

# Progress Update

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UofT Neutrino/DM Meeting  
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# Current Status of Aluminum Sourcing

Company	Aluminum Type	MOQ	Price/Piece (USD)	Mould Price (USD)	Add. Fees (USD)	Min Price (USD)	OD Tolerance	WT Tolerance	Shipping notes
Modern International Trade Co.	6061	10	410			4100	+2.40/-0.80 mm	+/- 10%	FOB Tianjin
Foshan Kaiya Aluminum Co.	6063-T5	1	114.56	2132	517.29 <sup>1</sup>	2763.85	+/- 2 mm		CIF Toronto
Ningbo City Beilun Fayi Metal Product Co.	6061	10	471.4			4714	+/- 2 mm		CIF Toronto <sup>2</sup>

1. Additional courier fee for small order, refunded when placing full order. For initial order, 45 day production time because of new mould. For full order, 15 day production time

2. Shipping was changed from FOB Shanghai to CIF Toronto: added 239.90 USD per piece. For 10 pieces, 20 day production time. For full order, 50 day production time

# $\nu_e$ CC1 $\pi^+$ Status

$$\nu_e + p/n \rightarrow e^- + \pi^+ + p/n$$

- Moved fully to T2K MC – no longer using atm MC
- Discrepancies between my 1Re/1Re1de selections and those in TN319 due to lack of BANFF tuning in my selection
  - Confirmed by Xiaoyue, who was kind enough to re-run her code without BANFF tuning
  - BANFF tuning not used in Mark Hartz's T2HKK studies, so I'll just try to implement it for T2K
- Just now beginning to re-examine cut optimization for the 2Re $\pi$  and 2Re $\pi$ 1de selections
  - Evis and wall
  - Distance between sub-events for 2Re $\pi$ 1de selection
  - $\pi^0$  rejection?
  - $\mu$  rejection?

# MLHEP2017

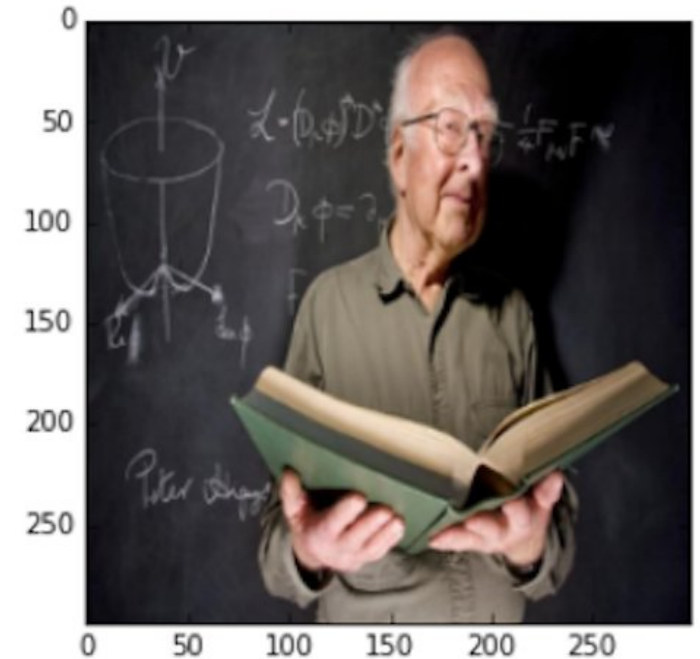
- Learned about machine learning and some of its possible applications to HEP
- If interested, slides and some python notebooks can be found here:
  - <https://github.com/yandexdataschool/mlhep2017>
- Topics covered: ML theory, overfitting, figures of merit, regularization, decision trees (bagging, RandomForest, boosting), neural networks (CNNs, RNNs), unsupervised learning, generative models

# Some thoughts

- Use Keras rather than directly using TensorFlow?
- In general, CNNs seem to be best at image classification
- Theo's code currently pre-processes rings to get them to be uniform
  - Have any studies been done on how CNNs handle raw event display data?
  - CNNs excel at extracting features, but currently the images have very few features for the CNN to extract
    - Perhaps look at using deeper CNNs on raw event display data?
    - Risk of overtraining – many methods available to combat this
- Looking forward to jumping into studying multi-ring events

# Some ML Fun

- Play around with a neural network:
  - <http://playground.tensorflow.org>



Peter Higgs is 36% cucumber

top-10 classes are:  
0.3624 cucumber  
0.1434 ear  
0.0900 zucchini  
0.0711 corn  
0.0242 tench  
0.0128 toilet seat  
0.0107 book jacket  
0.0103 barracouta  
0.0099 artichoke