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#### Using machine learning in searches for Lorentz invariance violation



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#### Outline

- Lorentz invariance violation (LIV)
- Motivation for employing Artificial neural networks (ANN)
- ANN for LIV
- Training ANN
- Results (work in progress)



#### Lorentz invariance violation (LIV)

• Attempt at measuring Quantum Gravity



# Lorentz invariance violation (LIV)

- Modified photon dispersion relation
- Usual starting point in searches for effects of QG with gamma rays



- Not a direct consequence of any particular QG model
- Simple way of parametrizing "out of the ordinary" behaviour

# Motivation for using ANN

- Possible consequences of LIV on gamma rays
  - Energy dependent photon group velocity
  - Modified reaction thresholds
  - Modified reaction cross sections
  - Vacuum birefringence
  - ...
- So far only one study of more than one LIV effect using the same data set (Abdalla+ 2019)
  - Effects not considered simultaneously!
- How can multiple effects be tested simultaneously?
  - NB: Single LIV effect can influence different stages of gamma-ray life
    - e.g. modified reaction threshold or cross section affect gamma-ray absorption on background radiation, but also extensive shower development (gamma-ray detection and reconstruction)

# Playground

- Mrk 501 flare from 2014 by H.E.S.S. (Abdalla+ 2019)
  - Simple temporal and spectral distributions
  - ~ 2000 gamma rays 1.3 TeV < E < 30 TeV</li>



#### ANN for LIV – First attempt

- Testing simultaneous LIV effect on photon group velocity & gamma-ray absorption
- Focussing on 2<sup>nd</sup> order correction to dispersion relation
- Input data: Light curve parameters (6) + spectral parameters (2)
- Output data: Light curve parameters (6) + spectral parameters (2) +  $E_{QG}$
- Assuming no source-intrinsic correlation
- Assuming perfect instrument

# Creating training data sets

- 10° gamma rays simulated with  $t_{\mbox{\scriptsize emit}},\, E_{\mbox{\scriptsize emit}}$ 
  - Light curve: sum of two Gaussians
  - Spectral distribution: power-law
- Performed 210 times for  $10^9 < E_{QG} < 10^{30}$
- 210 pools with 10<sup>6</sup> tuples:

-  $t_{emit}$ ,  $E_{emit}$ ,  $t_{detect}$ ,  $E_{detect}$ ,  $P_{survive}$ 



#### Creating training data sets

•  $P_{survive}$  – survival probability for photon of energy  $E_{detect}$ , and considering  $E_{QG}$ 



T. Terzić: ANN for LIV

# Creating training data sets

- Samples created by randomly picking from pool till  $N_{detect} = 2000$ 
  - Accounting for survival probability
  - Fit temporal and spectral distributions at detection to get input data: 6 pars for temporal + 2 for spectral distributions
  - Fit temporal and spectral distributions at emission to get output data: 1 par  $E_{QG}$  + 6 pars for temporal + 2 for spectral distributions
  - Repeat **100** times for each pool (21'000 samples)

#### ANN structure

- First attempts for development purposes
- ANN with 3 hidden layers:
  - Input layer: 8 neurons: Detected light curve parameters
     (6) + detected spectrum parameters (2)
  - Output layer: 9 neurons: Emitted light curve parameters
     (6) + emitted spectrum parameters (2) + E<sub>QG</sub>
  - Hidden layers: 1000 neurons
- 2500 iterations
- Manageable on laptop

#### Results 1

- All parameters mapped to [0, 1] interval
- Clear correlation between injected and reconstructed parameters describing spectrum and light curve



- Some improvement possible by adjusting the mapping  $[par_{min},\,par_{max}]\,\rightarrow\,[0,\,1]$ 



### Results 2

- Some connection between injected and reconstructed LIV parameter, but more work needed
- LIV parameter only in input layer
- Room for improvement:
  - Mapping  $[par_{min}, par_{max}] \rightarrow [0, 1]$
  - Parameter space coverage
  - ANN hyperparameter optimization



#### **Conclusion & Outlook**

- First attempt at simultaneous testing of several LIV effects
- ANN could be a way to go, however
- Make it work in the present form
  - Test different ANN settings (hyperparameters): Learning rate, Number of iterations, Number of hidden layers, Number of hidden units (nods), Choice of activation function
  - Test different choices of training sample
- Expand and investigate different approaches:
  - e.g. Individual events or bin contents as input data (instead of parameters)
  - Introduce realistic detectors (acceptance, energy resolution)
- Estimate sensitivity to systematic effects.
- Compare sensitivity to other analysis methods (e.g. likelihood)
- Combine multiple observations
- Introduce additional LIV effects (e.g. EAS development)
- Introduce additional free parameters (source intrinsic correlation, cosmology, background models, etc.)



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gravity meets qua	MoU- 043/24			
experiments are reaching the precision needed to test the interplay between gravity and quantum systems at ultra-low energies. Investigations of CSO Approval date - 17/05/2024				
these regimes, in particular once they are combined, will provide important clues towards the understanding of the full-fieldged theory of quantum				
gravity.				🗰 End date - 17/09/2028
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