



Searching for Dark Tridents with Convolutional Neural Networks

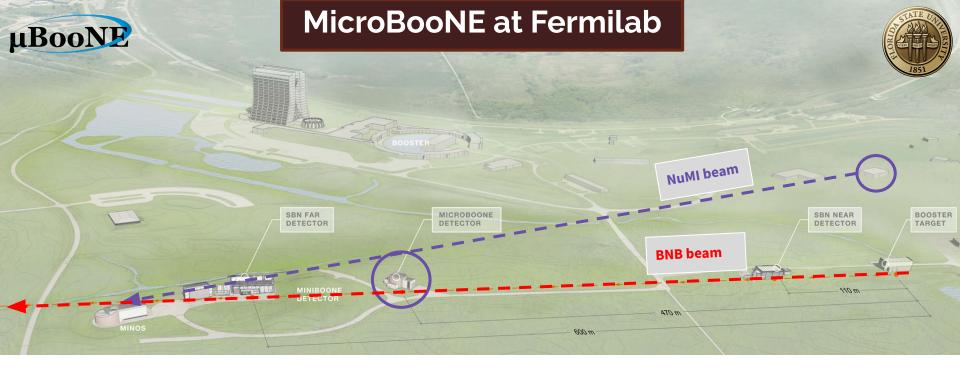
Luis Mora-Lepin on behalf of the MicroBooNE collaboration Neutrino Physics and Machine Learning Conference, ETH Zürich 25/06/2024







- The MicroBooNE detector and NuMI
- Overview of dark trident search
- A CNN for signal and background classification
- Results



MicroBooNE:

- Liquid argon time projection chamber (LArTPC)
- Active mass 85 tonnes
- Dimensions: 10.36 x 2.56 x 2.32 m³
- At surface level

Rich physics program:

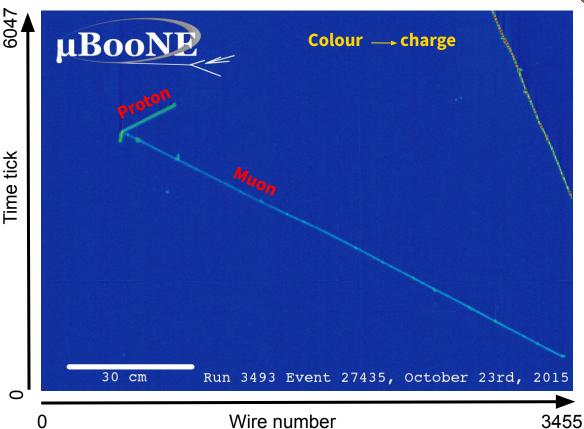
- Neutrino physics (Oscillations, cross section)
- BSM physics (This talk)
- LArTPC R&D



MicroBooNE Data



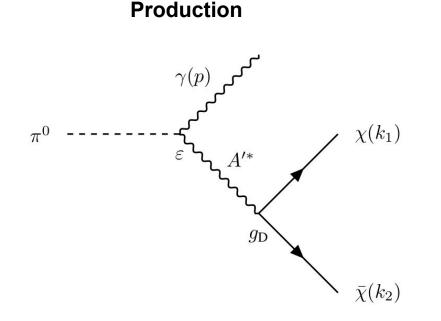
- MicroBooNE has three wire planes. Each plane produces a 2D view of the charged particles interacting with the detector volume
 - Spatial resolution of 3 mm per pixel
- Good calorimetric capacity
- The image shown here corresponds to the collection plane



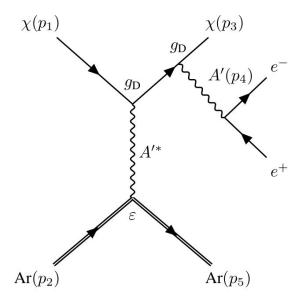
µBooNP Dark Tridents: A Dark Sector Portal



