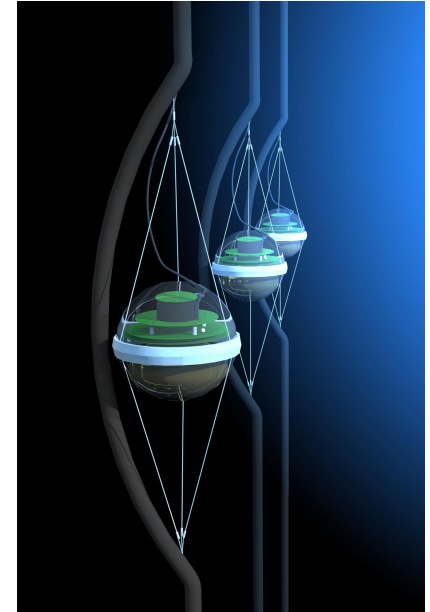


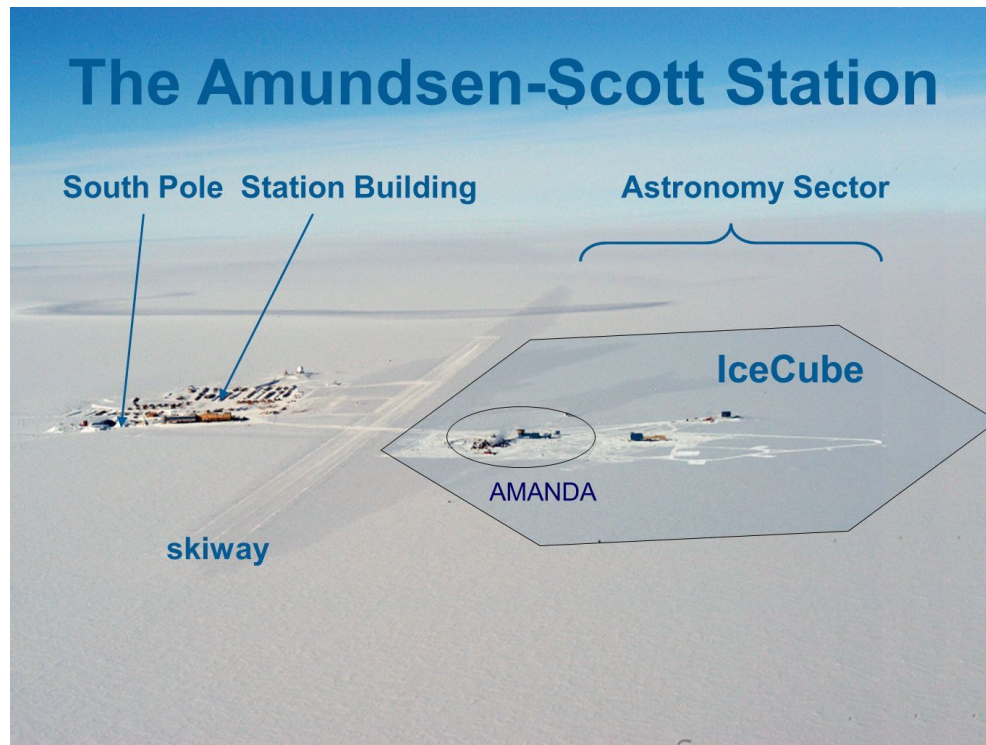
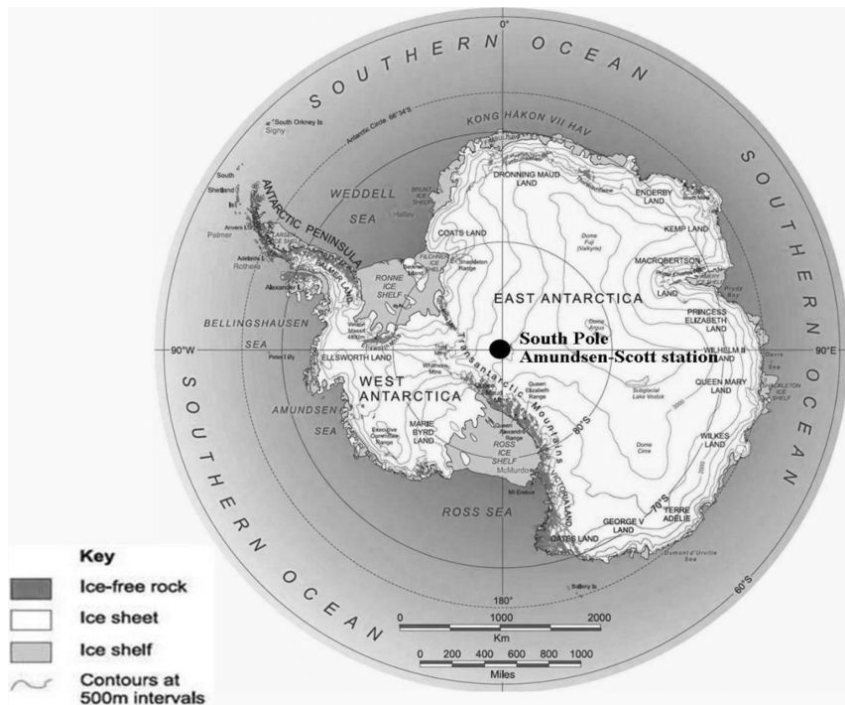
IceCube

