Application of the Topological Track Reconstruction to an idealised water-based liquid scintillator detector

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In unsegmented large-volume liquid scintillator detectors

Outline

- Motivation
- Simulation
- Light Separation
- Topological Track Reconstruction
- Machine Learning in JUNO
- Summary

Motivation: Introduction

Motivation for this work

- Idealised detector ⇒ full potential of new and advanced techniques: Large Area Picosecond Photodetectors (LAPPDs) and Water-based liquid scintillator (WbLS)
- Reconstructions and light separation algorithms for this detector ⇒ experience and results for future experiments like Theia
- Theia: Planned long-baseline beam neutrino experiment with a broad physics program
- Aiming to use WbLS as well as LAPPDs and PMTs

For Theia see also: M. Askins et al. "Theia: An advanced optical neutrino detector" (arXiv:1911.03501)

Motivation: Water-based liquid scintillator

Cherenkov light

- Prompt and directional emission
- Directional information with ring location and shape
- Particle identification via ring structure

Scintillation light

- Delayed and isotropic emission
- Low threshold and good energy reconstruction
- Shower reconstruction

WbLS

- Separation of light types gives access to all advantages.
- Further particle identification with the ratio of Cherenkov to scintillation light



[arXiv:1607.01671]

Simulation: Introduction

- Geant4 simulation of a small detector (Radius: $\sim 1.5 \,\mathrm{m}$, Height: $\sim 3.8 \,\mathrm{m}$)
- => comparable to ANNIE
- => little scattering and attenuation for optical photons
 - Detector completely covered with LAPPDs
 - Volume is filled with water or WbLS.
 - LAPPD model taken from ANNIE simulation with minor adjustments
 - Optical properties of WbLS taken from the Theia simulation
 - Example event: $500 \,\mathrm{MeV}$ muon in x direction in water starting at detector center.



Simulation: WbLS examples plots

- 500 ${\rm MeV}$ muon in \times direction starting at detector center
- Resulting in \sim 3900 Cherenkov and \sim 28,000 scintillation hits.
- Little scattering due to small detector $(\sim 6\% \text{ of hits are scattered or reflected.})$
- Cherenkov disk due to exiting muon





Detected angle (all hits)

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Light Separation: Concept

Method for separating Cherenkov and scintillation light

- 0 Cut hits later than $100\,\mathrm{ns.}$
- Smear vertex with an accuracy possible for vertex reconstructions: 20 cm and 1 ns
- Reconstruct particle direction via directional sum.
- Use direction for Cherenkov angle acceptance in samples on the track.
- Calculate expected time and compare with hit time.
- Weight hits according to the timing spectrum.

Angle truth to directional sum vs. number of hits



Light Separation: First result



- 12.9 % of all hits are Cherenkov hits.
- $\bullet~71\,\%$ of remaining hits with a cut at 0.66 are Cherenkov.
- => 58 % Improvement
 - 33.5 % of all Cherenkov, 2 % of all scintillation photons remain.
 - Room for improvement in angle acceptance, sampling of particle track and weighting function.

General concept of the Topological Track Reconstruction (TTR)

- $\bullet\,$ Known reference point in time $t_{\rm ref}$ and space $r_{\rm ref}$
- Assume straight particle path with velocity c_0 .
- Calculate possible locations \mathbf{x} of the particle at time $t(\mathbf{x})$.
- Developed by our group in Hamburg. See: Björn Wonsak et al. "Topological track reconstruction in unsegmented, large-volume liquid scintillator detectors" (arXiv:1803.08802)



Probability density functions



- Develop **p**robability **d**ensity **f**unctions (**p.d.f.**s) when taking more effects into account.
 - **(**) Isochrones come from the inversion of t(x).
 - Time uncertainty of scintillation light and response of photosensor
 - Oetection and propagation effects like angular acceptance and attenuation



Reconstruction method



- Create a p.d.f. for every hit and each PMT.
- Superimpose the p.d.f.s for every bin in volume.
- Gain probability mask showing most-likely origin of light.
- Treat prior iteration as truth; cut cells.
- Reconstruct again based on the previous iteration.
- Refine binning for more detailed result.



TTR: Two reconstructions in one

- Modify the reconstruction to work with pure Cherenkov detectors.
- Reconstruct pure Cherenkov event for proof of principle.



- Modify the reconstruction to have a scintillation and a Cherenkov part.
- With every iteration two reconstructions are performed: One considers hits to be Cherenkov photons, the other assumes hits to be Scintillation photons.
- Next slides: Perfect ordering from MC truth, Cherenkov reconstruction reconstructs only Cherenkov hits and vice versa.

