

Cosmology from the Abundance of SPT Clusters with DES and HST Weak Lensing

with Sebastian Grandis, Lindsey Bleem, Matthias Klein, Joe Mohr, Tim Schrabback and the South Pole Telescope (SPT) and Dark Energy Survey (DES) collaborations



Sebastian Bocquet, LMU Munich

Image credit: SPT 2024 winter-overs Josh + Kevin

Cosmology From the Big Bang to today



EDSU Tools 2024









Last Journey (on Mira supercomputer)

EDSU Tools 2024



Cluster Cosmology Why do we measure parameters?



- Understand the accelerated expansion & dark matter by testing the ΛCDM model (this talk)
 - Evolution from high-redshift to low-redshift universe.
 (No) S₈ tension?
 - Hubble parameter *H*⁰ tension? (not this talk)
- Understand neutrinos (e.g., hierarchy) through their imprint on large-scale structure
- Understand inflation by measuring tensor-to-scalar ratio r via BB mode in CMB
- Understand the formation of first stars/reionization

EDSU Tools 2024



Cluster Cosmology Why do we measure parameters?



EDSU Tools 2024



 Understand the accelerated expansion & dark matter by testing the Λ CDM model (this talk)

Evolution from high-redshift to low-redshift universe. (No) S₈ tension?

• Hubble parameter H_0 tension? (not this talk)

Understand neutrinos (e.g., hierarchy) through their imprint on large-scale structure

Understand inflation by measuring tensor-to-scalar ratio r via BB mode in CMB

Understand the formation of first stars/reionization





Cluster Cosmology The most massive collapsed objects $\gtrsim 10^{14} M_{\odot}$



Bullet Cluster. X-ray: NASA/CXC/CfA/M.Markevitch, Optical and lensing map: NASA/STScl, Magellan/U.Arizona/D.Clowe, Lensing map: ESO WFI EDSU Tools 2024

- Composition
 - 85–90% dark matter
 - 10–15% ordinary matter, of which
 - ~ 75% (gravitationally heated) gas
 - ~ 25% galaxies/stars
- Somewhat arbitrary (but useful) definition
 - Halo = *entire* thing
 - Cluster = galaxies & gas (what we see)



Large-Scale Structure and Cosmology

Standard Model





EDSU Tools 2024

Warm dark matter



Credit: Katrin Heitmann



Halo Mass Function dN(z)/dlnM — vanilla Λ CDM cosmology



EDSU Tools 2024



Halo Mass Function d*N(z)*/dln*M* – now *w* = -1.1 (instead of -1)





Halo Mass Function Impact of changing dark energy equation of state parameter by 0.1



EDSU Tools 2024





Credit: NASA/CXC/Cinestav/T.Bernal et al.





Credit: NASA, ESA, and J. Lotz, M. Mountain, A. Koekemoer, and the HFF Team (STScI)





Credit: NASA, ESA, the Hubble Heritage Team (STScI/AURA), J. Blakeslee (NRC Herzberg Astrophysics Program, Dominion Astrophysical Observatory), and H. Ford (JHU) http://www.spacetelescope.org/in

EDSU Tools 2024





Bleem et al. (2015)

Credit: NASA, ESA, and J. Lotz, M. Mountain, A. Koekemoer, and the HFF Team (STScI)





Credit: NASA, ESA, the Hubble Heritage Team (STScI/AURA), J. Blakeslee (NRC Herzberg Astrophysics Program, Dominion Astrophysical Observatory), and H. Ford (JHU) http://www.spacetel

EDSU Tools 2024

"Halo Observable Function"





halo observable function



 $\frac{dN}{dM}$ dNdMP(obs | M)dobs

$$\frac{dN}{dobs} = \int dM P(obs \mid M) \frac{dN}{dM}$$

Modeling Framework Observable—Mass Relations

- The bigger a halo, the stronger its SZ, X-ray, optical, lensing signal
 - Supported by theory and numerical simulations
 - These are average relations there is intrinsic scatter, because no two objects are the same
- For the experts:
 - Halo morphology and evolution lead to correlated scatter among observables

Simulations (Angulo+12)

We could use predictions from first principles (e.g., ٠ hydrostatic equilibrium) or numerical simulations

- We could use predictions from first principles (e.g., • hydrostatic equilibrium) or numerical simulations
 - Systematically limited by uncertain astrophysics

