

Quantum diffusion model: ingredients

- Generative model: learn hidden distribution from samples
- **Diffusion model**: connect target distribution to Gaussian through Markov chain
- Quantum neural network: trainable transformation on quantum states



"Quantum diffusion models" - Andrea Cacioppo, Lorenzo Colantonio, Simone Bordoni, Stefano Giagu





Quantum diffusion model: ingredients

- Generative model: learn hidden distribution from samples
- **Diffusion model**: connect target distribution to Gaussian through Markov chain
- Quantum neural network: trainable transformation on quantum states



"Quantum diffusion models" - Andrea Cacioppo, Lorenzo Colantonio, Simone Bordoni, Stefano Giagu







Quantum model (simulation)

Quantum model (simulation) **Classical autoencoder**





Learned distributions

Quantum model (hardware) **Classical autoencoder**







MULTI-BAND PHOTOMETRY AND REDSHIFT ESTIMATION FROM GALAXY IMAGES WITH NORMALIZING FLOWS

Laura Cabayol-Garcia (IFAE/PIC, Barcelona)

Cosmological analysis demands precise 3D mapping of the Universe

PIC

port d'informació científica

Measuring distance to galaxies (redshift)



Currently there are efforts on measuring photo-*z* directly from images with Al. We are developing a normalizing flow model to concurrently measure photometry and photometric redshifts directly from the astronomical images.



By simultaneously measuring photometry across multiple spectral bands and photometric redshifts, our model harnesses the full energy distribution of galaxies, which enables cross-band constraints maximizing the use of information in the data.



SBP: Single-band photometry. Each band is independent, the network does not learn from other bands to predict the flux of one band

MBP: Multi-band photometry. The network has information from all bands when making a prediction.

MBP+z: Multi-band photometry and redshift. The network has information from all bands when making a prediction. Predicts the photometry and the redshift simultaneously

The network benefits from knowing the full SED when making a prediction. It also benefits from predicting the photo-z simultaneously (MTL)

Check poster 36 tomorrow!

Kicking it Off(-shell) with Direct Diffusion

Anja Butter, Tomas Jezo, Michael Klasen, Mathias Kuschick, Sofia Palacios Schweitzer, Tilman Plehn arXiv: 2311.17175





Figure from CMS arXiv:1704.06142v2

Simulating leptonic $t\bar{t}$ decays precisely = Include off-shell processes



Direct Diffusion (DiDi)











LARA ALEGRE EXPLORING THE UNIVERSE WITH RADIO UNIVERSITÄT HEIDELBERG ZUKUNFT **SEIT 1386** VolkswagenStiftung ASTRONOMY AND AI Group led by Dr. Caroline Heneka





1100 Recombination Dark Ages 12 Cosmic Dawn













LARA ALEGRE EXPLORING THE UNIVERSE WITH RADIO UNIVERSITÄT HEIDELBERG ZUKUNFT SEIT 1386 VolkswagenStiftung ASTRONOMY AND A Group led by Dr. **Caroline Heneka**



The Full hydrogen history of the universe





10%





20%

45%

55%