IceCube

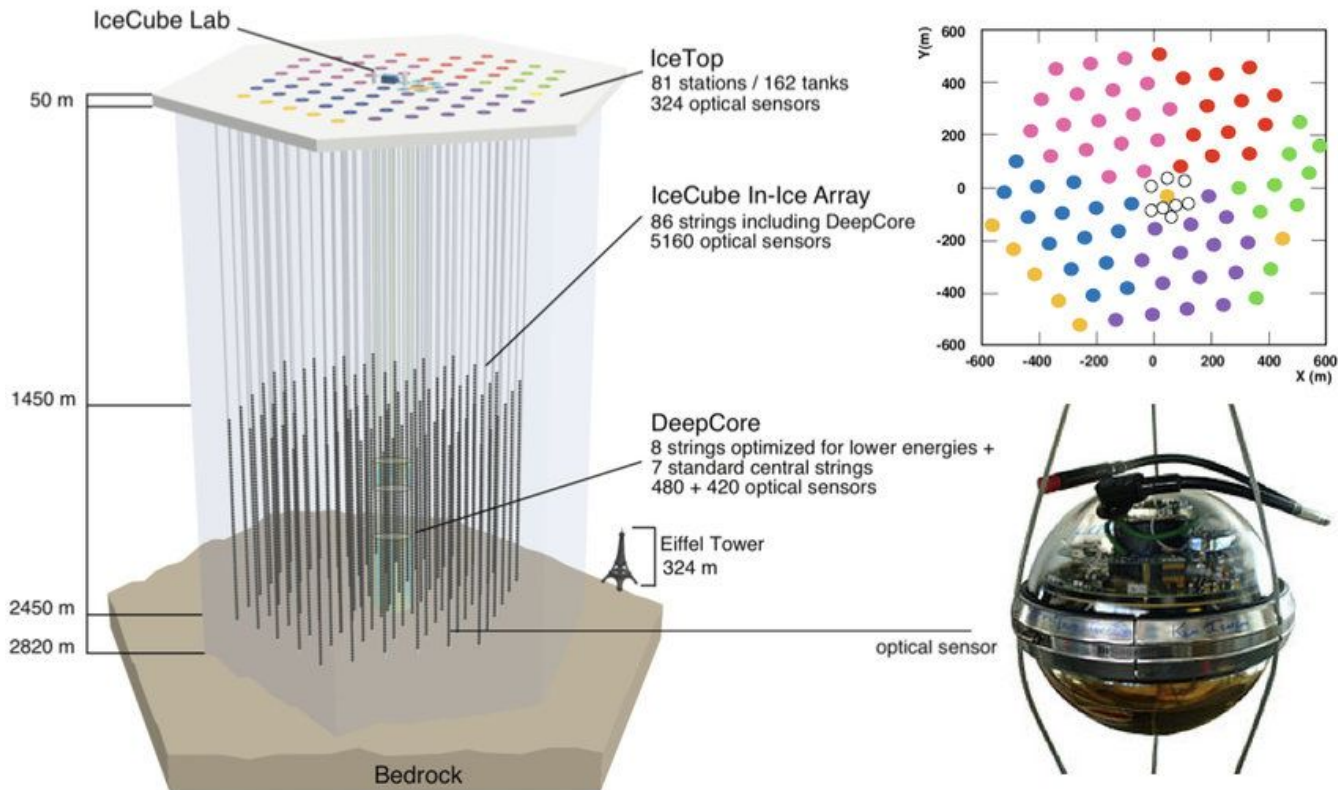
-

deep learning developments

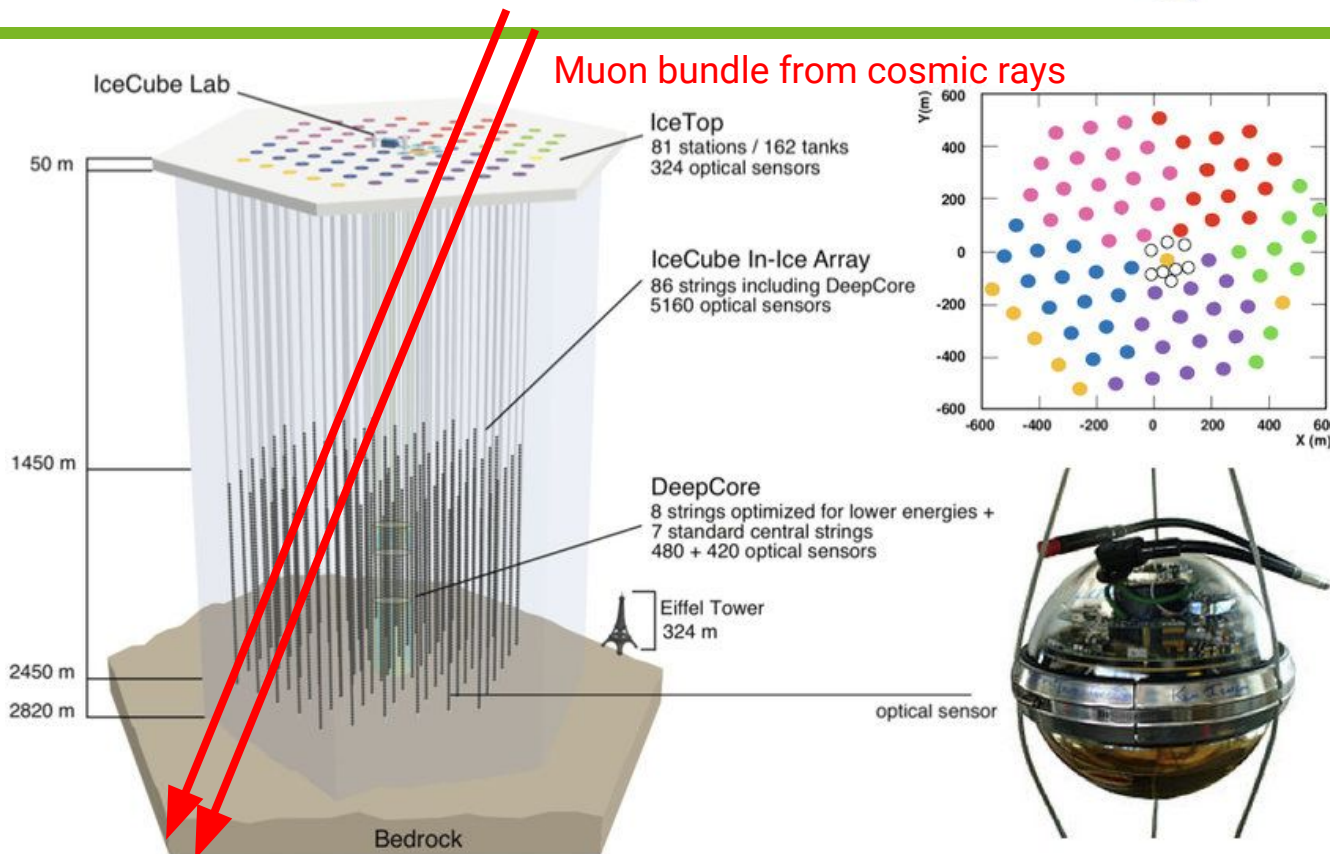




IceCube - the detector

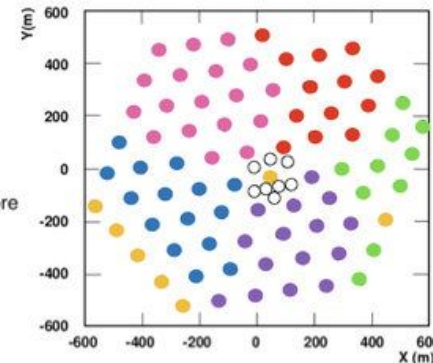
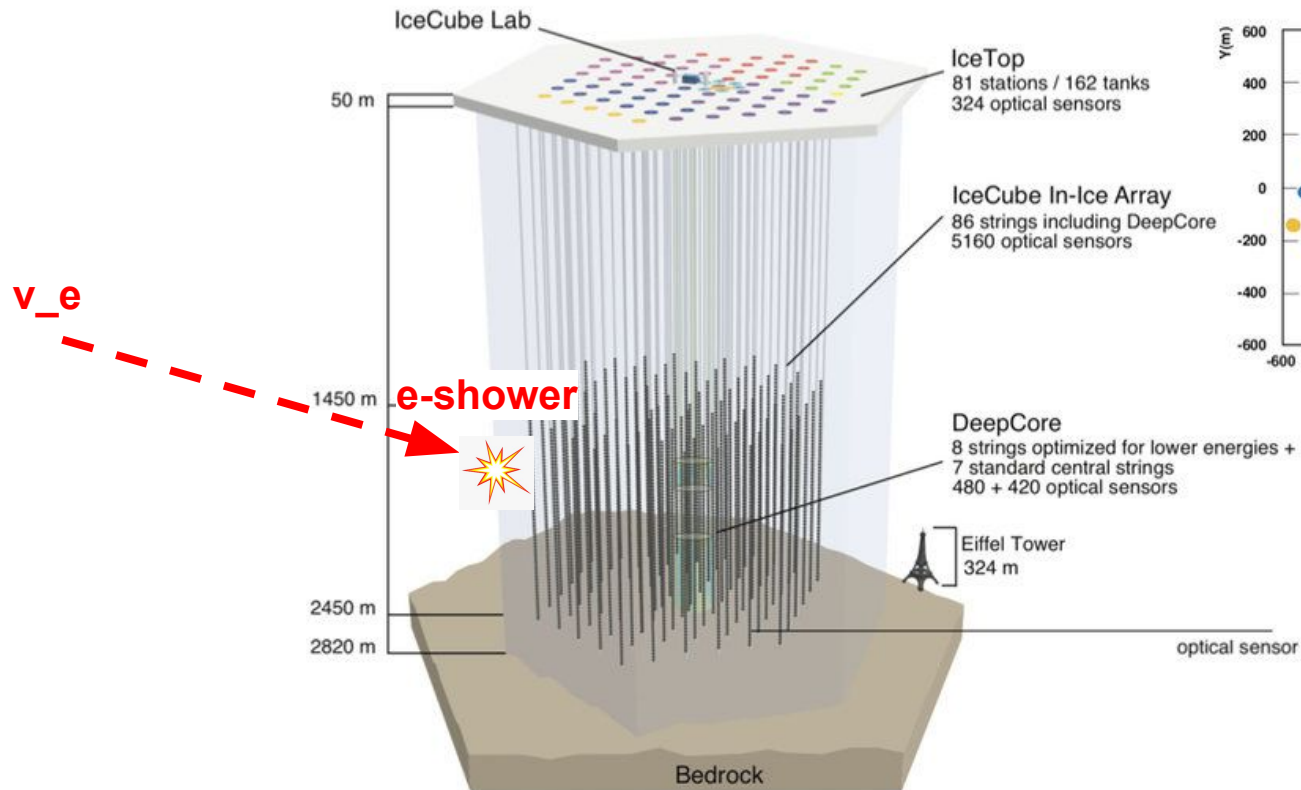