- DM candidate can be produced at fixed-target facilities through neutral meson decays
- Off-axis search of DM scattering has been proposed in: arXiv:1809.06388
- Interaction channel: DM scattering with the emission of an on-shell dark photon

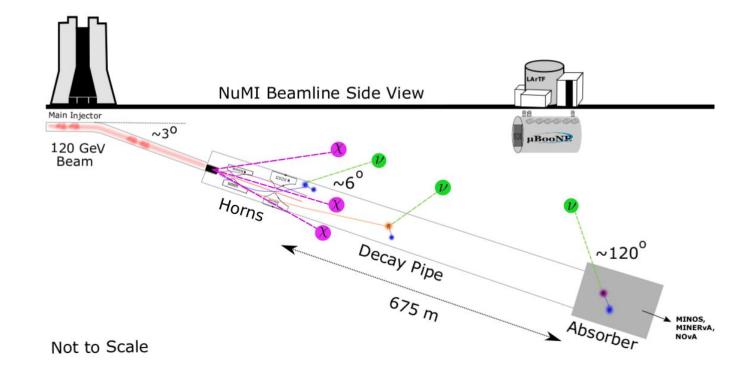


Scattering with Ar

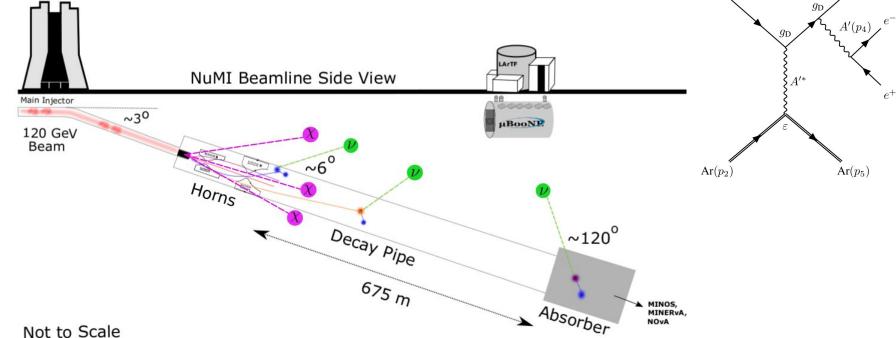












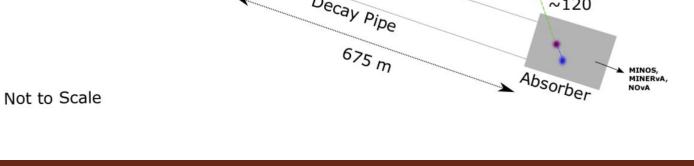


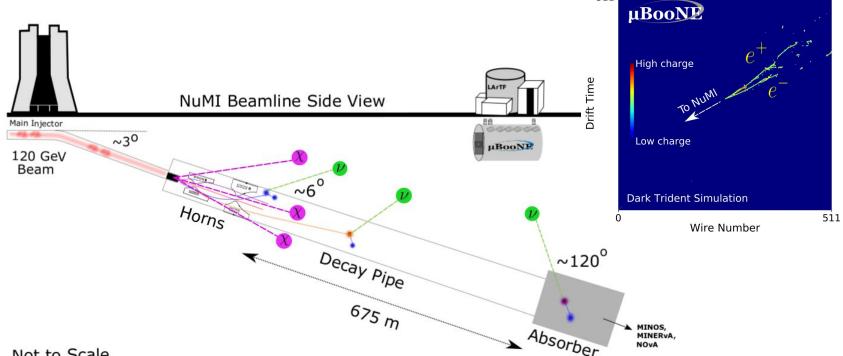


 $\chi(p_3)$

 $\chi(p_1)$

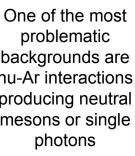
μBooNE

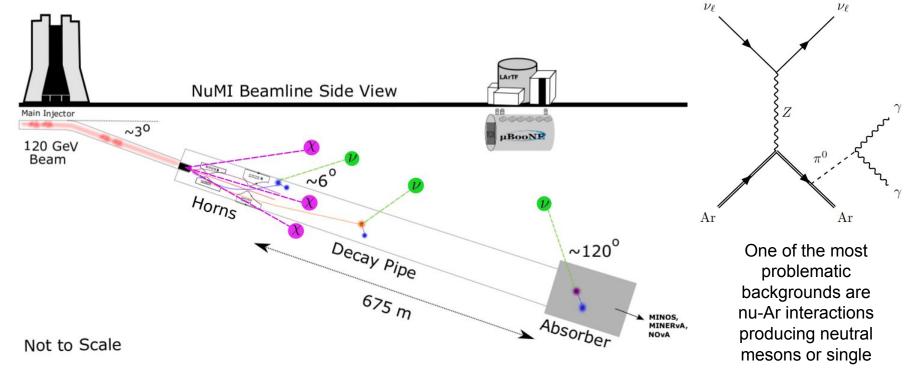




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μBooNE



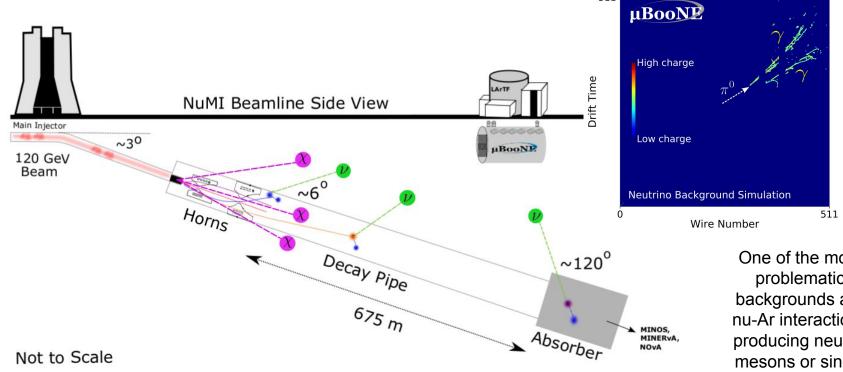






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One of the most problematic backgrounds are nu-Ar interactions producing neutral mesons or single photons

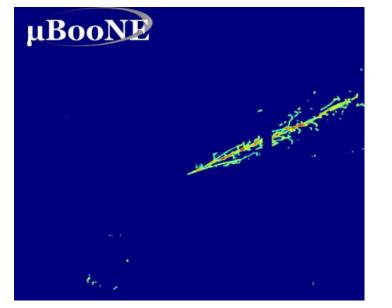


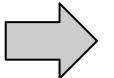












Signal or background?

