



## Project C3

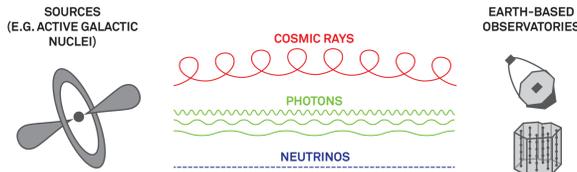
### Multi-level statistical analysis of high-frequency spatio-temporal process data

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## Big Data in Multi-Messenger Astronomy

Physics questions: origins of cosmic rays cosmic particle acceleration dark matter fundamental physics at high energies

### Common Challenges



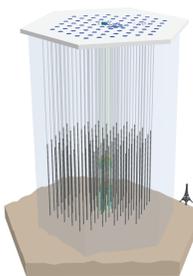
- ▶ Extreme amounts of data
- ▶ Real-time analyses required to trigger follow-up observations
- ▶ Indirect measurement processes for the quantities of interest
- ▶ Signal-to-background ratios from  $1 : 10^3$  to  $1 : 10^{10}$
- ▶ No annotated samples from the real world

The need for and the possibility of extensive physical simulations are both a blessing and a challenge.



### Neutrino Astronomy

- ▶ Extreme signal-to-background ratios  $\nu_\mu: 1 : 10^6, \nu_\tau: 1 : 10^{10}$
- ▶ **IceCube**
- ▶ Below the geographic South Pole
- ▶ 5160 light sensors in 1 km<sup>3</sup> of ice
- ▶ Complex high-dimensional and variable data
- ▶ Limited computational resources and bandwidth (100 GB/d)
- ▶ 3000 Events/s  $\Rightarrow$  1 TB/d



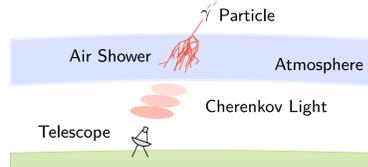
### Gamma-Ray Astronomy

- ▶ Cherenkov radiation measured
- ▶ Atmosphere used as the detector volume
- ▶ 1 gamma-ray : 10<sup>4</sup> hadrons
- ▶ **FACT**
- ▶ New detector technology
- ▶ ~70 Events/s  $\Rightarrow$  750 GB/d
- ▶ Monitoring  $\Rightarrow$  fast alerts
- ▶ **MAGIC**
- ▶ Stereoscopic system
- ▶ ~300 Events/s  $\Rightarrow$  1 TB/d



### Simulations

- ▶ Particle interactions
- ▶ Propagation of leptons
- ▶ Cherenkov light production
- ▶ Propagation of photons
- ▶ Detector optics and electronics

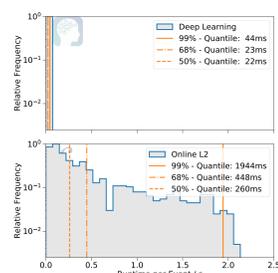


- ▶ Several cluster years of computations
- ▶ More than 5 PB simulated data
- ▶ Many events are discarded in the first analysis steps
- ▶ **Great potential for optimization**

### Real-Time Analysis



- ▶ **FACT [Bockermann et al. 2015]** (A1)
- ▶ Full real-time analysis chain for FACT
- ▶ Online application of pre-trained machine learning models
- ▶ **Best Paper Award ECML 2015**



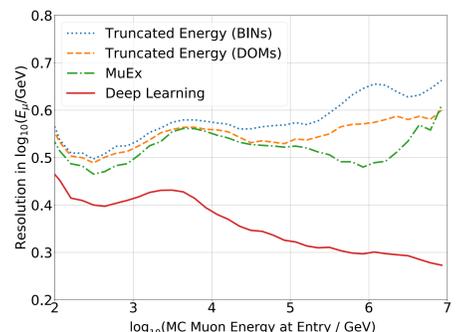
- ▶ **IceCube [Hünnefeld 2017]**
- ▶ Deep learning improves reconstruction while being orders of magnitude faster
- ▶ Constant application time prevents event pileup