IceCube - the detector



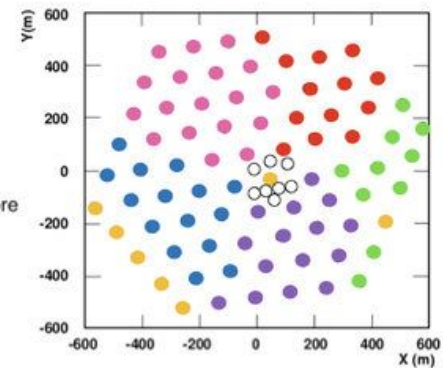
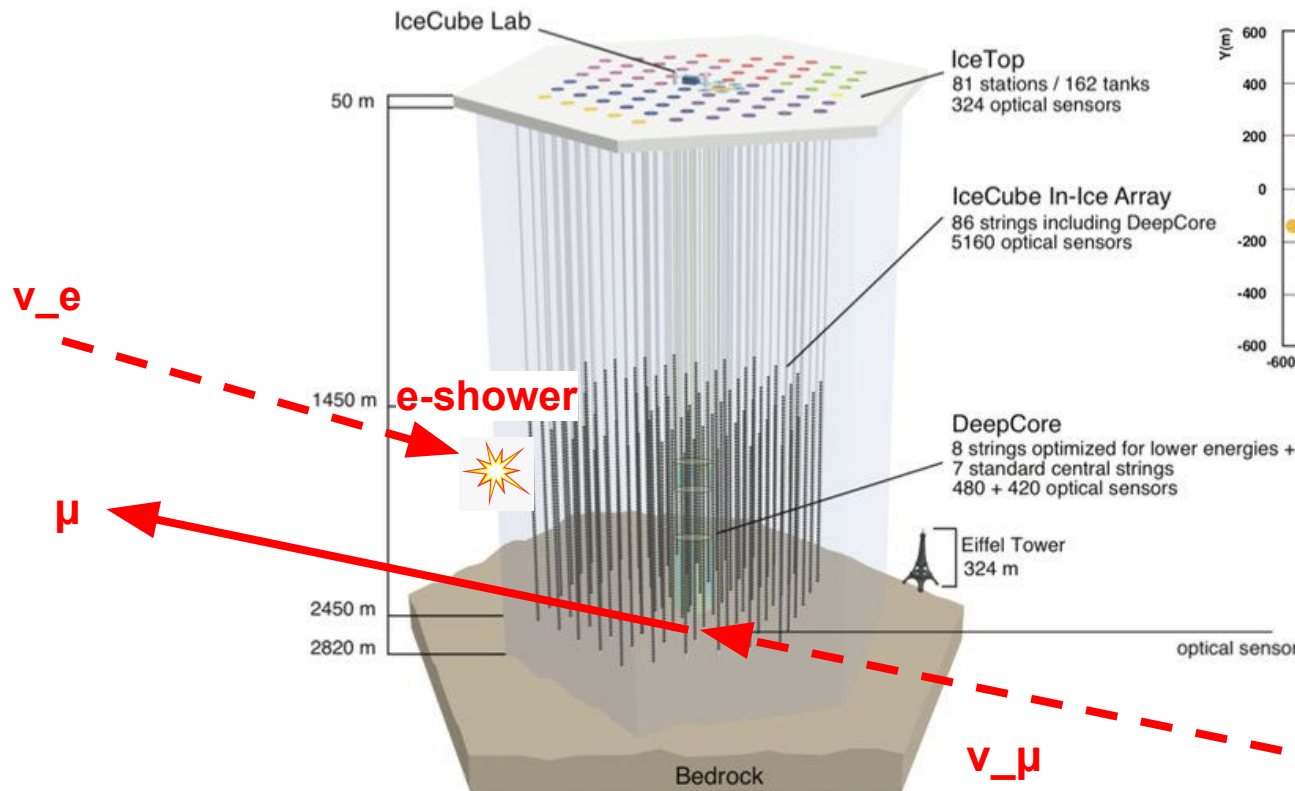
Digital
Optical
Module

IceCube - the detector



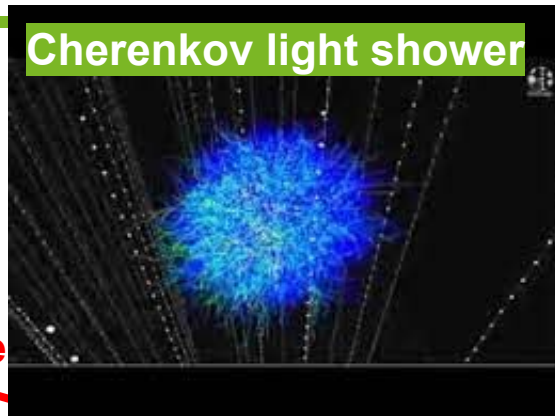
Digital
Optical
Module

IceCube - the detector



Digital Optical Module

Cherenkov light shower



v_e

1450 m **e-shower**



μ

2450 m

2820 m

velocity of charged particle larger than speed of light in ice ($\sim 0.75 * c$) -> emission of Cherenkov light

Cherenkov light muon



IceTop

81 stations / 162 t
324 optical sensor

IceCube In-Ice
86 strings including
5160 optical sensor

DeepCore

8 strings optimized for lower energies +
7 standard central strings
480 + 420 optical sensors