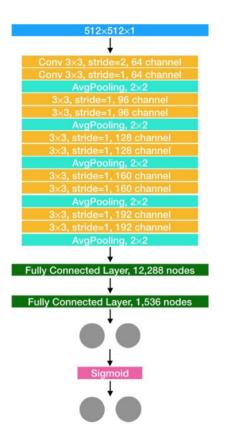
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A CNN for Background Rejection

- For this analysis we took advantage of the existing architecture developed for the Multi-Particle Identification Network (MPID)
- The CNN input are 512x512 MicroBooNE images cropped around the interaction vertex
- MPID's has filters with a size comparable to the activity expected on showers and tracks
- The final layer has been configured to output the probability of having either dark trident signal or background interactions

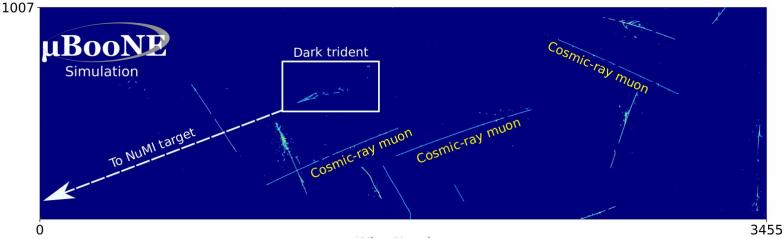
MPID details: Phys. Rev. D 103, 092003





Training Set Preparation





Wire Number

- no imagos aro sto
- The images are stored in a user-friendly format using the LArCV package
- A dedicated training/test set containing a benchmark signal sample and neutrino NCpi0 interactions was created. Cosmic rays tracks were also included