- We could use predictions from first principles (e.g., hydrostatic equilibrium) or numerical simulations
 - Systematically limited by uncertain astrophysics
- Weak-lensing-to-mass relation is known within few percents

EDSU Tools 2024

Idealized (exaggerated) situation

Unlensed

Lensed

By TallJimbo - Own work, CC BY-SA 3.0, https://commons.wikimedia.org/w/ index.php?curid=4150002

- We *could* use predictions from first principles (e.g., hydrostatic equilibrium) or numerical simulations
 - Systematically limited by uncertain astrophysics
- Weak-lensing-to-mass relation is known within few percents

EDSU Tools 2024

Idealized (exaggerated) situation

Unlensed

Lensed

index.php?curid=4150002

(b) Tangential shear profile of SPT-CL J0254-5857.

Weak-Lensing Mass Calibration **Robust observable – mass relations**

- We *could* use predictions from first principles (e.g., hydrostatic equilibrium) or numerical simulations
 - Systematically limited by uncertain astrophysics
- Weak-lensing-to-mass relation is known within few percents
 - Used to demonstrate that **hydrostatic mass** \neq halo mass \bullet
 - With lensing measurements of sample clusters, we empirically calibrate the observable – mass relations

EDSU Tools 2024

SPT Clusters with DES and HST Weak Lensing. I. Cluster Lensing and Bayesian **Population Modeling of Multi-Wavelength Cluster Datasets**

S. Bocquet,^{1,*} S. Grandis,^{2,1} L. E. Bleem,^{3,4} M. Klein,¹ J. J. Mohr,^{1,5} M. Aguena,⁶ A. Alarcon,³ S. Allam,⁷ S. W. Allen,^{8,9,10} O. Alves,¹¹ A. Amon,^{12,13} B. Ansarinejad,¹⁴ D. Bacon,¹⁵ M. Bayliss,¹⁶ K. Bechtol,¹⁷ M. R. Becker,³ B. A. Benson,^{18,4,19} G. M. Bernstein,²⁰ M. Brodwin,²¹ D. Brooks,²² A. Campos,²³ R. E. A. Canning,²⁴ J. E. Carlstrom,^{18, 4, 25, 3, 26} A. Carnero Rosell,^{27, 6, 28} M. Carrasco Kind,^{29, 30} J. Carretero,³¹ R. Cawthon,³² C. Chang,^{18,4} R. Chen,³³ A. Choi,³⁴ J. Cordero,³⁵ M. Costanzi,^{36,37,38} L. N. da Costa,⁶ M. E. S. Pereira,³⁹ C. Davis,⁴⁰ J. DeRose,⁴¹ S. Desai,⁴² T. de Haan,^{43,44} J. De Vicente,⁴⁵ H. T. Diehl,⁷ S. Dodelson,^{23,46} P. Doel,²² C. Doux,^{20,47} A. Drlica-Wagner,^{18,7,4} K. Eckert,²⁰ J. Elvin-Poole,⁴⁸ S. Everett,⁴⁹ I. Ferrero,⁵⁰ A. Ferté,⁵¹ A. M. Flores,^{9,8} J. Frieman,^{7,4} J. García-Bellido,⁵² M. Gatti,²⁰ G. Giannini,³¹ M. D. Gladders,^{18,4} D. Gruen,¹ R. A. Gruendl,^{29,30} I. Harrison,⁵³ W. G. Hartley,⁵⁴ K. Herner,⁷ S. R. Hinton,⁵⁵ D. L. Hollowood,⁵⁶ W. L. Holzapfel,⁵⁷ K. Honscheid,^{58,59} N. Huang,⁵⁷ E. M. Huff,⁴⁹ D. J. James,⁶⁰ M. Jarvis,²⁰ G. Khullar,^{4,18} K. Kim,¹⁶ R. Kraft,⁶¹ K. Kuehn,^{62,63} N. Kuropatkin,⁷ F. Kéruzoré,³ S. Lee,⁴⁹ P.-F. Leget,⁴⁰ N. MacCrann,⁶⁴ G. Mahler,^{65,66} A. Mantz,^{8,9} J. L. Marshall,⁶⁷ J. McCullough,⁴⁰ M. McDonald,⁶⁸ J. Mena-Fernández,⁴⁵ R. Miquel,^{69,31} J. Myles,^{9,40,51} A. Navarro-Alsina,⁷⁰ R. L. C. Ogando,⁷¹ A. Palmese,²³ S. Pandey,²⁰ A. Pieres,^{6,71} A. A. Plazas Malagón,^{40,51} J. Prat,^{18,4} M. Raveri,⁷² C. L. Reichardt,¹⁴ J. Roberson,¹⁶ R. P. Rollins,³⁵ A. K. Romer,⁷³ C. Romero,⁷⁴ A. Roodman,^{40,51} A. J. Ross,⁵⁸ E. S. Rykoff,^{40,51} L. Salvati,^{75,76,77} C. Sánchez,²⁰ E. Sanchez,⁴⁵ D. Sanchez Cid,⁴⁵ A. Saro,^{78,77,76,79,80} T. Schrabback,^{81, 2} M. Schubnell,¹¹ L. F. Secco,⁴ I. Sevilla-Noarbe,⁴⁵ K. Sharon,⁸² E. Sheldon,⁸³ T. Shin,⁸⁴ M. Smith,⁸⁵ T. Somboonpanyakul,^{86,40} B. Stalder,⁶¹ A. A. Stark,⁶¹ V. Strazzullo,^{76,87,77} E. Suchyta,⁸⁸ M. E. C. Swanson,²⁹ G. Tarle,¹¹ C. To,⁵⁸ M. A. Troxel,³³ I. Tutusaus,⁸⁹ T. N. Varga,^{90, 5, 91} A. von der Linden,⁸⁴ N. Weaverdyck,^{11,41} J. Weller,^{5,91} P. Wiseman,⁸⁵ B. Yanny,⁷ B. Yin,²³ M. Young,⁹² Y. Zhang,⁹³ and J. Zuntz⁹⁴ (the DES and SPT Collaborations)