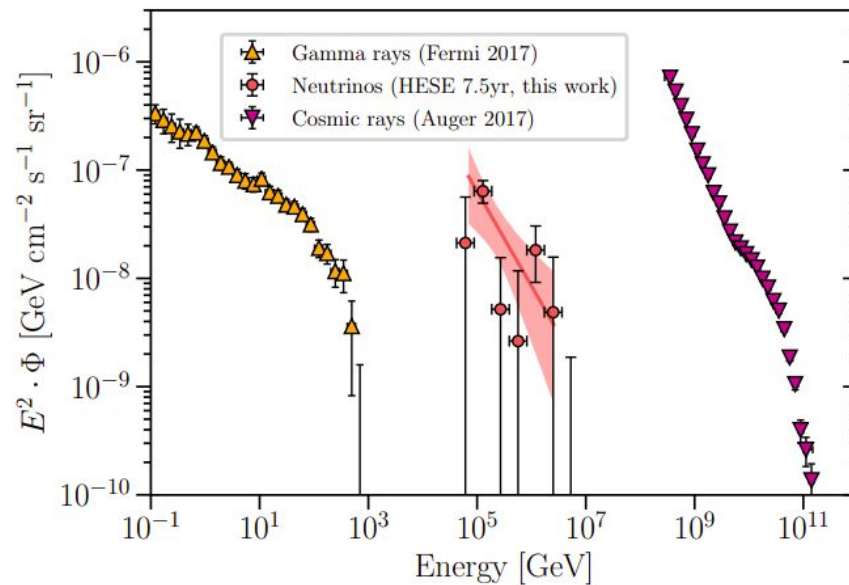
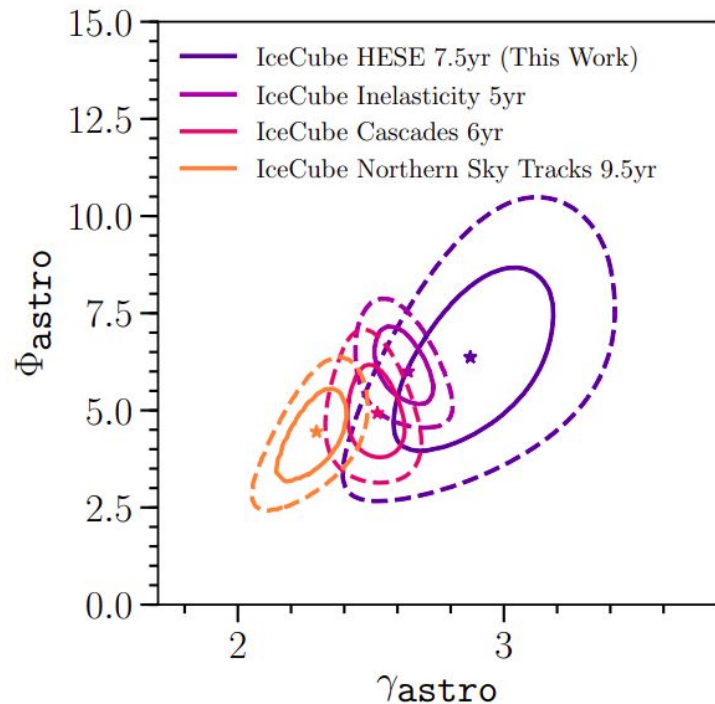
Eiffel Tower
324 m

optical sensor

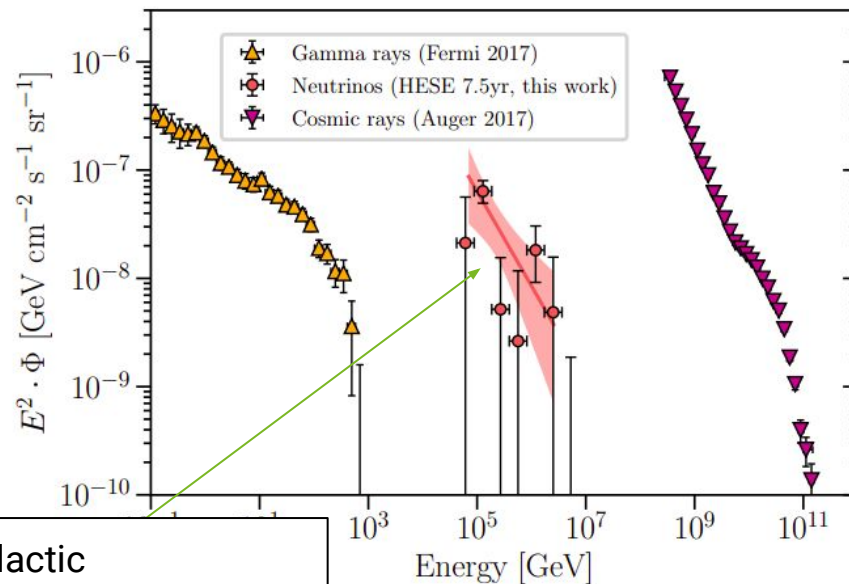
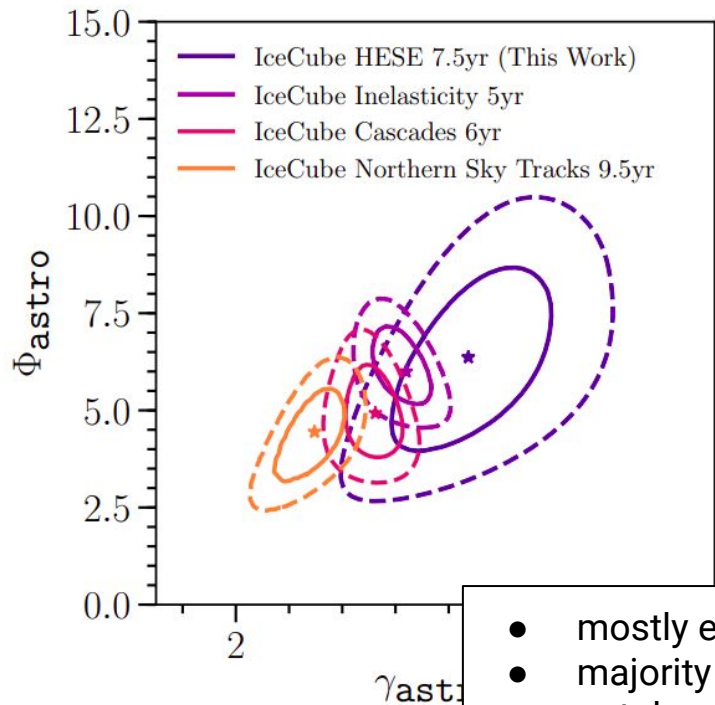


Digital
Optical
Module

Some recent physics results (7.5 years starting event data)



Some recent physics results (7.5 years starting event data)



- mostly extragalactic
- majority not from GeV-blazars
- not dominated by GRBs

Some recent physics results (TXS 0506+056 - the first point source)

[[Previous](#) | [Next](#) | [ADS](#)]

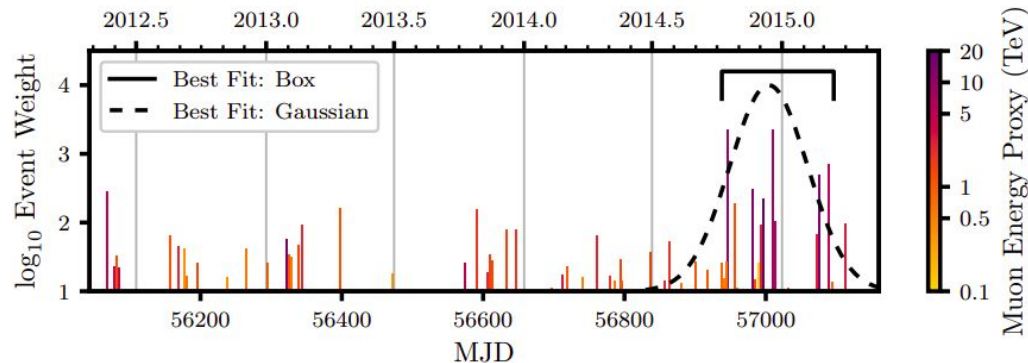
Fermi-LAT detection of increased gamma-ray activity of TXS 0506+056, located inside the IceCube-170922A error region.

