Low-surface-brightness science and machine-learning for morphological analysis in LSST

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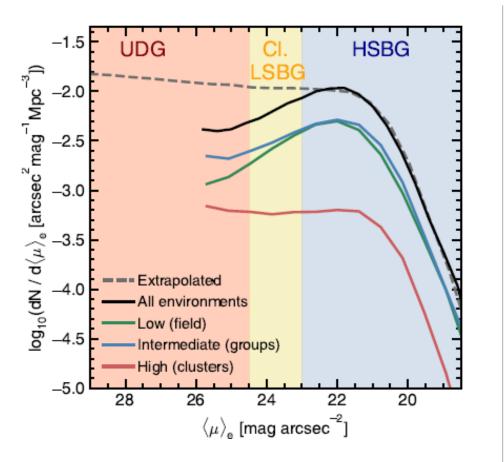
LSST:UK All Hands meeting 13 May 2019

#### Large Synoptic Survey Telescope Galaxies, Dark Matter, and Black Holes: Extragalactic Roadmap

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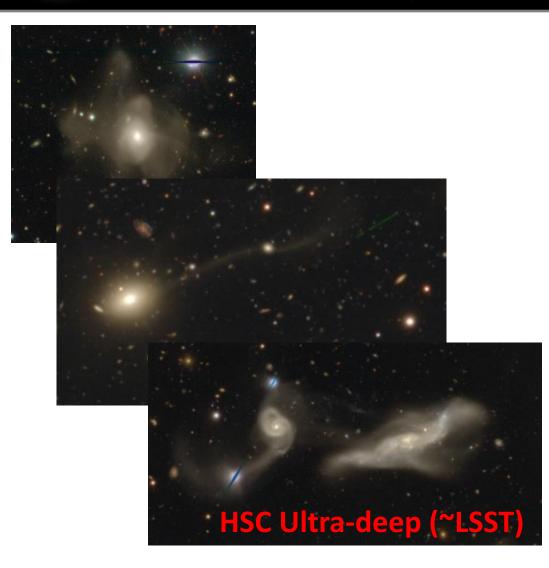
### The significance of the LSB Universe Most galaxies are LSB i.e. undetected by today's surveys



#### Martin, Kaviraj +19

- At 10<sup>10</sup> MSun ~20% of galaxies are LSB
- At 10<sup>8</sup> MSun ~90% of galaxies are LSB
- LSB galaxies are the norm rather than the exception
- Without a good understanding of LSB galaxies our understanding of galaxy evolution remains incomplete

## The significance of the LSB Universe LSB tidal features are key tracers of our theoretical paradigm



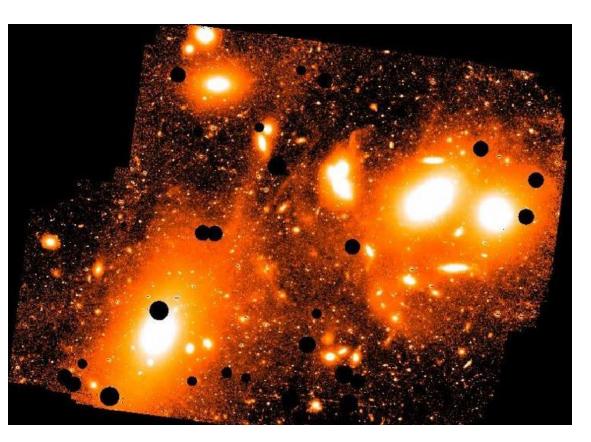
- LSB tidal features encode galaxy assembly histories
- Most mergers have low mass ratios which produce faint features
- Key tracers of our structure formation paradigm

## The significance of the LSB Universe LSB tidal features are key tracers of our theoretical paradigm



#### **Duc** +11

## The significance of the LSB Universe Intra-cluster light (ICL)



Mihos +05

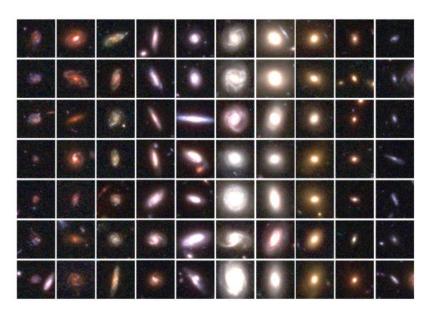
- Galaxy clusters test our cosmological model
- But a significant fraction of baryons in the low-surfacebrightness ICL
  - Utility of clusters closely linked to how well we understand the evolution of the ICL

# The significance of the LSB Universe

- Statistical LSB studies using deep-wide surveys are LSST's niche
- Huge unexplored discovery space a new frontier in galaxy evolution studies
- But...impossible without proper preparatory work
- Project pipelines are not optimised for LSB studies and no project effort planned in this area

# Galaxy morphology

### A fundamental quantity in observational cosmology



- Galaxy morphology is a fundamental parameter for all galaxy evolution studies
- Also key for a plethora of science in observational astrophysics
- ...e.g. an important prior in photo-z pipelines which underpin weak lensing studies and contextual data for transient lightcurve classifications
- Important for the science goals of several science collaborations e.g. Galaxies, AGN, strong lensing, transients etc.
- But...significant challenges due to LSST's data volume and cadence