arXiv:2310:12213 — PRD accepted

Image credit: SPT 2018 winter-overs Adam & Joshua

SPT Clusters with DES and HST Weak Lensing. II. Cosmological Constraints from the Abundance of Massive Halos

S. Bocquet,^{1,*} S. Grandis,^{2,1} L. E. Bleem,^{3,4} M. Klein,¹ J. J. Mohr,^{1,5} T. Schrabback,^{2,6} T. M. C. Abbott,⁷ P. A. R. Ade,⁸ M. Aguena,⁹ A. Alarcon,³ S. Allam,¹⁰ S. W. Allen,^{11, 12, 13} O. Alves,¹⁴ A. Amon,^{15, 16} A. J. Anderson,¹⁰ J. Annis,¹⁰ B. Ansarinejad,¹⁷ J. E. Austermann,^{18,19} S. Avila,²⁰ D. Bacon,²¹ M. Bayliss,²² J. A. Beall,¹⁸ K. Bechtol,²³ M. R. Becker,³ A. N. Bender,^{3, 4, 24} B. A. Benson,^{24, 4, 10} G. M. Bernstein,²⁵ S. Bhargava,²⁶ F. Bianchini,^{11, 12, 13} M. Brodwin,²⁷ D. Brooks,²⁸ L. Bryant,²⁹ A. Campos,³⁰ R. E. A. Canning,³¹ J. E. Carlstrom,^{24,4,32,3,29} A. Carnero Rosell,^{33,9,34} M. Carrasco Kind,^{35,36} J. Carretero,²⁰ F. J. Castander,^{37,38} R. Cawthon,³⁹ C. L. Chang,^{4,3,24} C. Chang,^{24,4} P. Chaubal,¹⁷ R. Chen,⁴⁰ H. C. Chiang,^{41,42} A. Choi,⁴³ T-L. Chou,^{4,32} R. Citron,⁴⁴ C. Corbett Moran,⁴⁵ J. Cordero,⁴⁶ M. Costanzi,^{47,48,49} T. M. Crawford,^{4,24} A. T. Crites,⁵⁰ L. N. da Costa,⁹ M. E. S. Pereira,⁵¹ C. Davis,¹¹ T. M. Davis,⁵² J. DeRose,⁵³ S. Desai,⁵⁴ T. de Haan,^{55, 56} H. T. Diehl,¹⁰ M. A. Dobbs,^{41, 57} S. Dodelson,^{30, 58} C. Doux,^{25, 59} A. Drlica-Wagner,^{24, 10, 4} K. Eckert,²⁵ J. Elvin-Poole,⁶⁰ S. Everett,⁶¹ W. Everett,⁶² I. Ferrero,⁶³ A. Ferté,¹³ A. M. Flores,^{12,11} J. Frieman,^{10,4} J. Gallicchio,^{4,64} J. García-Bellido,⁶⁵ M. Gatti,²⁵ E. M. George,⁶⁶ G. Giannini,^{20,4} M. D. Gladders,^{24,4} D. Gruen,¹ R. A. Gruendl,^{35, 36} N. Gupta,⁶⁷ G. Gutierrez,¹⁰ N. W. Halverson,^{62, 19} I. Harrison,⁶⁸ W. G. Hartley,⁶⁹ K. Herner,¹⁰ S. R. Hinton,⁵² G. P. Holder,^{36, 70, 57} D. L. Hollowood,⁷¹ W. L. Holzapfel,⁷² K. Honscheid,^{73, 74} J. D. Hrubes,⁴⁴ N. Huang,⁷² J. Hubmayr,¹⁸ E. M. Huff,⁶¹ D. Huterer,¹⁴ K. D. Irwin,^{13,12} D. J. James,⁷⁵ M. Jarvis,²⁵ G. Khullar,^{4, 24} K. Kim,²² L. Knox,⁷⁶ R. Kraft,⁷⁵ . Krause,⁷⁷ K. Kuehn,^{78, 79} N. Kuropatkin,¹⁰ F. Kéruzoré,³ O. Lahav,²⁸ A. T. Lee,^{72,80} P.