ATel #10791; *Yasuyuki T. Tanaka (Hiroshima University), Sara Buson (NASA/GSFC), Daniel Kocevski (NASA/MSFC) on behalf of the Fermi-LAT collaboration*
on 28 Sep 2017; 10:10 UT
Credential Certification: David J. Thompson (David.J.Thompson@nasa.gov)

Subjects: Gamma Ray, Neutrinos, AGN

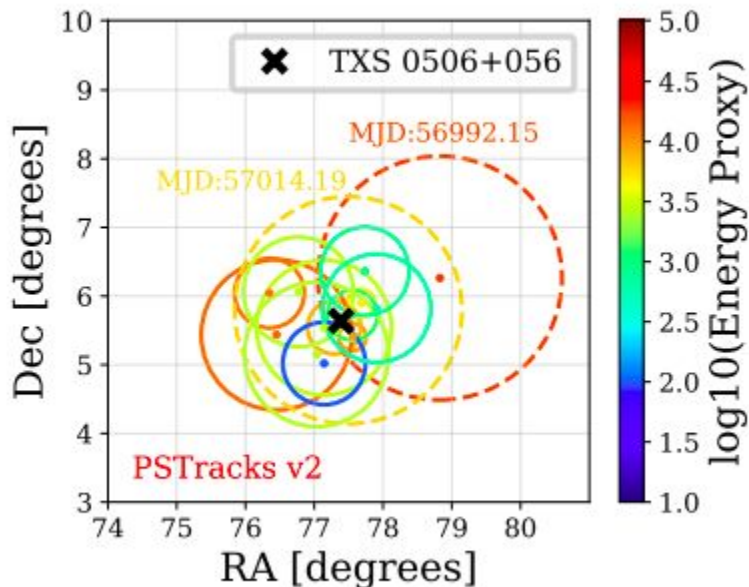


Follow up of ATEL
Gave independent > 3 sigma
Evidence for time dependent neutrino
Emission in historic data

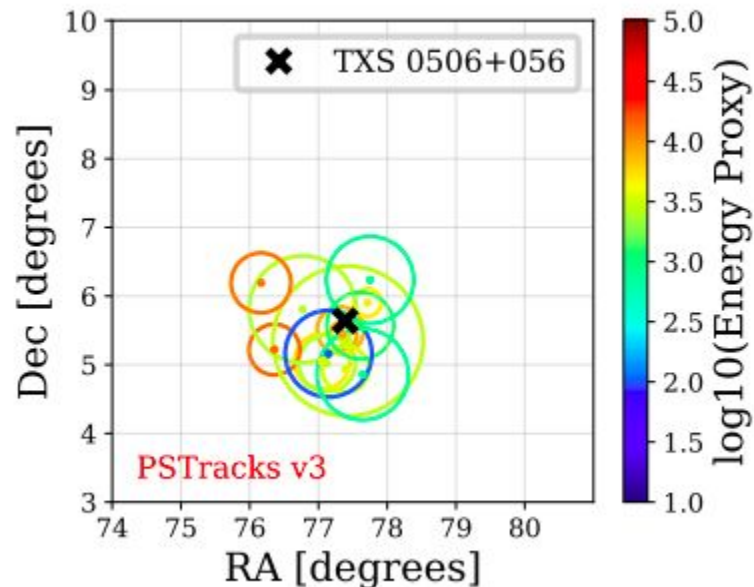


Some recent physics results (Revisiting of the time dependent TXS analysis)

Old processing (> 3 sigma)

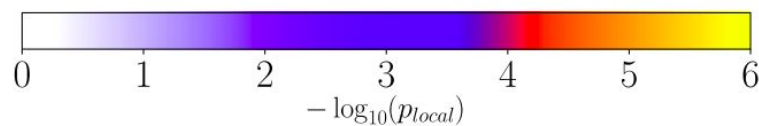
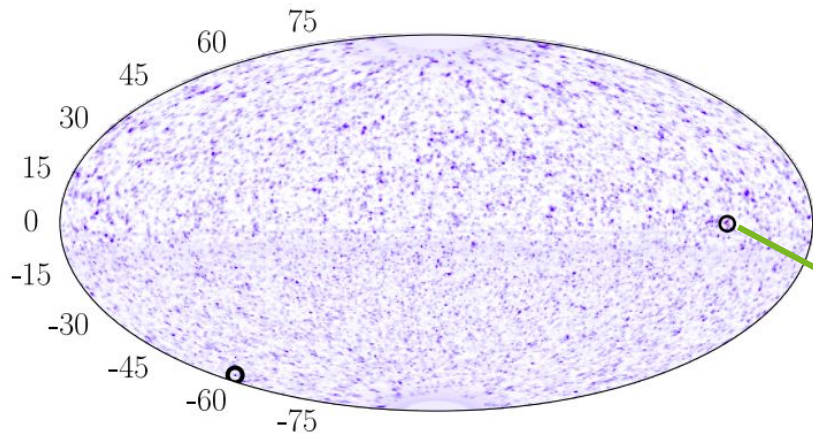


new processing (< 3 sigma)

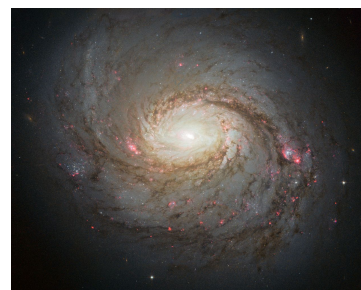


But: combination with ATEL still very non-trivial

Some recent physics results (Another interesting source candidate - NGC 1068)



| | | | | | | | |
|---------------------|------------|--------------|--------------|-------------|------------|-------------|-------------|
| TXS 0518+211 | BLL | 80.44 | 21.21 | 15.7 | 3.8 | 0.92 | 6.6 |
| TXS 0506+056 | BLL | 77.35 | 5.70 | 12.3 | 2.1 | 3.72 | 10.1 |
| PKS 0502+049 | FSRQ | 76.34 | 5.00 | 11.2 | 3.0 | 0.66 | 4.1 |
| S3 0458-02 | FSRQ | 75.30 | -1.97 | 5.5 | 4.0 | 0.33 | 2.7 |
| PKS 0440-00 | FSRQ | 70.66 | -0.29 | 7.6 | 3.9 | 0.46 | 3.1 |
| MG2 J043337+2905 | BLL | 68.41 | 29.10 | 0.0 | 2.7 | 0.28 | 4.5 |
| PKS 0422+00 | BLL | 66.19 | 0.60 | 0.0 | 2.9 | 0.27 | 2.3 |
| PKS 0420-01 | FSRQ | 65.83 | -1.33 | 9.3 | 4.0 | 0.52 | 3.4 |
| PKS 0336-01 | FSRQ | 54.88 | -1.77 | 15.5 | 4.0 | 0.99 | 4.4 |
| NGC 1275 | AGN | 49.96 | 41.51 | 3.6 | 3.1 | 0.41 | 5.5 |
| NGC 1068 | SBG | 40.67 | -0.01 | 50.4 | 3.2 | 4.74 | 10.5 |
| PKS 0235+164 | BLL | 39.67 | 16.62 | 0.0 | 3.0 | 0.28 | 3.1 |
| IC 13807 | FSRQ | 30.48 | 38.80 | 0.0 | 3.8 | 0.20 | 2.6 |