 Images are compressed by a factor of 6 on the time tick axis

μBool

Time Ticks

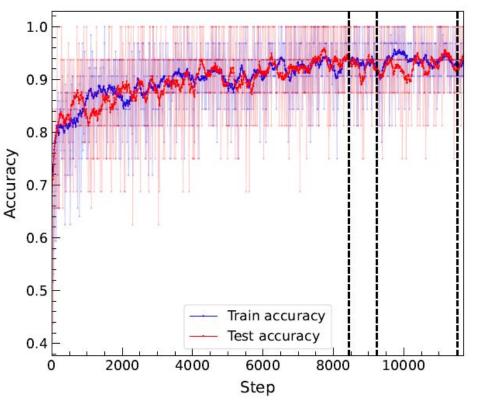
- Cropping around of the interaction vertex obtaining an image of 512x512 pixels
- For training true vertex is used. In contrast for the actual analysis we use reco vertex (provided by pandora)



Network Training



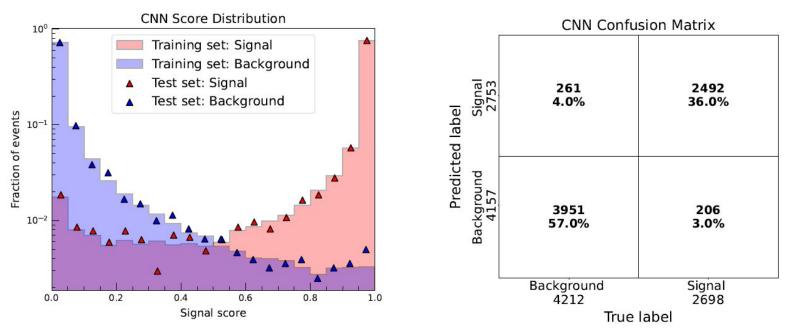
- Training set size: 62879
- Hardware: NVIDIA V100
- Iterations: 11786 (~5 epochs)
- Time: ~4 hours
- Batch size: 32
- Learning rate: 0.001
- Adam optimizer
- Dropout layers and L2 regularization were implemented to control the overfitting
- Binary cross entropy loss and accuracy were used to monitor the training





Training Results



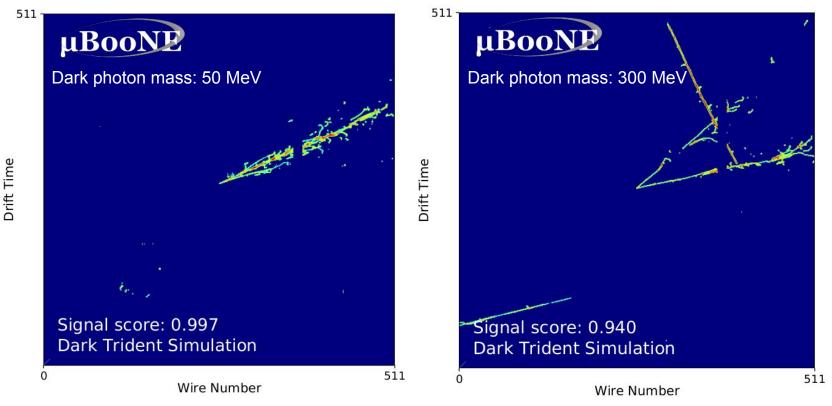


- The CNN achieves good separation of signal and background
- It reaches ~93% of accuracy
- Good generalization to signal samples simulated with different masses
- No signs of overfitting