-F. Leget,⁸¹ D. Li,^{18,13} H. Lin,¹⁰ A. Lowitz,²⁴ N. MacCrann,⁸² G. Mahler,^{83,84} A. Mantz,^{11,12} J. L. Marshall,⁸⁵ J. McCullough,⁸¹ M. McDonald,⁸⁶ J. J. McMahon,^{4,32,24} J. Mena-Fernández,⁸⁷ F. Menanteau,^{35,36} S. S. Meyer,^{4,32,24,29} R. Miquel,^{88,20} J. Montgomery,⁴¹ J. Myles,⁸⁹ T. Natoli,^{24,4} A. Navarro-Alsina,⁹⁰ J. P. Nibarger,¹⁸ G. I. Noble,⁹¹ V. Novosad,⁹² R. L. C. Ogando,⁹³ Y. Omori,⁴ S. Padin,⁹⁴ S. Pandey,²⁵ P. Paschos,²⁹ S. Patil,¹⁷ A. Pieres,^{9,93} A. A. Plazas Malagón,^{81,13} A. Porredon,⁹⁵ J. Prat,^{24,4} C. Pryke,⁹⁶ M. Raveri,⁹⁷ C. L. Reichardt,¹⁷ J. Roberson,²² R. P. Rollins,⁴⁶ C. Romero,⁷⁵ A. Roodman,^{81,13} J. E. Ruhl,⁹⁸ E. S. Rykoff,^{81,13} B. R. Saliwanchik,⁹⁹ L. Salvati,^{100,101,102} C. Sánchez,²⁵ E. Sanchez,¹⁰³ D. Sanchez Cid,¹⁰³ A. Saro,^{104,102,101,105,106} K. K. Schaffer,^{4,29,107} L. F. Secco,⁴ I. Sevilla-Noarbe,¹⁰³ K. Sharon,¹⁰⁸ E. Sheldon,¹⁰⁹ T. Shin,¹¹⁰ C. Sievers,⁴⁴ G. Smecher,^{41,111} M. Smith,¹¹² T. Somboonpanyakul,¹¹³ M. Sommer,⁶ B. Stalder,⁷⁵ A. A. Stark,⁷⁵ J. Stephen,²⁹ V. Strazzullo,^{101,102} E. Suchyta,¹¹⁴ G. Tarle,¹⁴ C. To,⁷³ M. A. Troxel,⁴⁰ C. Tucker,⁸ I. Tutusaus,¹¹⁵ T. N. Varga,^{116, 5, 117} T. Veach,¹¹⁸ J. D. Vieira,^{36, 70} A. Vikhlinin,⁷⁵ A. von der Linden,¹¹⁰ G. Wang,³ N. Weaverdyck,^{14,53} J. Weller,^{5,117} N. Whitehorn,¹¹⁹ W. L. K. Wu,¹³ B. Yanny,¹⁰ V. Yefremenko,³ B. Yin,³⁰ M. Young,⁹¹ J. A. Zebrowski,^{4, 24, 10} Y. Zhang,⁷ H. Zohren,⁶ and J. Zuntz¹²⁰ (the SPT and DES Collaborations)

arXiv:2401.02075 — PRD accepted

The South Pole Telescope (SPT)

10-meter sub-mm quality wavelength telescope

90, 150, 220 GHz and 1.6, 1.2, 1.0 arcmin resolution

2007: SPT-SZ

960 detectors 90,150,220 GHz

2012: SPTpol

1600 detectors 90,150 GHz +Polarization

2017: SPT-3G

~15,200 detectors 90,150,220 GHz +Polarization

Find clusters Sunyaev-Zel'dovich (SZ) Effect

EDSU Tools 2024

Clean and well-understood selection of cluster candidates

Out to highest redshifts where clusters exist!