10 year time independent analysis
 With updated source list (based on gamma emission) finds new hot spot in northern sky: NGC 1068 ... TXS is second hot spot

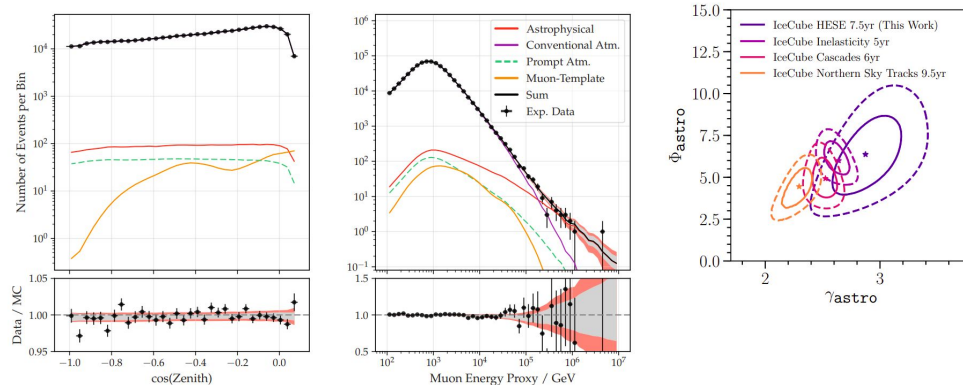
<https://doi.org/10.1103/PhysRevLett.124.051103>

“prototypical Seyfert AGN”

Post-trial p-value: 2.9 sigma

Reconstruction have crucial impact

Reconstructions impact variables in analyses ...
could we do this better?



For certain online alerts, we perform
Time consuming likelihood scans (~ 12 hours +)

Issue: we know they include certain biases, and do
not include certain systematics (ice optical
properties)

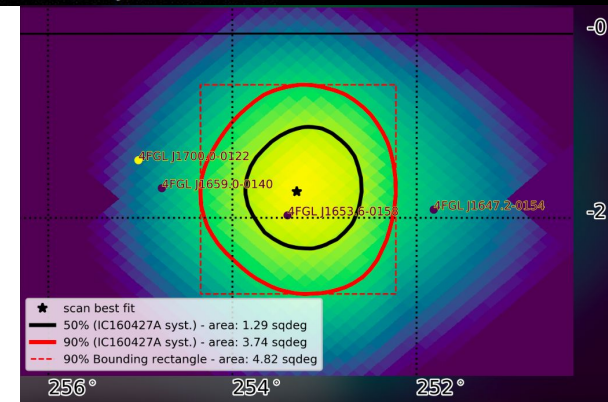
-> conservative over correction

[Previous | Next | ADS]

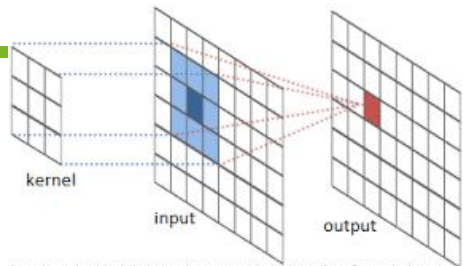
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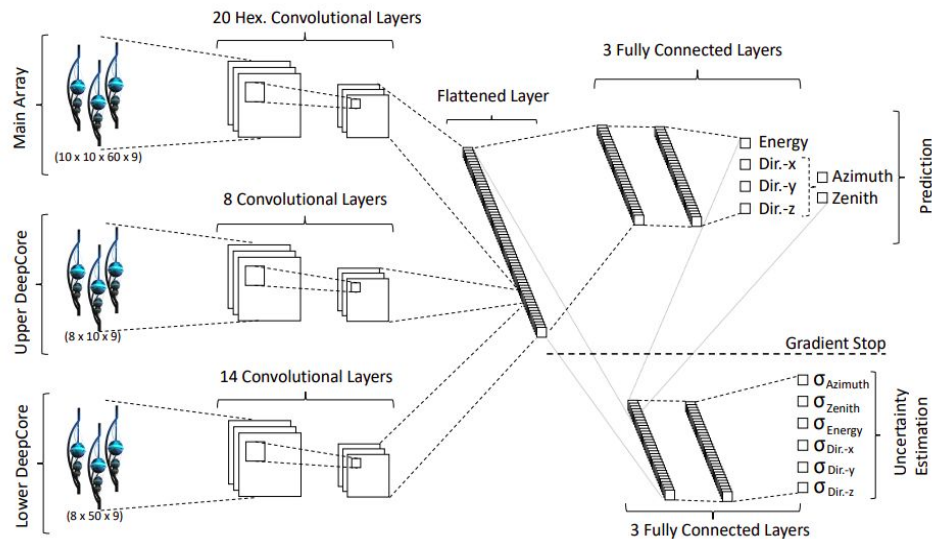
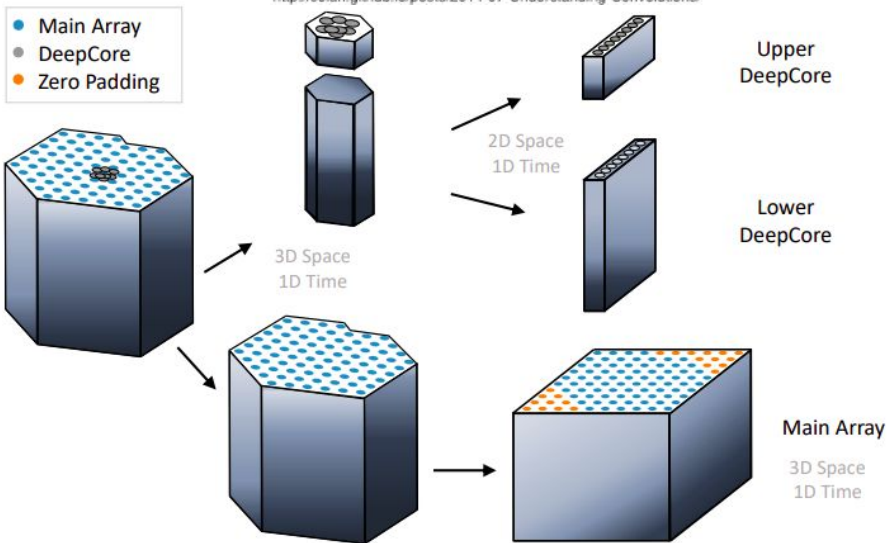
Subjects: Gamma Ray, Neutrinos, AGN