Part 1: Enabling low-surface-brightness science using LSST

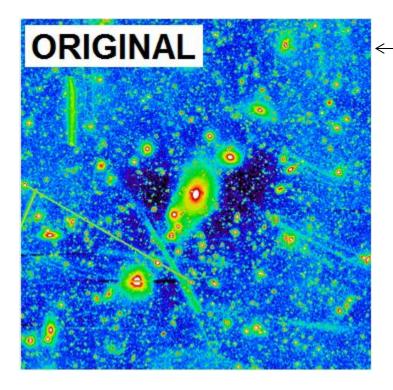
#### The problem:

- LSB structures acutely susceptible to sky over-subtraction and shredding by de-blenders
- Means that galaxy population is incomplete and LSB structures are removed/truncated/shredded -> LSB science impossible.
- Project pipeline known to suffer these effects (optimised for photometry i.e. smaller spatial scales than LSB structures) and no LSB-enabling work is planned by Project

#### The solution (provided by this WP):

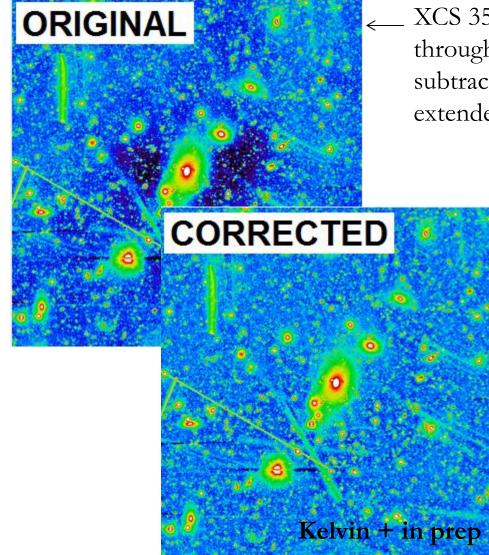
- Optimised sky subtraction to preserve LSB structures at any spatial scale
- Machine-learning algorithms for mitigating shredding

## Phase B project Part 1: Enabling low-surface-brightness science using LSST



XCS 35 in HSC Deep DR1 i-band reduced through the current LSST pipeline. Sky oversubtraction is visible around bright and extended sources.

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XCS 35 in HSC Deep DR1 i-band reduced through the current LSST pipeline. Sky oversubtraction is visible around bright and extended sources.

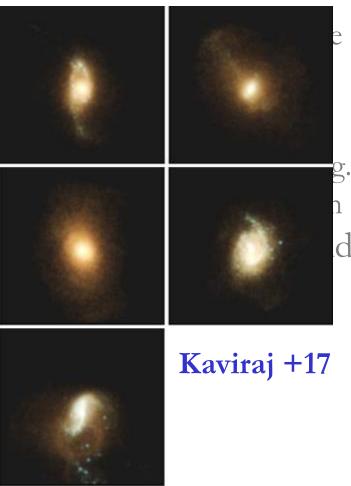
- Fit 2D Sersic models to sources
- Characterise the expected flux in the wings
- Residual between model and science map used to define over-subtraction threshold
- Values below threshold added back

Part 1: Enabling low-surface-brightness science using LSST

- Explore optimal background subtraction strategies for multiple spatial scales
- Benchmark on mock images from cosmological simulations e.g. Horizon-AGN (Kaviraj +17) inserted into HSC and ComCam frames to ensure proper representation of noise/background sources/camera effects in the data

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- Benchmark on mock images from cosmological simulations e.g. Horizon-AGN (Kaviraj+17) inserted into HSC and ComCam frames to ensure proper representation of noise/background sources/camera effects in the data
- The results of this WP will be critical for the global LSST community

Part 2: Measuring galaxy morphologies in LSST

#### The problem:

- LSST data volumes are unprecedented
- Makes visual classification (even using systems like Galaxy Zoo) intractable
- Requirement: **fast and accurate** automated techniques benchmarked via visual classification
- LSST's short cadence is an extra hurdle repeatedly producing training sets for supervised ML impractical

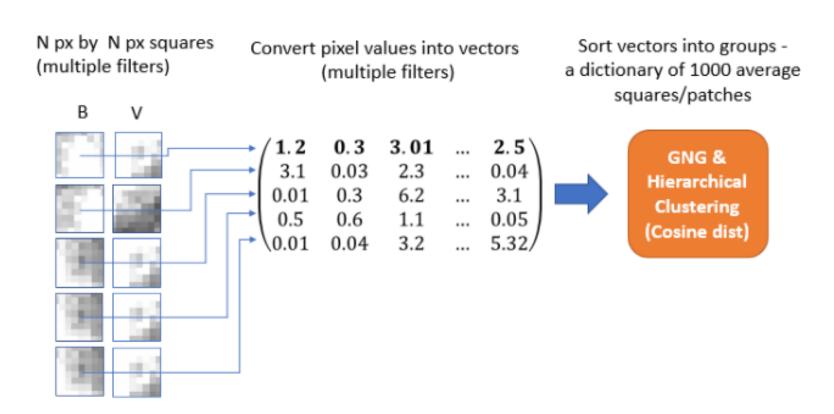
#### The solution (provided by this WP):