Reconstruction result WbLS: Iteration 0



Reconstruction result WbLS: Iteration 5



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TTR: Status

Status

- Splitting of the reconstruction worked well; results for the raw reconstruction look like expected.
- Raw reconstruction: no usage of propagation effects except angular acceptance of photodetectors
- Characteristics of both light types can be distinguished.
- Probability mask too confined on middle of the track

Issue

- Results for emitted light look unexpected; probability mask has high entries at edge, artefacts.
- Emitted light algorithm: use all propagation effects and the Look-Up-Tables (LUTs) to calculate the number of emitted photons for each cell.
- Worked well for larger detectors like LENA, problematic with this small WbLS detector for particles leaving detector.
- Currently under investigation

TTR problem: Emitted light at iteration 5



Reminder: Directional information in pure water



- For each cell take hits giving a contribution to cell content into account.
- Project hits on unit sphere around cell.
- Gain directional information via circular Hough transform and directional sum.
- Needs adjustments and improvements in the new version of the TTR.

Machine Learning I: Direction reconstruction in JUNO

Work of Hauke Schmidt

- $\bullet\,$ JUNO has $\sim 3\,\%$ Cherenkov light \Rightarrow direction reconstruction
- Motivation: background suppression for sources with known location, especially solar neutrinos
- Assumption of known vertex
- => Time of flight correction
 - Projection of hit coordinates on unit sphere around the vertex
- => Angular coordinate for every hit
 - Modification of distance of hit point to origin according to time information
- => Time deformed sphere around vertex
 - Network PointNet (see arXiv:1612.00593)
 - Data represented as PointCloud (PMT positions, hit times)



Results



- $\bullet\,$ No dependence on position: energy of $8\,{\rm MeV}$ gives with vertex uncertainty and TTS error of 90 $^\circ.$
- => Direction can be confined to half of the detector at $8 \,\mathrm{MeV}$.
- => Great result for amount of Cherenkov light in JUNO
- => Expected to be very useful in a WbLS-detector.

Machine learning II: introduction and simulated data

Work of Rosmarie Wirth

Goal

• Reconstruct track and find shower with Machine Learning.

Simulation

- Toy MC simulating scintillation along random track with a high emission point (Peak)
- $\bullet\,$ Cubic detector with $4\,\mathrm{m}$ edge length
- $\bullet~100$ PMTs with $1\,\mathrm{ns}$ time resolution per wall
- First stage: Dynamic Graph CNN (Yue Wang et al. "Dynamic Graph CNN for Learning on Point Clouds" arXiv:1801.07829)
- Second stage: fully connected layers

Output goals

- Coordinate reconstruction (start, peak, end)
- Voxel reconstruction





Simulated Data

Goal I: Coordinate reconstruction

Promising results:

Position	Mean distance
Start	$0.16\pm0.20\mathrm{m}$
Peak	$0.22\pm0.14\mathrm{m}$
Exit	$0.21\pm0.11\mathrm{m}$

 \Rightarrow Track and shower identification work.

Outlook

- Implementation for OSIRIS sub-detector of JUNO
- More realistic data
- Use case: Muon flux and veto validation



mean distance between label and predicted coordinate

Goal II: Voxel reconstruction

- (99.24 ± 2.84) % of the path voxels discriminated to 7 % of detector volume (empty means ≤ 200 photons)
- Mean distance reconstructed peak to label $(0.36\pm0.47)\,\mathrm{m}$
- For 85 % of the analysed peaks, distance reconstructed peak to label $\leq 0.5\,\mathrm{m}$
- Reconstructed photon emission distribution is not reliable: number of photons in empty voxels too high and vice versa



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Summary

Summary

- Simulation in place for simulating events and LUTs
- Light separation algorithm shows first results.
- TTR works well with WbLS in raw reconstruction.
- Directional information can be extracted within the TTR.
- $\bullet\,$ ML: Direction reconstruction can confine direction to half of the detector for 8 ${\rm MeV}$ electron events in JUNO.
- ML II: Coordinate and voxel reconstruction are working well with toy MC data.
- Additional work for LAPPD simulation in ANNIE and other service work (not shown here)

Outlook

Outlook

- Improve light separation algorithms and study different energies and particles.
- Fix problems with emitted light algorithm in TTR.
- Use TTR for separation.
- Adjust directional information algorithms to new TTR version.
- Use advantages of light separation.
- Contract is extended for 6 months.

Thank you for your attention.

Light Separation I: Wavelength

- Cherenkov photons in blue/green wavelengths
- $\bullet\,$ Scintillation photons in UV/blue wavelengths
- => Use wavelength filtering for separation.







Detected Wavelength

Dichroicon

- Dichroic filter: Reflection/Transmission dependent on wavelength
- Two photosensor with Winston cones: One for Cherenkov wavelengths, one for scinitllation wavelengths

Light Separation II: Position

- Cherenkov photons emitted at a characteristic angle of \sim 38 40 $^{\circ}.$
- Scintillation photons emitted isotropic.
- => Use spatial information for separation.
 - In an experiment excellent granularity in direction of Cherenkov cone is needed.
- => LAPPDs are capable of this feature.
 - For non-beam events: Track reconstruction is necessary to get direction.