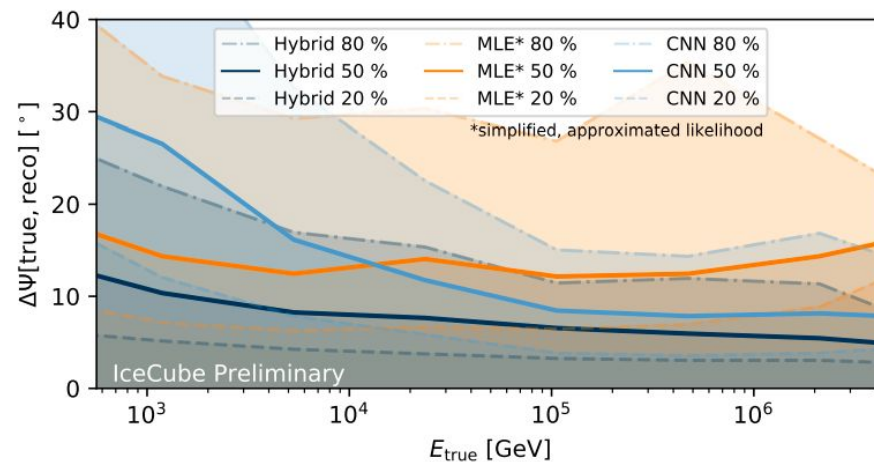
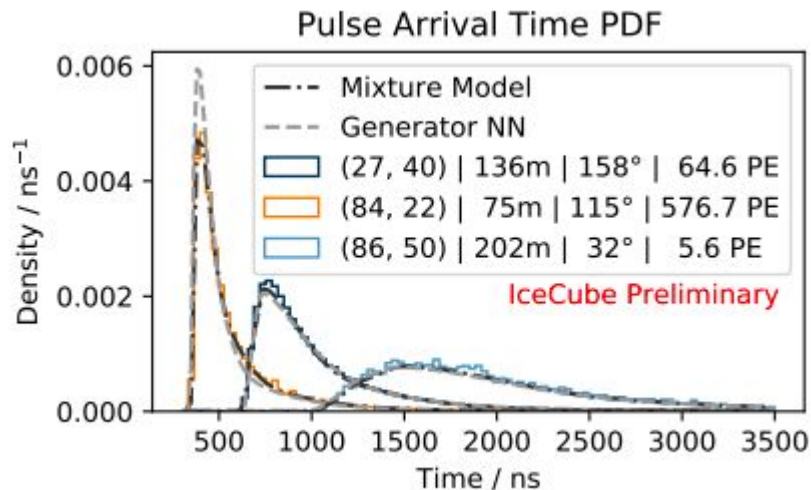


Deep learning for showers



<http://colah.github.io/posts/2014-07-Understanding-Convolutions/>

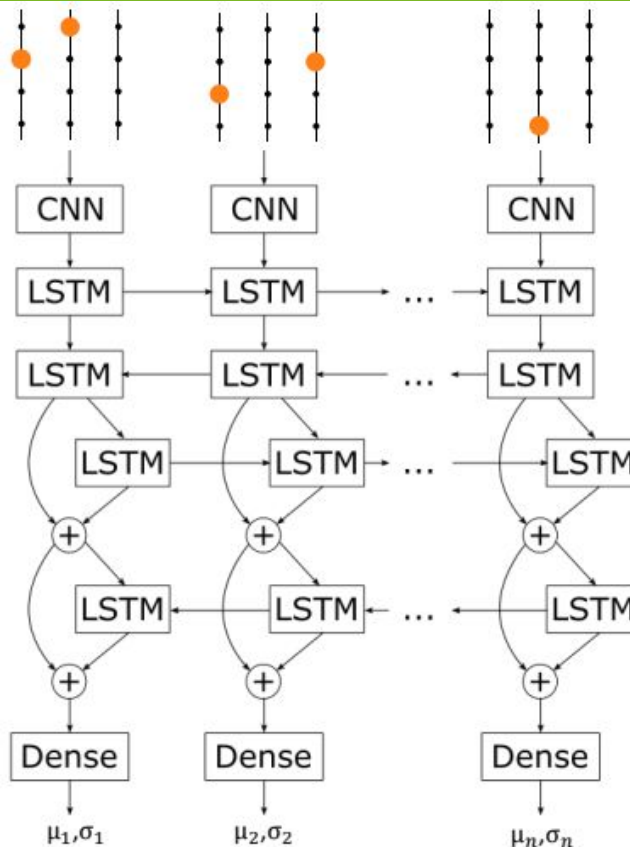
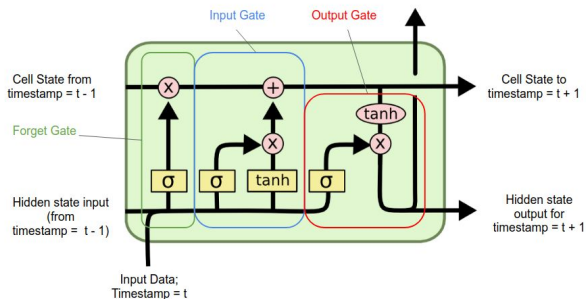




$$\mathcal{L}_{\text{event}}(\vec{x} = \{\vec{c}, \vec{t}\} | \vec{\xi}) = \prod_d \text{Poisson} \left(\sum_i c_{d,i} | \lambda_d(\vec{\xi}) \right) \cdot \prod_i P_d(t_{d,i} | \vec{\xi})^{c_{d,i}}$$

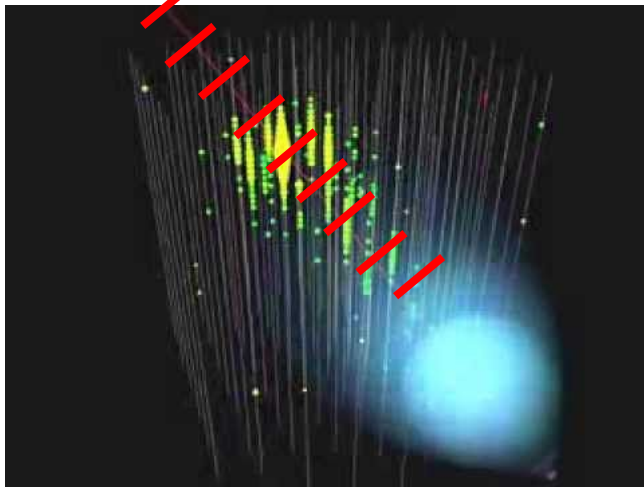
PoS-ICRC2021-1065

CNN / RNN mixture - high energy muons (G. Wrede, ECAP)