SPTpol @ 150 GHz

Why use SZ-selected clusters? Three approaches: X-ray, Optical, SZ

EDSU Tools 2024

Sebastian Bocquet — LMU Munich

20

The Dark Energy Survey 5000 deg² galaxies & weak lensing

Catalog of SPT-selected cluster candidates needs

- Confirmation
- Cluster redshifts
- Weak-lensing (mass) measurement
 all of which DES was designed for
 (here we use DES Year 3 data = Y3)

SZ Cluster Selection + Optical Confirmation

-> clean and deep cluster catalog

- Confirm SPT cluster candidates by measuring redshift (photo-z) and optical richness λ
- For the experts:
 - Get rid of chance associations (with SPT noise fluctuation)
 - Calibrate probability of chance association by measuring (λ , z) at random locations
 - Establish $\lambda_{min}(z)$ to achieve target purity (> 98%)

MCMF machinery (Klein+18,23; Bleem+24)

EDSU Tools 2024

SPT(SZ+pol) Cluster Sample 1,005 confirmed clusters above *z* > 0.25 over 5,200 deg²

EDSU Tools 2024

SPT Clusters and the Dark Energy Survey 3,600 deg² overlap

EDSU Tools 2024

Cluster lensing analysis Shear profiles

- Almost 700 SPT clusters (redshift 0.25–0.95) with DES Y3 shear
 - Analysis uses individual cluster shear profiles
 - Stacked for visualization purposes
 - For the experts:
 - Same source selection as in DES Y3 3x2pt
 - Same photo-*z* and shear calibrations
 - Radial range: $0.5 < r [h^{-1}Mpc] < 3.2 / (1 + z)$ (avoid cluster centers, stay in 1-halo term regime)
- 39 high-redshift clusters (redshift 0.6 1.7) with the Hubble Space Telescope Schrabback+18, Schrabback, Bocquet+21, Zohren, Schrabback, Bocquet+22

EDSU Tools 2024

Likelihood Function I Bayesian Population Modeling

Let us generate a cluster dataset!

EDSU Tools 2024

Likelihood Function Bayesian Population Modeling

Let us generate a cluster dataset!

Differential multi-observable cluster abundance

$$\frac{d^4 N(\boldsymbol{p})}{d\xi \, d\lambda \, d\boldsymbol{g}_{\mathrm{t}} \, dz} = \int \dots \int dM \, d\zeta \, d\tilde{\lambda} \, dM_{\mathrm{WL}} d\Omega_{\mathrm{s}} P(\xi \,|\, \zeta) P(\lambda \,|\, \tilde{\lambda}) P(\boldsymbol{g}_{\mathrm{t}} \,|\, M_{\mathrm{WL}}) P(\zeta, \lambda, M_{\mathrm{WL}} \,|\, M, z, \boldsymbol{p}) \frac{d^2 N(\boldsymbol{p})}{dM \, dV} \frac{d^2 V(z)}{dz \, dz}$$

$$\text{marginalize over}$$

$$\text{latent variables}$$

EDSU Tools 2024

Likelihood Function II Poisson likelihood function: $\mathscr{L}(k \text{ events } | \text{ rate } \mu) \propto \mu^k e^{-\mu} \Rightarrow \ln \mathscr{L} = k \ln(\mu) - \mu$

EDSU Tools 2024

$\ln \mathscr{L}(\boldsymbol{p}) = \sum_{i} \ln \frac{d^4 N(\boldsymbol{p})}{d\xi \, d\lambda \, d\boldsymbol{g}_{\mathrm{t}} \, dz} \bigg|_{\xi = \lambda \, q = \overline{z}} - \int \dots \int d\xi \, d\lambda \, d\boldsymbol{g}_{\mathrm{t}} \, dz \, \frac{d^4 N(\boldsymbol{p})}{d\xi \, d\lambda \, d\boldsymbol{g}_{\mathrm{t}} \, dz} \Theta_{\mathrm{s}}(\xi, \lambda, z) + \mathrm{const} \, .$