Training Results





Occlusion analysis

µBooNE

Signal score: 0.997

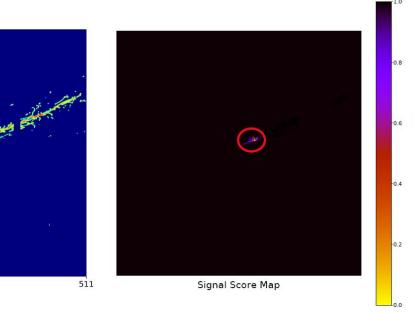
Dark Trident Simulation

Wire Number

511

Drift Time

- In this test a zone of nxn pixels in the image is 'hidden' before passing it to the CNN
- The CNN score will vary if important pixels of the image are occluded
- Pixels at the beginning of the showers contribute with important features



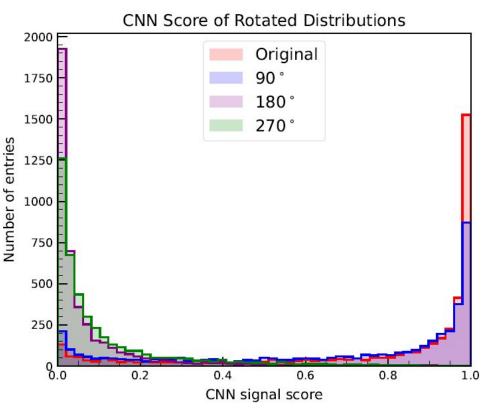




Angular Dependance



- The electron-positron shower direction typically point back to the NuMI target
- On the other hand, less neutrinos are expected from this direction as a result of the focusing horns
- We study if the CNN is able to infer the typical dark trident direction respect to NuMI
- Note: The MPID network was trained with isotropic angular distributions

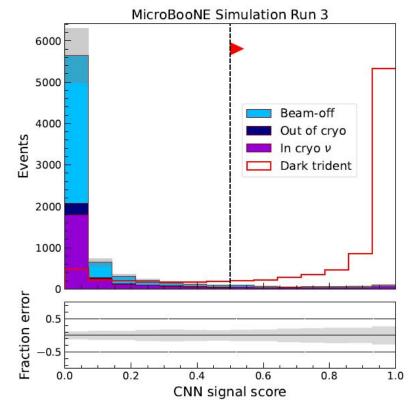




Inference Over Analysis Datasets



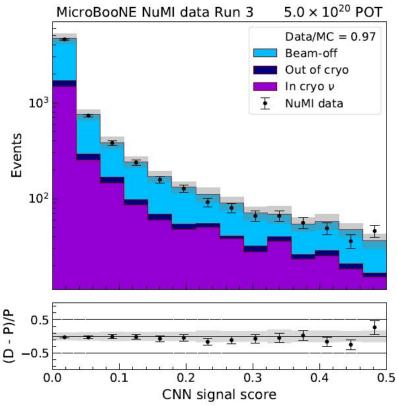
- The background prediction is composed of cosmic-ray (beam-off) interactions, and neutrino interactions produced inside and outside the cryostat
- The CNN also generalizes to events coming from these three background samples





Inference Over Analysis Datasets

- The background prediction is composed of cosmic-ray (beam-off) interactions, and neutrino interactions produced inside and outside the cryostat
- The CNN also generalizes to events coming from these three background samples
- We also checked the performance over data collected using the NuMI beam over a control region (CNN score < 0.5)