- *Unsupervised* ML + benchmarking via visual classification
- UML compresses arbitrarily large galaxy pop. into small number of `morphological clusters'
- ...and these **clusters** are then benchmarked against visual classification

Part 2: Measuring galaxy morphologies in LSST

## An Unsupervised Approach

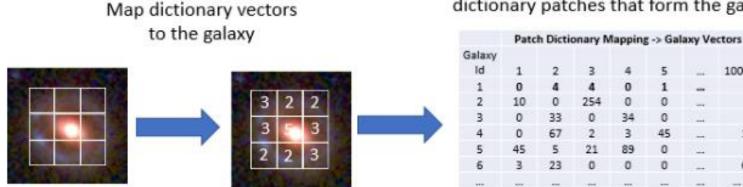
### Step 1 – Build a dictionary of patches



Part 2: Measuring galaxy morphologies in LSST

## An Unsupervised Approach

### Step 2 – Create Galaxy Vector Representations



Each vector is a histogram of the dictionary patches that form the galaxy

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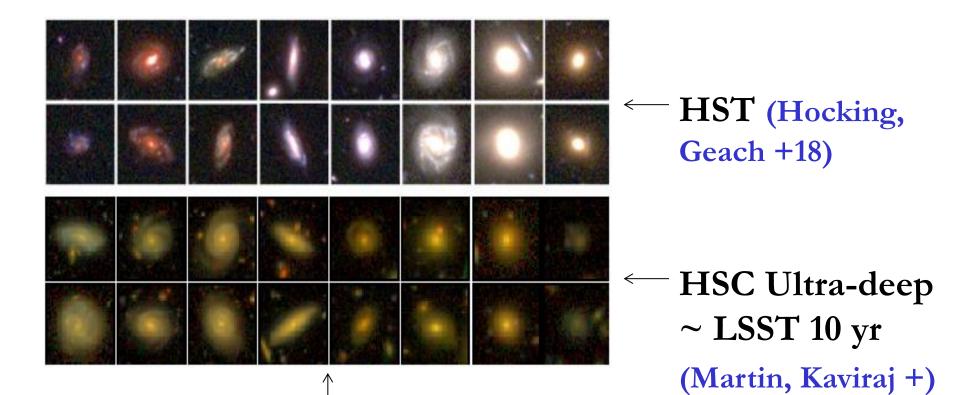
Part 2: Measuring galaxy morphologies in LSST

## An Unsupervised Approach

### Step 3 – Sort Galaxy Vectors into Collections

Sort galaxies into groups and create a         Galaxy Vectors         Galaxy       Galaxy       Gittionary of galaxies         1       0       4       0       1        000         2       10       0       254       0       0        0         3       0       33       0       34       0        0         Hierarchical         Clustering       Colspan="2">Clustering															
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Part 2: Measuring galaxy morphologies in LSST



Tested on bright galaxies only

Part 2: Measuring galaxy morphologies in LSST

- Successfully tested on relatively bright galaxies in HST-CANDELS and HSC Udeep
- Adapt for LSB galaxies (which dominate the number density)
- Develop star-galaxy separation capability (works reasonably well on HST data but untested on HSC/LSST type data)

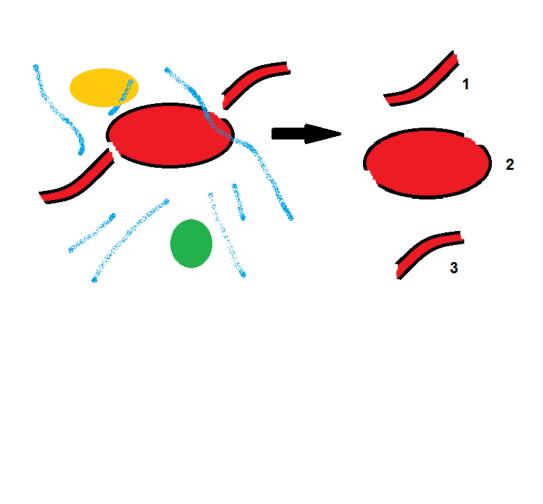
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- Develop capability for strong lens detection (works on HST data but untested on HSC/LSST type imaging)
- Develop capability for mitigating shredding

## Phase B project Part 2: Measuring galaxy morphologies in LSST



- If sky is over subtracted then galaxies and tidal features can be shredded
- Since it works at the **pixel level,** the UML algorithm can mitigate this
- ...because tidal features inherit properties from their parent galaxies

### LSB science and UML for galaxy morphology Summary

Principal aims:

- Enable LSB science using LSST by developing optimised background modelling and sky subtraction at all spatial scales (funded in Phase B)
- (2) Develop unsupervised machine-learning infrastructure for morphological analysis
  - Morphological classification (not funded)
  - Star-galaxy separation (not funded)
  - Lens identification (not funded)
  - Mitiating shredding (funded in Phase B)