### Improved Event Reconstruction through Machine Learning

- ▶ **FACT [Nöthe 2017]**
- ▶ All reconstruction tasks are now solved using random forests (classification & regression)
- ▶ Greatly improved angular resolution
- ▶ Background suppression improved by a factor of 2
- ▶ 50 % gain in sensitivity
- ▶ Evaluation of deep learning started

- ▶ **IceCube [Hünnefeld 2017]**
- ▶ Deep learning improved reconstruction in many areas
- ▶ Random forests improved sensitivity of search for rare tau neutrino events by a factor of 2

- ▶ **MAGIC [Mielke 2017]**
- ▶ Deep learning improved energy resolution by a factor of 2 in the higher energies

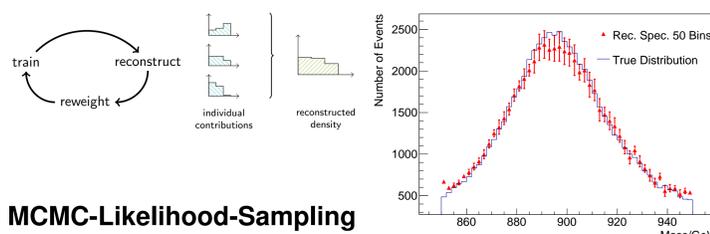


Resolution of energy reconstruction algorithms for IceCube. The novel deep learning reconstruction outperforms other methods by large margins.

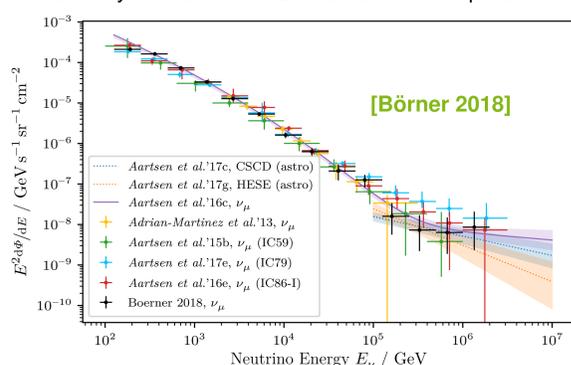
### Novel Unfolding Techniques

- ▶ Indirect measurements result in limited resolution and acceptance
- ▶ Unfolding methods **reconstruct the probability distribution** of energies (spectrum) from the observed quantities

- ▶ **Dortmund Spectrum Estimation Algorithm** (A1 C5)
- ▶ Individual information by multivariate estimation of class probabilities
- ▶ Applied to FACT and LHCb [Bunse et al. 2018] [Ruhe et al. 2016]

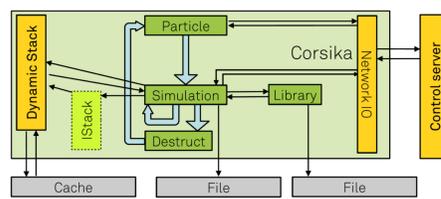


- ▶ **MCMC-Likelihood-Sampling**
- ▶ Estimates uncertainty and correlation
- ▶ Accounts for systematic uncertainties as nuisance parameters



### Increasing Simulation Efficiency

- ▶ **Abort unnecessary simulations** as early as possible [Baack 2016a]
- ▶ Flexible stack implementation allows for prioritization and discarding of particles
- ▶ Perfect agreement between observations and simulations is hard to achieve
- ▶ Improving agreement by feature selection
- ▶ **Remove features that distinguish observations from simulations** [Børner et al. 2017]



- ▶ Active learning for simulation parameters that help training models the most [Bunse et al. 2017]
- ▶ Network interface allows for parallelization of single shower simulations and optimal usage of cluster resources

- ▶ **IceCube**
- ▶ Showers without high-energy muons are aborted
- ▶ Run-time reduced from 169.5 ms to 2.5 ms per shower

- ▶ **FACT [Baack 2016b]**
- ▶ Early abortion of particles unlikely to produce detectable Cherenkov radiation
- ▶ Run-time reduction of  $\approx 80\%$  per shower

