# Preparing for LSB science with the Rubin Observatory



Characterisation of LSB tidal features from mock images

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### Deep imaging in the era of the Rubin Observatory/LSST

- Hyper Suprime-Cam  $(\mu_{r}^{\lim}(3\sigma,10''\times10'') > 30.5 \text{ mag arcsec}^{-2})$  and other pathfinder instruments give us an idea of what Rubin Observatory will be capable of
  - Such instruments recover many more LSB features around galaxies compared with SDSS ( $\mu_{r}^{lim}(3\sigma, 10'' \times 10'') \sim 24 \text{ mag arcsec}^{-2}$ ) and therefore a more complete record of past low mass accretion events



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- Rubin Observatory will greatly increase the sample size of galaxies with very deep observations
  - Detailed ACDM predictions will allow us to understand the capabilities of this new dataset and make predictions for
    - Frequency and distribution of tidal features as a function of halo mass
    - Biases from orientation, redshift, etc.
    - Surface brightness distribution of tidal features



### The New Horizon Simulation (Dubois+21)

- New Horizon is a high resolution cosmological simulation
  - $\circ$  Contiguous volume of (16 Mpc)<sup>3</sup>
  - High spatial and stellar mass resolution of 34 pc / 10<sup>4</sup> M<sub>o</sub>
  - Sufficient mass resolution to resolve the stellar halo around <MW mass galaxies



- Decompose galaxy stellar haloes into:
  - 1) Dense tidal substructures
  - 2) Diffuse light / debris



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- Very diffuse light in the stellar halo is inaccessible at expected LSST SB limits
  - It accounts for 25% of the total halo light on average



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- A large fraction of flux in more massive galaxies is likely to be detected
- The number of galaxies with detectable tidal features also falls with redshift so that <10% flux in the stellar haloes of MW mass galaxies is detected by z=0.2
  - The low mass / high redshift Universe will remain inaccessible

- ~50 volunteers visually classified tidal features mock Rubin Observatory images
  - Classified for a range of:
    - Limiting surface brightness (single visit → 10 year depth + 35 mag arcsec<sup>-2</sup> to probe beyond the limits of LSST)
    - **Redshift** ( $z = 0.05 \rightarrow 0.8$ )
    - Orientations (projected along *xy*, *xz*, *yz*)

double nucleus	merger remnant	bridge	tidal tail	stellar stream	shell	plume

- Sources of uncertainty
  - Limiting surface brightness and surface brightness dimming





- At sufficient depth, almost 100% of objects were found to have some kind of tidal feature
  - Even at low masses, most galaxies undergo frequent interactions, but not necessarily mergers (Martin+2021)
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  - Mergers and tails are more robustly detected
  - The detection of streams and shells depends more on mass and depth

- Tidal streams and shells are less common in but also weaker in less massive galaxies.
  - his reflects a possible observational bias since the tidal features present in galaxies with a smaller number of tidal features are also likely to be weaker and are therefore more likely to go undetected.

- More sources of uncertainty
  - Orientation
  - Inherent ambiguity in tidal feature classification



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- In most cases, deeper imaging means classifiers and more likely to agree with each other and agree across projections
- However, for some categories, increasing the depth makes classification ambiguous
  - As depth improves, morphologies can become more complex, introducing uncertainty in precise characterisation



#### Conclusions

- After its 10 year survey, LSST will have sufficient depth to resolve a significant fraction of the flux found in tidal substructures of MW galaxy stellar haloes
- Around 75% of flux lies in these denser tidal features rather than more diffuse tidal debris which lie beyond the surface brightness limits accessible to LSST
- At sufficient depth, almost 100% of galaxies (M<sub>\*</sub>/M<sub>o</sub><10<sup>9.5</sup>) possess tidal features
  - But most detectable tidal features are hosted by high mass galaxies at relatively low redshift
- Surface brightness limits, galaxy orientation, redshift, etc. have a clear effect on the ability of expert classifiers to visually identify and characterise tidal features
- Concurrence between classifiers generally improves with deeper imaging but morphologies can become more complex, introducing uncertainty in precise characterisation

#### Please look out for out paper coming soon... Contact: garrethmartin@arizona.edu

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