Systematic errors in weak lensing surveys and the utility of random galaxy catalogues

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Outline

Weak gravitational lensing

Galaxy intrinsic alignments (IA)

Direct measurements and modelling of IA

• Photo-*z* randoms

 Self-calibration with 3(+)x2-point analyses Correcting clustering biases with Organised Randoms





Weak Cosmological Lensing — 'Cosmic Shear'



 Cosmic shear must tackle challenges wrt shape, redshift & covariance estimation, noise, systematics...

 $C_{\kappa}(\ell) = \frac{9H_0^4}{4c^4} \Omega_{\mathrm{m},0}^2 \sigma_8^2 \int_0^{\chi_{\mathrm{s}}} \mathrm{d}\chi \left(\frac{\chi(\chi_{\mathrm{s}}-\chi)}{\chi_{\mathrm{s}}}\right)^2 P_{\delta}\left(\frac{\ell}{\chi},\chi\right)$



Image courtesy of Bell Labs/Lucent



Intrinsic Alignments

Local galaxies





Background sources

GG

LSS







• We are interested in GG

 Intrinsic alignments give other contributions: II, GI

 We know that red/blue galaxies align differently

• What we observe is some weighted linear combination of all contributions



 Ignoring IA will result in large cosmological parameter biases

 We need a descriptive model, and the data/ statistics to calibrate it

 Currently most popular is the non-linear alignments (NLA) model — issues with wide priors; lack of complexity; degradation of constraints; limited to linear scales







Background source

LSS

 w_{g+}





KiDS – Kilo Degree Survey



Driver+'11, arXiv:1009.0614

$$w_{g+} \propto b_g \int P_{\delta I}(A_{IA}, \beta)$$

 $w_{gg} \propto b_g^2 \int P_{\delta} P_{\delta} Fixed \Lambda CDM model(s)$
 $cosmology$

• Blue galaxies (top) unaligned Red galaxies (bottom) strongly aligned with structure • We constrain A_{IA} (and β) above 6Mpc/h Red signals vary greatly below ~6Mpc/h – satellite/central galaxies align differently

$$P_{\delta I}(k,z) = -A_{IA} \cdot \left\langle \frac{L}{L_{\text{piv}}} \right\rangle^{\beta} \cdot C_1 \frac{a^2 \bar{\rho}(z)}{D(z)} P_{\delta}(k,z)$$









 Red galaxy fraction & central galaxy fraction determine the amplitude of linear-scale alignments in data; both of these evolve with redshift

Fortuna, HJ, + '20, arXiv:2003.02700

- Galaxy colours and central/satellite status • correlate with luminosity
- Do intrinsic alignments also depend on • luminosity?
- (NB: Georgiou, HJ, + '19 also found IA to differ as a function of waveband, if things were not already complicated enough)





- Bright red galaxies do seem to have luminosity-dependent alignments
- 2 regimes? Broken power-law? This would not be captured by a simple redshift dependence for IA: enter halo model
- Need additional exploration of the faintend: enter PAUS





PAUS – Physics of the Accelerating Universe Survey PAUcam @ William Herschel Telescope (WHT) on La Palma ullet40 optical narrow-bands between 4500-8500Å + 6 broad-bands from CFHTLS Aiming for ~100 deg² of targeted observation over several non-contiguous fields Achieving photometric redshift accuracy of ~0.3% Recently finished a pilot study of galaxy IA + clustering in the 19 deg² W3 field (arXiv:2010.09696)













Randoms

 Random galaxy catalogues tell us what is special about the positions of real galaxies

- Galaxies cluster together under gravity; to be quantitative, we must compare their distribution in ratio to a random, i.e. un-clustered, field
- This means we want random points over the same range in 3D as we have galaxies, but they cannot be locally clustered
- Any large-scale structure statistics interested in the positions of objects must make use of randoms











V_{max} randoms – Cole '11; Farrow, +'15



red blue

- We want the mean number density at each redshift; use the luminosity function $\phi(L) = 1/\Sigma_i V_{\max,i}(L)$
- Calculate a maximum redshift z_{max} for each object, given *k*+*e*-corrections:

 $M_{z=0} = m_{obs.} - \mu_{obs.} - k_0(z_{obs.}) + Q z_{obs.}$ $= m_{\text{limit}} - \mu_{\text{max}} - k_0(z_{\text{max}}) + Q(z_{\text{obs.}} - z_{\text{max}})$

 Scatter 'clones' of each galaxy uniformly within the corresponding $V_{\rm max}$, or Gaussiandistribute them around the parent z_{spec} ('windowed' randoms)



V_{max} randoms with photo-z - HJ, + (in prep.)

