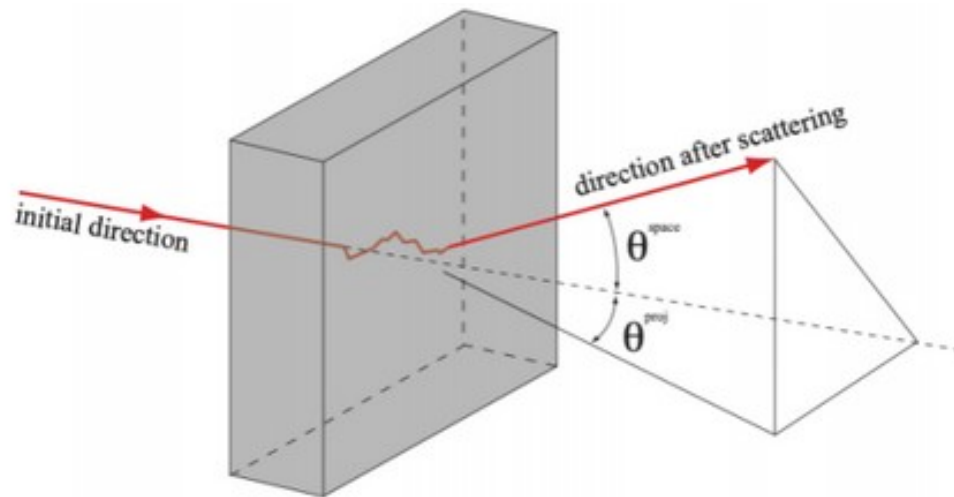
« passing through origin »



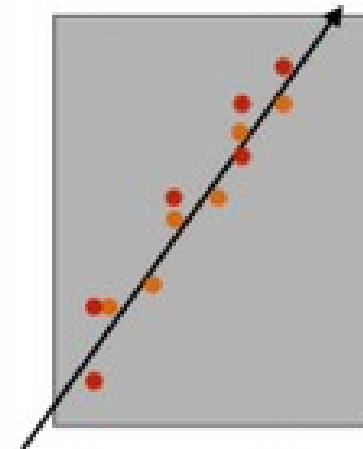
« homogeneous magnetic field »



Scattering in detector



Energy (hence momentum) loss



LHC / HL-LHC Plan