Likelihood Function II

Poisson likelihood function: $\mathscr{L}(k \text{ events } | \text{ rate } \mu) \propto \mu^k e^{-\mu} \Rightarrow \ln \mathscr{L} = k \ln(\mu) - \mu$

EDSU Tools 2024

 $\ln \mathscr{L}(\boldsymbol{p}) = \sum_{i} \ln \left| \frac{d^4 N(\boldsymbol{p})}{d\xi \, d\lambda \, d\boldsymbol{g}_{\mathrm{t}} \, dz} \right|_{\xi = \lambda \, q_{\mathrm{t}} \, z} - \int \dots \int d\xi \, d\lambda \, d\boldsymbol{g}_{\mathrm{t}} \, dz \left| \frac{d^4 N(\boldsymbol{p})}{d\xi \, d\lambda \, d\boldsymbol{g}_{\mathrm{t}} \, dz} \right|_{\Theta_{\mathrm{s}}(\xi, \lambda, z)} + \mathrm{const} \, .$

Likelihood Function II Poisson likelihood function: $\mathscr{L}(k \text{ events } | \operatorname{rate} \mu) \propto \mu^k e^{-\mu} \Rightarrow \ln \mathscr{L} = k \ln(\mu) - \mu$

can be re-written as

EDSU Tools 2024

$\int \dots \int d\xi \, d\lambda \, dg_{t} \, dz \, \frac{d^{4} N(p)}{d\xi \, d\lambda \, dg_{t} \, dz} \Theta_{s}(\xi, \lambda, z) + \text{const.}$

Likelihood Function II Poisson likelihood function: $\mathscr{L}(k \text{ events } | \text{ rate } \mu) \propto \mu^k e^{-\mu} \Rightarrow \ln \mathscr{L} = k \ln(\mu) - \mu$

$$\ln \mathscr{L}(\boldsymbol{p}) = \sum_{i} \ln \frac{d^4 N(\boldsymbol{p})}{d\xi \, d\lambda \, d\boldsymbol{g}_{\mathrm{t}} \, dz} \Big|_{\xi_i, \lambda_i, g_{\mathrm{t}, i}, z_i} - \int \dots \int d\xi \, d\lambda \, d\boldsymbol{g}_{\mathrm{t}} \, dz \, \frac{d^4 N(\boldsymbol{p})}{d\xi \, d\lambda \, d\boldsymbol{g}_{\mathrm{t}} \, dz} \Theta_{\mathrm{s}}(\xi, \lambda, z) + \mathrm{const} \, .$$

can be re-written as

$$\ln \mathscr{L}(\boldsymbol{p}) = \sum_{i} \ln \int_{\lambda_{\text{cut}}}^{\infty} d\lambda \frac{d^3 N(\boldsymbol{p})}{d\xi \, d\lambda \, dz} \Big|_{\xi_i, z_i} - \int_{z_{\text{cut}}}^{\infty} dz \int_{\xi_{\text{cut}}}^{\infty} d\xi \int_{\lambda_{\text{cut}}}^{\infty} d\lambda \frac{d^3 N(\boldsymbol{p})}{d\xi \, d\lambda \, dz} + \sum_{i} \ln \left[\frac{\frac{d^4 N(\boldsymbol{p})}{d\xi \, d\lambda \, dg_t \, dz}}{\int_{\lambda_{\text{cut}}}^{\infty} d\lambda \frac{d^3 N(\boldsymbol{p})}{d\xi \, d\lambda \, dz}} \right]_{\xi_i, z_i} + \text{cornection}$$

$$\frac{\frac{d^4 N(p)}{d\xi \, d\lambda \, dg_{\rm t} \, dz}}{\int_{\lambda_{\rm cut}}^{\infty} d\lambda \, \frac{d^3 N(p)}{d\xi \, d\lambda \, dz}} = \frac{P(\lambda, g_{\rm t}, \xi, z \, | p)}{P(\lambda > \lambda_{\rm cut}, \xi, z \, | p)} \equiv P(\lambda, g_{\rm t} \, | \, \lambda > \lambda_{\rm cut}, \xi, z, p)$$

conditional "mass calibration likelihood"

EDSU Tools 2024

28

Sebastian Bocquet — LMU Munich

nst.