• Redshift errors \rightarrow errors in *z*_{max}

• Given some spectroscopic objects, we can sample from $n(z_{\text{spec}} | z_{\text{phot}} \pm \delta z)$ for each galaxy

• Mitigate error in z_{max} by generating a distribution $P(V_{\text{max}})$ encoding photoz errors



V_{max} randoms with photo-z - HJ, + (in prep.)



• Creating ensemble randoms from $P(V_{max})$ we avoid over-filling low redshifts and we compensate photo-z degeneracies

• These randoms prevent photo-z induced tilting of measured correlation function \downarrow





PAUS W3 -- all galaxies in $0.1 < z_{phot.} < 0.8$





• With our fancy randoms we can measure IA + clustering in PAUS (+ mock GAMA)

• Probing fainter objects \rightleftharpoons more satellites, over a longer redshift baseline

• Red alignments lost in low S/N ($\leq 2\sigma$); may recover these with full PAUS area

• Blue galaxies again unaligned





3x2pt analysis

 Weak cosmological lensing — cosmic shear — is a powerful probe of the cosmic matter distribution, but is subject to a strong degeneracy between σ_8 and Ω_m

 Jointly analysing cosmic shear with galaxy clustering, and their cross-correlation: galaxy-galaxy lensing (GGL), helps to break the degeneracy and tighten parameter constrains

 The inclusion of galaxy positional statistics also helps with self-calibration of astrophysical and systematic biases, e.g. intrinsic alignments, photo-z and galaxy bias, via nuisance parameterisations





Observing conditions can mess with your positional statistics - HJ, + (in prep.)

- Variable conditions \leftrightarrow PSF, or Galactic extinction, or Milky Way stars, or.... can cause you to systematically fail to detect galaxies
- With a spatially inhomogeneous selection function, galaxy positional statistics can be biased
- We must attempt to mitigate these biases lest they contaminate our cosmological inference















Self-organising maps (SOMs) — HJ, + (in prep.)









SOMs: unsupervised artificial neural networks designed to project high-dimensional data onto a 2D map, preserving topological features of the space

- Useful for dimensionality reduction, classification, data visualisation...
- Training the SOM on survey systematic-tracers, we bin the map into $N_{\rm HC}$ 'hierarchical clusters'
- We can then estimate an expected galaxy density contrast for each cluster \equiv noncontiguous area of sky with correlated systematics





- $\delta_{\rm syst}$ and distribute clones to reflect the systematic fluctuations: Organised Randoms
- remove density field biases
- We interpolate systematic-tracer variables from KiDS-Bright ($r \leq 20$) onto dozens of FLASK simulations, and probabilistically apply the systematic density fluctuations inferred from data

SOM setup: 800C

• We can now map the galaxy density contrast back onto the sky — different $N_{
m HC}$ and systematics yield different

• Using these randoms to measure galaxy positional statistics, we should cancel the systematic fluctuations and



 We are able to reliably corrected clustering biases in KiDS-like mock samples

 Performance scales excellently with number density/systematic pathology, as we see with the faint (shear) sample

 Organised randoms are relatively robust to incomplete systematics information & nonoptimal scale sensitivity, in particular for the faint sample





- randoms
- variable redshift distributions, reflected in the randoms
- galaxy samples used for KiDS cosmic shear

• Bonus: through distribution of clones, we retain the ability to mimic galaxy selection effects in the

• Expanding the redshift axis and selecting on observed colours, we see underdense pointings, and

•These benefits will be fully explored in an upcoming tomographic clustering analysis of the faint



Summary

• We don't understand intrinsic alignments very well yet

 Future analyses must focus upon the variability of alignments with respect to centrals/satellites and galaxy luminosity, with a focus on the unconstrained faintend

 Require new models to accommodate this variability in shear analyses, e.g. perturbation theory, EFT, halo modelling

More complex randoms can be useful for the accurate measurement of statistics

 Promising new methods for clustering bias-cancellation using "organised" randoms from self-organising maps







PAUS Qz₅₀ blue PAUS blue GAMA blue

 10^{1}



 Same as previous figure, but now comparing with signals from best 50% of photo-z.





Completed-KiDS Forecast — demonstrating potential impact of IA priors



HJ+ '19, arXiv:1811.09598 for details of forecast setup







If red vs. blue dominates alignment profiles, why do the full-sample GAMA fits disagree?

HJ+ '19, arXiv:1811.09598







Galaxies tend to have colour gradients along the radial direction — any gravitational shielding should dissipate along the same direction

DEIMOS requires an elliptical Gaussian weight function to suppress image noise; we can fix the physical scale being measured





Georgiou, HJ+ '19, arXiv:1809.03602

• Bluer *g*-band shapes more aligned than rband • Difference comparable to total *r*-band signal • Redder *i*-band shapes also more aligned than *r*-band??





r-band

elliptical galaxies only!

old stars



g-band r-band i-band?

elliptical galaxies only!

star-forming old stars dust?

