<u>N.O. Pinciroli Vago</u>, P. Fraternali DeepGraviLens: a multi-modal architecture for classifying gravitational lensing data

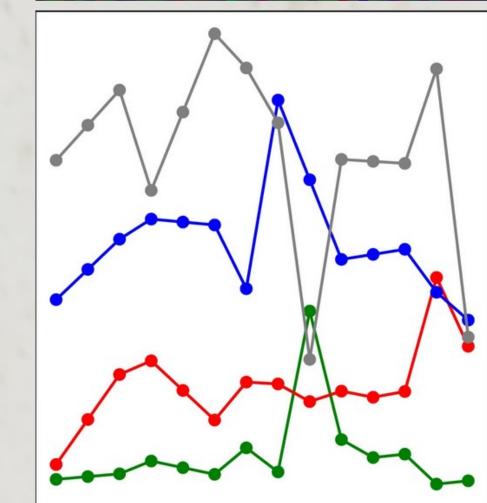
Goal: finding gravitational lenses of transients

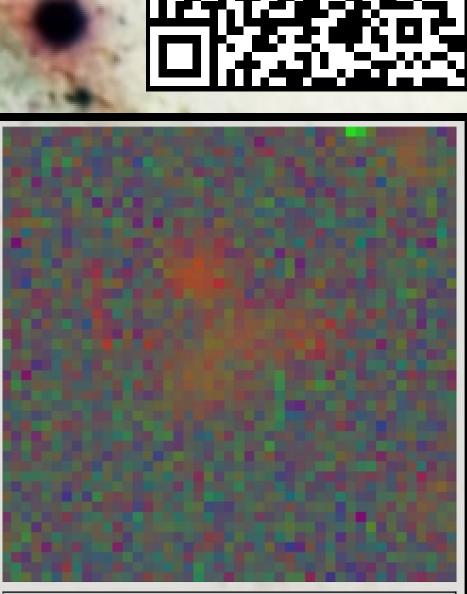
- <u>Why</u>: transients are time-dependent ⇒ images and time series to capture their evolution
- <u>Tools</u>: ensemble of multi-modal neural networks for multi-class classification (no lens, galaxy-galaxy lens, lensed SNIa, lensed SNCC)
- <u>Results</u>: better accuracy (up to +11%), successful testing on real data
- Ongoing work: fine-grained classification, redshift estimation

Datasets

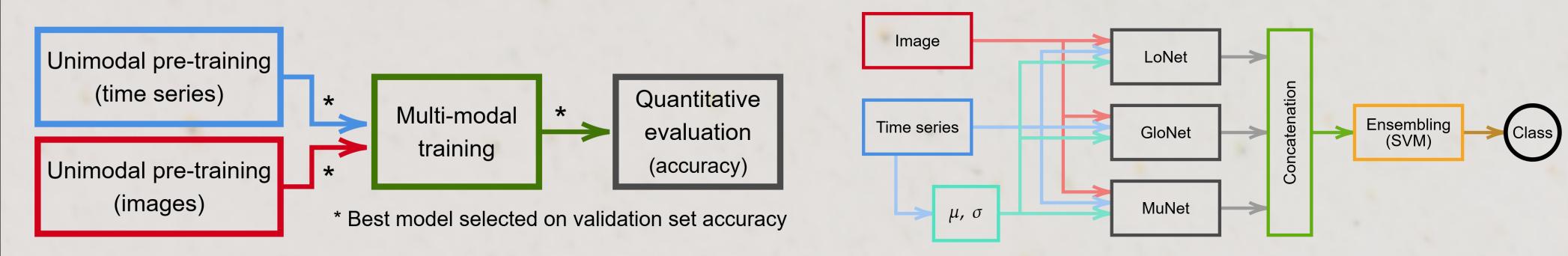
4 datasets (DESI-DOT, LSST, DES-deep, DES-wide) 4 classes (no lens, galaxygalaxy lens, lensed SNIa, lensed SNCC)

20,000 simulated observations per dataset Test on 3,000 simulated observations and **real** observations

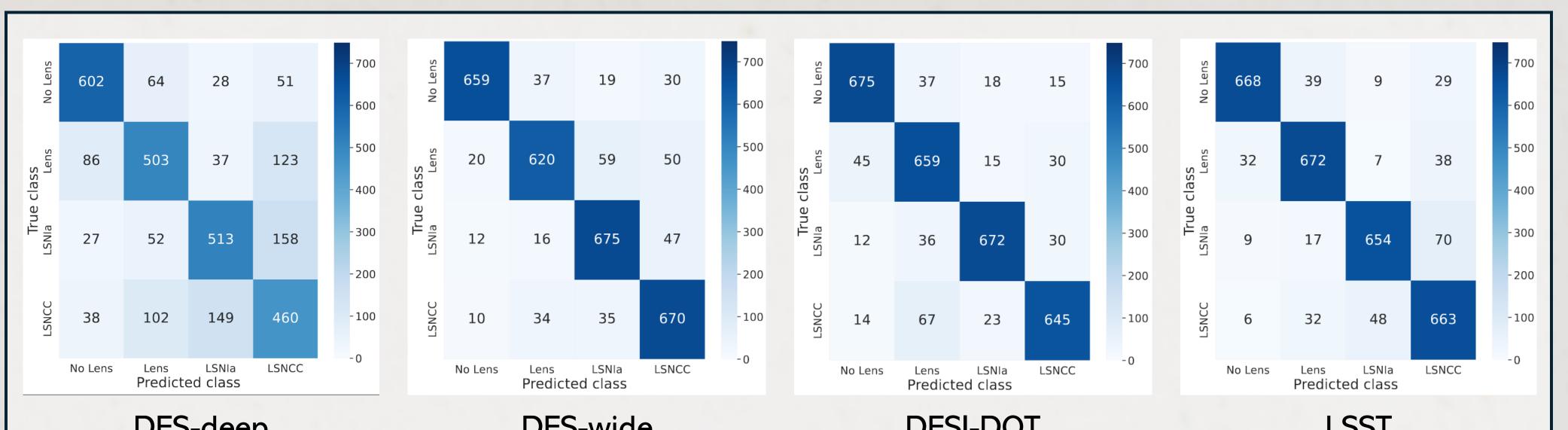




Method



Results



 Conclusions Multimodality is useful (up to +7.3% accuracy wrt single modality) Subnetworks are complementary (⇒ ensembling improves accuracy up to +12%) Overall improvement wrt state of the art (+11% accuracy) 			(88.8% accuracy)
 Multimodality is useful (up to +7.3% accuracy wrt single modality) Subnetworks are complementary (⇒ ensembling improves accuracy up to +12%) 	Conclusions		
- Subnetworks are complementary (\Rightarrow ensembling improves accuracy up to +12%)	Conclusions		
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			-12%)

- Ongoing work on multi-output regression:
 Redshift estimation
 - Finer-grained classification