EDSU Tools 2024

Bocquet+24I

Pipeline Verification

using mock datasets created from the model

- Create synthetic clusters from the halo \bullet mass function using observable — mass relations
- Analyze several statistically independent mock realizations
- Pipeline recovers input values
- We correctly implemented the analysis framework!

Robustness Tests during Blind Analysis Phase All chains were blinded by applying the same unknown parameter offset

EDSU Tools 2024

30

Robustness Tests during Blind Analysis Phase All chains were blinded by applying the same unknown parameter offset

full sa	ample, fully marginalized -	•
Systematics	full sample -	•
full s	sample, HMF uncertainty -	-
Cluster sub-samples	0.25 < z _{cluster} < 1 -	•
0.25	$5 < z_{\text{cluster}} < 1, \zeta_{\text{min}} : 1 \rightarrow 2$ -	•
	0.25 < <i>z</i> _{cluster} < 1, <i>ξ</i> > 5 -	
DES lensing $0.25 < z_{cluster} < 1$, DE	S WL boost: DNF \rightarrow BPZ -	-
0.25 < <i>z</i> _{cluster} < 1, DES \	WL center: MCMF \rightarrow SPT -	_
$0.25 < z_{cluster} < 1$, DES WL r	$r_{\min}: 500 \rightarrow 800 [h^{-1} kpc]$	

EDSU Tools 2024

30

Does the model describe the data? Binned and stacked data for visualization

Does the model describe the data? Binned and stacked data for visualization

Mean recovered model (and uncertainties) from full analysis. No significant signs of problems.

EDSU Tools 2024

ACDM with massive neutrinos

- In combination with Plan

Bocquet+24II SPT clusters + WL SPT(SZ+pol) clusters Planck18 + (DES Y3 + HST) WL SPT clusters + WL + *Planck*18 ACT DR-6 lensing Planck18 TTTEEE DES Y3 3x2pt | BAO ----0.90 0.85 σ_8 0.80 0.75 [0.3 0.2 0.1 0.3 0.4 0.1 0.2 0.3 0.25 0.30 0.35 0.75 0.80 0.85 $\sum m_{v}$ [eV] Ω_{m} Ω_{m} σ_8

• Competitive constraints, especially on $S_8^{\text{opt}} \equiv \sigma_8 \left(\Omega_{\text{m}}/0.3\right)^{0.25}$

• No evidence for " S_8 tension" with Planck (1.1 σ)

nck
$$\sum m_{\nu} < 0.18 \,\mathrm{eV} \,(95 \,\% \,\mathrm{C} \,. \,\mathrm{L})$$

Sebastian Bocquet — LMU Munich

32

Tracing the Growth of Structure Phenomenological test

- Five bins in redshift with equal number of clusters
- Fit for independent amplitudes $\sigma_8(z)$
- With loose prior on Ω_m from the sound horizon at recombination θ_*
- Good agreement with ΛCDM model and *Planck* parameters from *z* = 0.25 to *z* = 1.8

Sebastian Bocquet — LMU Munich

33

Outlook select work by PhD students

Mazoun, Bocquet, Garny, Mohr, Rubira, Vogt 24

Asmaa Mazoun

Interacting dark sector models

Analysis of SPT+DES dataset ongoing (Mazoun+ in prep.)

EDSU Tools 2024

1014

 $M[h^{-1}M_{\odot}]$

1015

1016

Sophie Vogt

FIG. 1. The critical overdensity δ_{crit} for spherical collapse in f(R) gravity (Eq. (12)) for different values of $\log_{10} |f_{R0}|$ at collapse redshift $z_c = 0$ in colored solid lines. The dashed black line represents δ_{crit} in a corresponding GR cosmology (Eq. (13)).

f(R) and nDGP models

Analysis of SPT+DES dataset done (Vogt+ in prep.)

Outlook: Joint Constraints SPT Cluster Abundance + DES 3x2 pt

- Joint analysis
 - Cosmological covariance
 - Shared (lensing) systematics
 - Addressed w/ Chun-Hao To, Elisabeth Krause, Sebastian Grandis
- Expect powerful constraints on z < 2 large-scale structure
- Ideal complement to high-redshift CMB measurements by
 Planck

Outlook: Joint Constraints SPT Cluster Abundance + DES 3x2 pt

- Joint analysis
 - Cosmological covariance
 - Shared (lensing) systematics
 - Addressed w/ Chun-Hao To, Elisabeth Krause, Sebastian Grandis
- Expect powerful constraints on z < 2 large-scale structure
- Ideal complement to high-redshift CMB measurements by
 Planck