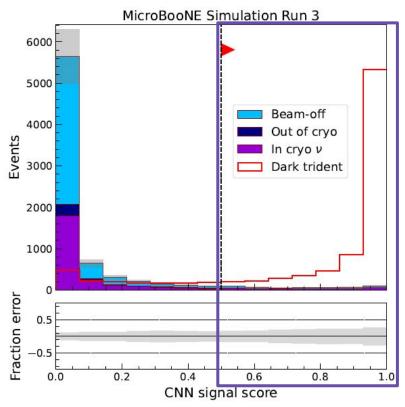






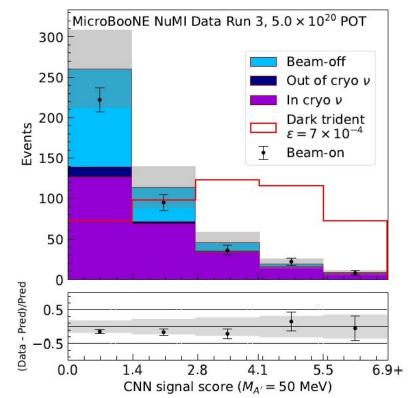


- The CNN score region above 0.5 is used to probe the dark trident hypothesis
- The scores are passed through a logit transformation which maps the interval (0.5,1.0) to (0, infinity)





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- The number of candidates found in the NuMI data is consistent with the background expectation



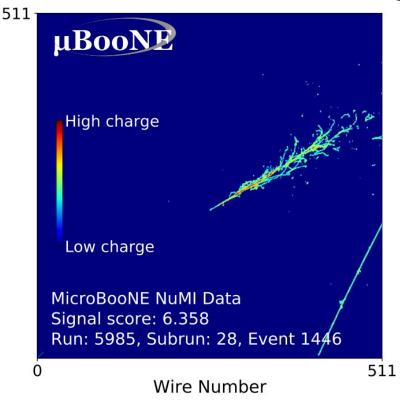


Results

Drift Time



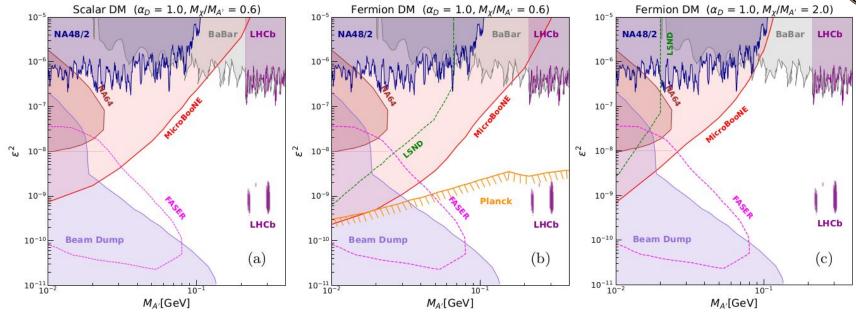
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MicroBooNE Limits





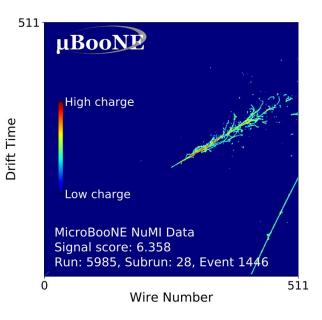
- The results obtained were used to set constraints to different combinations of the parameter space using the CLs method
- The limits obtained by MicroBooNE are the most stringent ones for dark photon masses below 100 MeV



Conclusions

- We have successfully implemented a CNN to discriminate dark trident interactions from neutrino interactions and cosmic rays
- The CNN was used to search for dark trident candidates in dataset collected using the NuMI beam
- A few candidates were found but the number is consistent with background expectation. New constraints on the model parameter space were obtained
- This technique can be generalized to other BSM models and LArTPC experiments
- Results published by PRL: <u>Phys.Rev.Lett.</u> 132 (2024) 24, 241801









Backup

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Training and Test Set



| Type of event | Number of events in training set | Number of events in test set |
|----------------------------|----------------------------------|------------------------------|
| Dark trident | 12601 | 1399 |
| Dark trident + Cosmic rays | 11776 | 1299 |
| ${ m NC}\pi^0$ | 12601 | 1399 |
| $NC\pi^0$ + Cosmic rays | 12601 | 1415 |
| Cosmic rays only | 13300 | 1399 |



The MicroBooNE Detector







MicroBooNE Dataset



