

Randomized Computer Vision Approaches for Pattern Recognition in Timepix and Timepix3

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Aim of the talk



Give brief overview of Timepix pixel detectors.
Describe previous analysis works.

□ Introduce *new* analysis work:

- Extend current approaches to utilize new information,
- CV-inspired overlap detection and separation,
- Randomized regression of spatial trajectories in 2D and 3D,
- ML: particle species classification.

Introduction



Timepix detectors



Active pixel detector developed by Medipix collaboration, CERN
256 x 256 pixels with 55 µm pitch (1.98 cm² sensitive area)
Hybrid design:

- Sensor thin layer of semiconductive material (Si, GaAs, CdTe),
- ASIC signal pre-amplifiers and counting circuitry (CMOS-based).





Operation modes





Detector networks at LHC

TPX05



□ ATLAS

- Medipix (run 1),
- Timepix (run 2),
- Timepix3 (run 2)

□ MoEDAL, LHCb

- Timepix (run 2),
- Timepix3 (run 2)





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Notable applications at LHC



Determination of absolute luminosity and induced radioactivity

Ref: A. Sopczak *et al.*, "Precision Measurements of Induced Radioactivity and Absolute Luminosity Determination With TPX Detectors in LHC Proton–Proton Collisions at 13 TeV," in *IEEE Transactions on Nuclear Science*, vol. 65, no. 7, pp. 1371-1377, July 2018. <u>Characterization of mixed</u> radiation fields at different <u>positions</u>

Ref: Bergmann, B., *et al.* "ATLAS-TPX: a two-layer pixel detector setup for neutron detection and radiation field characterization." *Journal of Instrumentation* 11.10 (2016): P10002.



Notable applications in space

□ Proba-V satellite (2013)

- LEO, altitude: 820 km
- SATRAM = Timepix-based spacecraft radiation monitoring platform

□ VZLUSAT-1 nanosatellite (2017)

- LEO, altitude: 450 km
- Timepix-based X-ray telescope on board







Data examples





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Morphological clustering

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□ Idea: separate pixels into tracks by their spatial adjacency



Ref: Holy, T., et al. "Pattern recognition of tracks induced by individual quanta of ionizing radiation in Medipix2 silicon detector." *Nuclear Instruments and Methods in Physics Research Section A: Accelerators, Spectrometers, Detectors and Associated Equipment* 591.1 (2008): 287-290.

Morphological classification



Various thresholds on linearity, roundness, convex hull, etc.
Powerful & simple discrimination between track types.
Often combined with *a priori* info (about beam, converters, etc.)

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 Dots photons and electrons (10 keV)
Small Blobs photons and electrons
Curly Tracks electrons (MeV range)
Heavy Blobs heavy ionizing particles with low range
Heavy Tracks heavy ionizing particles (photons)
Straight Tracks energetic light charged particles (MIP)

Ref: Holy, T., et al. "Pattern recognition of tracks induced by individual quanta of ionizing radiation in Medipix2 silicon detector." *Nuclear Instruments and Methods in Physics Research Section A: Accelerators, Spectrometers, Detectors and Associated Equipment* 591.1 (2008): 287-290.

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Extension points



- Both techniques originally developed for Medipix2 (binary pixels)
 - ToT information not utilized
 - Susceptible to pile-up \rightarrow acquisition time tuning

□ 6 classes have proven insufficient for more complex tracks, e.g.:

δ-ray emissions,

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- collisions, fragmentation events,
- exotic particles





2D Analysis



CV-inspired ideas



□ Re-formulate task as an <u>object detection problem</u>

Treat ToT as "intensity" and frame as "image"

Photography:





CV-inspired ideas



Re-formulate task as an <u>object detection problem</u>

Treat ToT as "intensity" and frame as "image"

Combine data with a priori known information in order to achieve 3D results.

Photography:

Timepix:





Object class: heavy ions



Desirable properties:

- Large momentum \rightarrow assumption of linear trajectory
- Hidden parameters can be approximated with a line segment.
- Charge sharing produces continuous pattern \rightarrow suitable for LO.



Heavy ion tracks

Cross-section:













Detector frame:



MoEDAL, detector #4 2015-09-12



60000

40000

20000

Energy [keV]

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Segmented track + fit:



... Entry point



- $\boldsymbol{\theta}$... Incidence angle
- φ ... Azimuth

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 $-\frac{dE}{dx} \cong \frac{\sum_{i} E_{i}}{trajectory \ length}$

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Hough Transform: Overlap separation example



- □ <u>Question:</u> with Hough, are RANSAC & LO *really* necessary?
- Benchmarked performance, spatial and angular accuracy.
- Generated random frames from manually curated dataset of over 1K experimentally observed tracks.