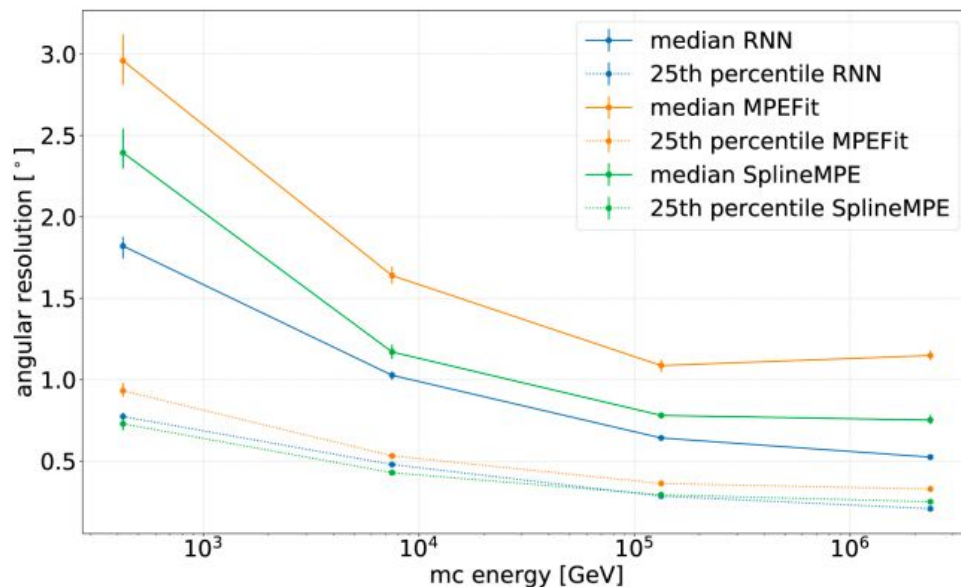
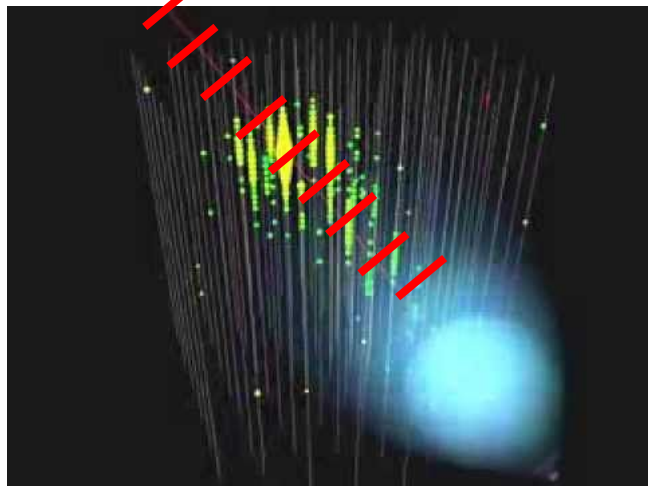
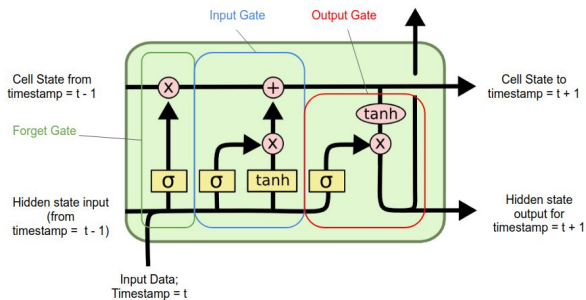


+ Second Network that Takes as input The predicted Positions from The LSTM

Direction + Uncertainty

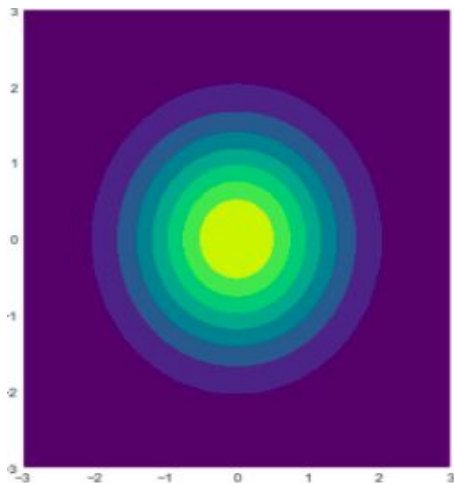


CNN / RNN mixture - high energy muons (G. Wrede, ECAP)



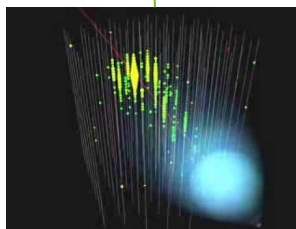
Normalizing flows for Posterior prediction (T. Glösenkamp, ECAP)

base space

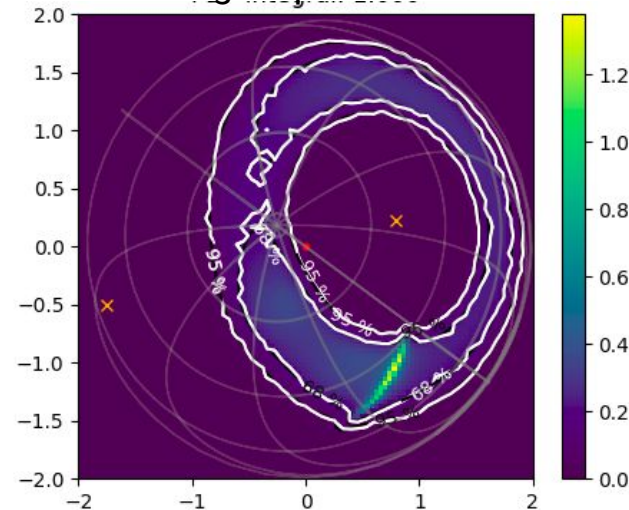


$$\ln(q(z_o)) = \ln(p_b(\rho^{-1}(z_o))) - \ln(\det(J_{\hat{z}}^{\rho\phi}))$$

$$z_o = \rho_{\phi(x)}(z)$$



target space



- 1) Systematics
 - 2) Coverage control
 - 3) Goodness-of-Fit
- (2008.05825)

Advantages of deep learning

- 1) Fast
- 2) Easier to include systematics
- 3) Better performance
Performance ML vs standard likelihood?

Showers



Throughgoing tracks (direction)



Starting Tracks



Energy estimation



Low energy events (<200 GeV)

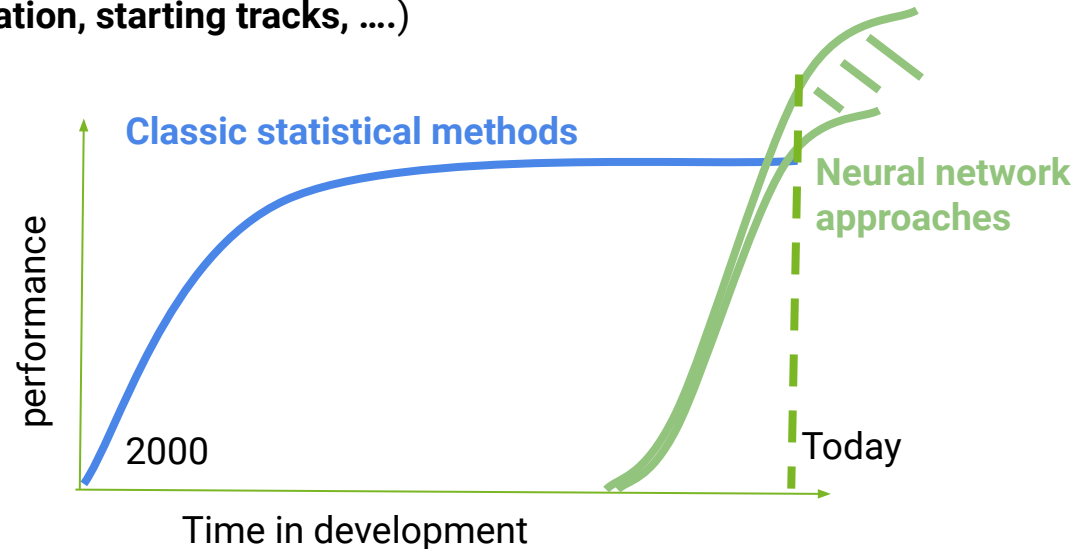


-
- Determining origin of neutrino flux ongoing effort in IceCube
 - Deep-Learning based reconstructions are taking over (including ECAP activities)
 - Already offer faster and more precise reconstructions for a large number of use cases (**shower-type topologies, energy estimation, starting tracks,**)

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Classical ML approaches run into bottleneck
(**20+ years** of history in IceCube/Amanda)

New Machine learning methods
(only **~2-3 years** of investigation) already better in many areas



New physics results and improvements of existing results guaranteed in the next years !