The South Pole Telescope (SPT)

10-meter sub-mm quality wavelength telescope
90, 150, 220 GHz and
1.6, 1.2, 1.0 arcmin resolution

2007: SPT-SZ

960 detectors 90,150,220 GHz

2012: SPTpol

1600 detectors 90,150 GHz +Polarization

2017: SPT-3G

~15,200 detectors 90,150,220 GHz +Polarization

The 10,000 deg² SPT-3G Survey(s)

Survey	Area	Years observed	Noise level (T)				
	$[deg^2]$		[μ K-arcmin]				
			95 GHz	150 GHz	220 GHz	Coadded	
SPT-3G Main	1500	2019-2023, 2025-2026	2.5	2.1	7.6	1.6	
SPT-3G Summer	2600	2019-2023	8.5	9.0	31	6.1	
SPT-3G Wide	6000	2024	14	12	42	8.8	

Wide is still 2—3 times deeper than SPT-SZ!

Slide from Lindsey Bleem

The 10,000 deg² SPT-3G Survey(s)

Survey	Area	Years observed	Noise level (T)				
	$[deg^2]$		[μK -arcmin]				
			95 GHz	150 GHz	220 GHz	Coadded	
SPT-3G Main	1500	2019-2023, 2025-2026	2.5	2.1	7.6	1.6	
SPT-3G Summer	2600	2019-2023	8.5	9.0	31	6.1	
SPT-3G Wide	6000	2024	14	12	42	8.8	

Wide is still 2-3 times deeper than SPT-SZ!

Slide from Lindsey Bleem

Summary

- Cluster abundance as a cosmological probe
- SZ-selection + weak-lensing mass calibration • = excellent control over systematics
- Latest analysis of SPT (SZ+pol) clusters with • DES Y3 + HST lensing is competitive and compatible with other probes
- Next few years will be spectacular (SPT-3G, • advACT, SO, eROSITA, DES Y6, KiDS, HSC, Euclid, LSST, CMB-S4, etc.)

Image credit: CTIO/NOIRLab/NSF/AURA/D. Munizaga

Cluster Lensing Model Grandis, Bocquet+21

- Simple model shear(Mass) based on NFW profile
- Biased and noisy estimator (e.g., Becker&Kravtsov 11)
- Solution: Introduce latent variable M_{WL} and establish $M_{WL}-M_{halo}$ relation such that shear(M_{WL}) is unbiased
- establish $M_{WL} M_{halo}$ relation using hydrodynamic simulations to capture baryonic effects
- Incorporate all known sources of uncertainty in lensing measurements into $M_{\rm WL}-M_{\rm halo}$

Calibrate mean $M_{WL} - M_{halo}$ relation and intrinsic scatter $\langle \ln\left(\frac{M_{WL}}{M_0}\right) \rangle = b_{WL}(z) + b_M \ln\left(\frac{M_{200c}}{M_0}\right)$ $\ln \sigma_{\ln M_{WL}} = \frac{1}{2} \left[s_{WL}(z) + s_M \ln\left(\frac{M_{200c}}{M_0}\right) \right]$

EDSU Tools 2024

- 5,259 clusters over 12,791 deg2
- Low-redshift lenses
- DES Y3 lensing SNR = 65

R.A. [deg]

eRASS1 cluster abundance >5,000 X-ray selected clusters

- Mass calibration driven by DES Year 3 lensing data (SNR 65) (Grandis, Ghirardini, <u>Bocquet+24</u>)
- eROSITA largely follows our approach
 - Individual cluster likelihoods
 - *M*wl-*M*halo relation
 - DES Year 3 lensing analysis (but they also use KiDS and HSC data)
- Simultaneous constraints on Ω_m , σ_8 , w, $\sum m_v$
- Cluster cosmology using ICM-selected clusters works!

EDSU Tools 2024

Ghirardini+24

EDSU Tools 2024

 SPT-3G data gets to ~Planck depth on 1500d field with a ~week of data.

Observe 1500d field every ~2 days for 6 years

The SPT-3G SZ Catalog

- First catalog from 2019-2020 data

EDSU Tools 2024

JSObrin

Slide adapted from Lindsey Bleem Sebastian Bocquet — LMU Munich

High S/N (>30σ) detection of CMB cluster lensing!

S.Bocquet+L. Bleem

0.3

 Ω_{m}

0.7

Prakrut Chaubal

0.2