Light Separation III: Timing

- Cherenkov photons emitted prompt.
- Scintillation photons emitted delayed.
- Velocity dependent on wavelength: Scintillation light travels slower
- => Use time information for separation.
 - Need for fast photosensors
- => LAPPDs are fast enough.
- \Rightarrow LAPPDs are a good tool for Cherenkov and scintillation light separation using spatial and timing information!



Hit time (all hits)

Light Separation IV: outlook

Two main ideas:

- Use algorithms before or in reconstruction to sort hits.
- **②** Use probability information within the reconstruction.

Algorithm way

- Generate several hundreds of events for looking into overall timing/spatial profile.
- Reconstruct track direction for using spatial information or use the topological track reconstruction for this purpose.
- Combination of timing and spatial cuts/weights for separation.

Separation within reconstruction

- Scintillation hits get less weight in Cherenkov reconstruction and vice versa at track.
- This information might be usable for separation.
- Furthermore, directional information like directional sum to estimate amount of Cherenkov versus scintillation light in cells.

Directional information

Goal

- Use the working principle of the TTR to gain directional information
- Find for every cell in volume direction vectors for secondary/shower identification
- Direction vectors might be usable for Cherenkov light identification

Method

- Project all contributing PMTs to cell on unit sphere.
- Two Methods: Circular Hough transform and directional sum



Directional sum and Hough transform

Directional Sum

- Add up the unit vectors of contributing PMTs.
- Gives two information:
 - Direction of the directional sum
 - Length of the directional sum
- Both information can be useful.



Hough transform

- Project unit sphere on angle plane.
- Draw circle around PMT position with radius corresponding to Cherenkov angle.
- Full circles useful for leaving tracks



[https://www.mathworks.com]



Directional Sum and MC truth in Hough transform

Preliminary results



- First information along the track with both methods
- Agreement with the event
- Directional sum needs improvement for end of track.

Preliminary results II and status

- Not only direction, also "quality" of directional sum and Hough transform useful
- Parameters: Length of directional sum and number of circles overlapping for Hough transform
- Indicate amount of Cherenkov light
- => Useful for wbLS application



Status: Directional information can be found and look promising. Needs testing and validation.

Motivation III: Large Area Picosecond Photodetectors (LAPPDs)

- Photodetectors with two Microchannel Plates and 30 anode strips with $20~{\rm cm}\cdot 20~{\rm cm}$ size
- Excellent time resolution of $<100\,\mathrm{ps}$ JUNO PMTs have $\sim 1\,\mathrm{ns}$
- Spatial resolution of $< 1 {\rm cm}$ JUNO PMTs no granularity: $> 50 {\rm \, cm}$
- Quantum efficiency $> 20\,\%$ JUNO PMTs $\sim 30\,\%$





Simulation: wbLS examples plots II



Time of emission (tof corrected)

Direction reconstruction

- $\bullet\,$ In JUNO $\sim 3\,\%$ of the emitted light is Cherenkov light.
- This opens an opportunity for a direction reconstruction.
- Motivation: background suppression for neutrinos of sources with known location, especially solar neutrinos
- No usage of the TTR
- Training sample: 100,000 electron events $(3 \, {\rm MeV})$ at the center of JUNO (using detector simulation)
- Validation and evaluation: 10,400 events for every $1\,{\rm Mev}$ step between 1 and $8\,{\rm MeV}$
- Time of flight correction
- $\bullet\,$ Time cut $5.5\,\mathrm{ns}$ after time of flight correction
- Usage of known vertex position



[Determination of Supernovae Direction with Reconstructed Positron Information; DOI: 10.22323/1.244.0067]

Hit distribution



- Not using a CNN because
 - Edge effects if trying to parameterize the 3D sphere to 2D cartesian coordinates
 - 3D CNN would contain lots of entries with a zero

 \Rightarrow massive amount of memory and running time





Results

- Network PointNet (see arXiv:1612.00593); Framework: **TensorFlow**
- Data represented as PointCloud (PMT positions, hit times)
- Implementation based on **Dynamic Graph** CNN with modifications:
 - No rotation and moving of input
 - Reduce output to three values
 - Add quadratic normalisation to output
 - Cosine function as loss function
 - Use Convolution, MaxPooling and Dense layer instead of ReduceMax and fully connected layer



Network architecture

- First stage: Dynamic Graph CNN (see arXiv:1801.07829)
- Second stage: Fully connected layers
- Notable feature: Node max pooling (take maximum of each node)



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Reconstruction result wbLS: Iteration 10



Hits across the parallel coordinate I



Hits across the parallel coordinate II







Strip 15 right side



Hits across transverse coordinate



/oltage [m//]

- 12

25 Time [ns]

Hits across transverse coordinate II





Right side of the LAPPD





Hits with different time at same spot





Right side of the LAPPD



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Hits with different time at same spot II





Right side of the LAPPD

Strip 15 right side



Diagonal Hits across LAPPD



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Influence of training and time resolution



- Training in center yields worse results, especially farther outside.
- Training on z-axis normalised to the time of flight is best option.
- Performance improves with time resolution of the PMTs.