Hough Transform: Overlap separation example



- □ <u>Question:</u> with Hough, are RANSAC & LO *really* necessary?
- Benchmarked performance, spatial and angular accuracy.
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Algorithm	n	$\langle FR \rangle$	$\langle E(x_{\mathrm{far},1}) \rangle$	$\langle E(x_{\mathrm{far},2}) \rangle$	$\langle E(\varphi) \rangle$	$\langle E(\theta) \rangle$	$\langle t \rangle$ [ms]
RANSAC	5	0.107	4.344	4.266	5.244	10.085	4,492.2
	10	0.253	6.413	6.384	9.030	10.826	9,123.3
	20	0.544	10.264	9.776	12.012	11.786	$20,\!626.7$
	50	0.772	14.852	14.572	14.655	10.226	38,161.3
LO-RANSAC	5	0.104	4.194	3.953	5.169	11.709	$24,\!105.5$
	10	0.242	6.096	6.219	8.549	12.250	$57,\!995.2$
	20	0.532	9.777	9.311	11.620	12.682	125,795.0
	50	0.770	14.034	13.888	13.960	10.167	$294,\!498.0$
SA-RANSAC	5	0.105	4.026	3.938	5.115	12.205	20,882.6
	10	0.238	6.088	6.064	7.833	12.539	19,708.4
	20	0.528	9.768	9.163	10.912	12.578	39,827.8
	50	0.769	14.324	14.111	12.535	9.892	83,365.4





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Particle Identification



Motivation



□ dE/dx includes information about particle species and momentum



Motivation



□ dE/dx includes information about particle species and momentum.



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Results (5 classes)





Configuration:

- k-NN classifier (k=7)
- 32 uniform dE/dx bins

Results:

• Accuracy = 0.89



Results (5 classes + rejection)

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Configuration:

- k-NN classifier (k=7)
- 32 uniform dE/dx bins
- Confidence >= 0.95

Results:

- Accuracy = 0.94
 - Rejection Ratio = 0.25

WIP: 3D Analysis



Timepix3 vs. Timepix



□ Simultaneous ToT and ToA operation mode

Data-driven readout

- Hits are reported <u>during</u> acquisition → need to work with incomplete information in online scheme
- Temporal consistency: hits may be out of order
- Large data throughput (theoretically up to 80 MHz)

□ Increased time resolution

- Fast ToA clock at 640 MHz \rightarrow resolution: 1.56 ns
- TDC synchronization \rightarrow coincidence analysis



Online morphological clustering



- □ Conventional morphological clustering *cannot* be used in Timepix3 \rightarrow The algorithm needs to be extended.
- □ Data-driven mode: no frames, pixels are reported continuously □ Out-of-order timestamps: hit buffering in a priority queue □ Geometric DS: efficient adjacency test



→ Advertisement: L. Meduna et al. Real-time Timepix3 data clustering, visualization and classification with a new Clusterer framework



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3D event reconstruction



\Box Estimation of relative depth from fToA \rightarrow 3D point clouds

Ref: Bergmann B. *et al*, "3D track reconstruction capability of a silicon hybrid active pixel detector" *The European Physical Journal* C 77.6 (2017): 421.

RANSAC and Hough methods easily generalize to 3D.
Results:



Conclusion



Previously developed pattern recognition methods updated for new detectors (Timepix, Timepix3):

- Utilized ToT, fToA (increased precision, online clustering alg., 3D reconstruction)
- Updated clustering for data-driven readout mode \rightarrow online algorithm

□ New methods for HETE analysis developed and evaluated:

- Randomized \rightarrow accurate results, good performance
- Trans-dimensionality \rightarrow well-suited for overlap separation (simulated up to 20)
- ML \rightarrow basic feature model has shown accuracy up to 93.5%

□ New methods tested in various applications:

- Angular, dE/dx analysis (MoEDAL)
- Particle identification (Heidelberg Ion-Beam Therapy Center)

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Further work



□ Analysis of non-linear trajectories (under assumption of continuity)

- Polynomial / spline interpolation
- Principal curves

Detection of interactions between events within detector.

- Branch detection
- Prong detection

□ Application of deep learning (2D CNNs).

\Box Bus saturation \rightarrow offload to readout hardware (FPGA).

Thank you for your attention. Questions?

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Literature:

P. Manek. Machine learning approach to ionizing particle recognition using hybrid active pixel detectors, Master Thesis, Faculty of Electrical Engineering, Czech Technical University

