Clustering Wedges: An Alternative Approach to Measuring H(z) and $D_A(z)$

Eyal Kazin

In collaboration with: Tamara **Davis**, Chris **Blake**, Ariel **Sánchez**







What you will get from this talk

- For the non-expert:
 - How we **measure** the **geometry** of the Universe with galaxy clustering
 In other words: the Alcock-Paczynski test on the Baryonic
 - Acoustic Feature to constrain H(Z), $D_A(Z)$
- For the **expert**:
 - practical aspects of **binning** your correlation functions: Multipoles, **Wedges**, RR(µ)



Kazi





For usage of data, mock catalogues:

Marc Manera, Cameron McBride, The Sloan Digital Sky Survey, The WiggleZ Dark Energy Survey

The WiggleZ Group





SWINBURNE UNIVERSITY OF TECHNOLOGY

Large Scale Structure Workshop, Trieste, August 1st 2012





For usage of data, mock catalogues:

Marc Manera, Cameron McBride, The Sloan Digital Sky Survey, The WiggleZ Dark Energy Survey

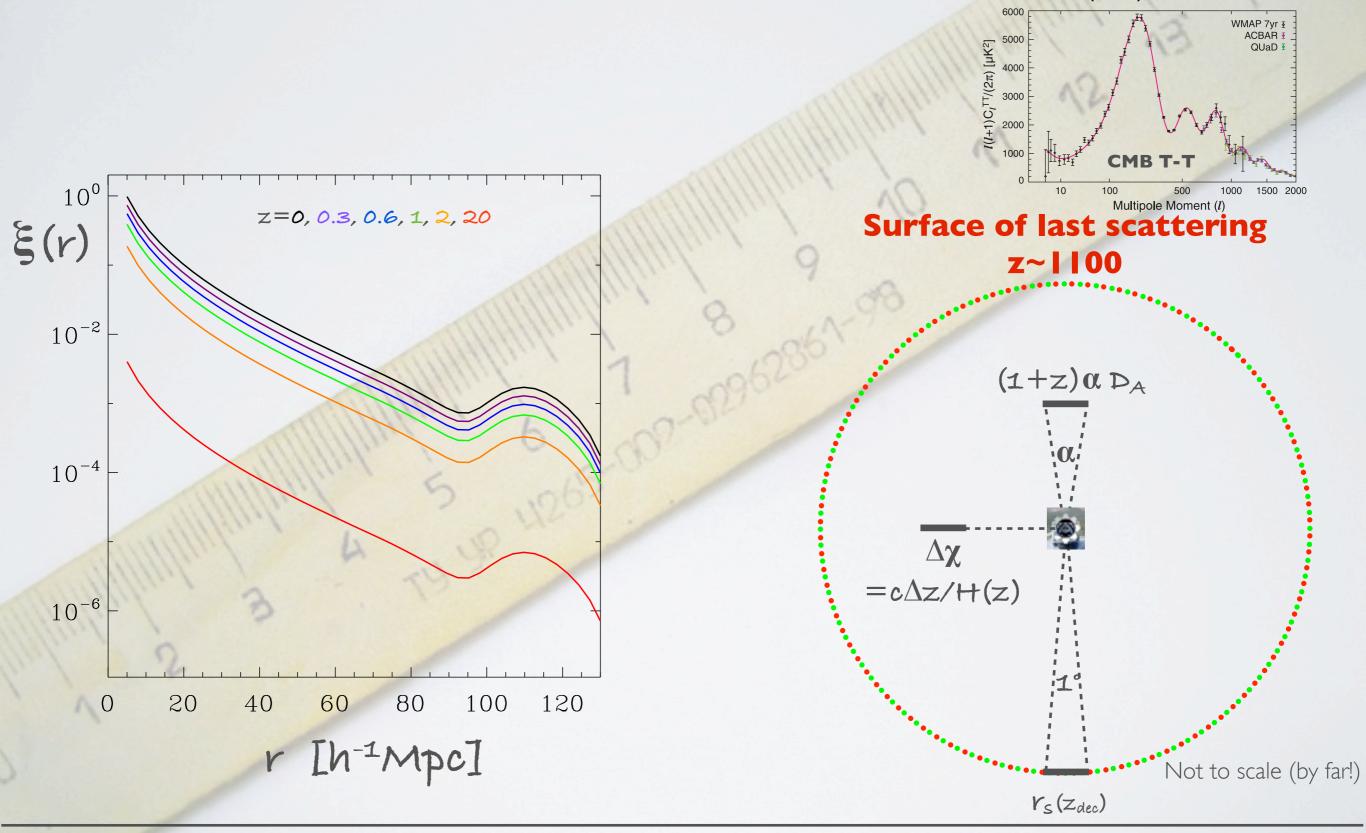
The Sloan Digital Sky Survey



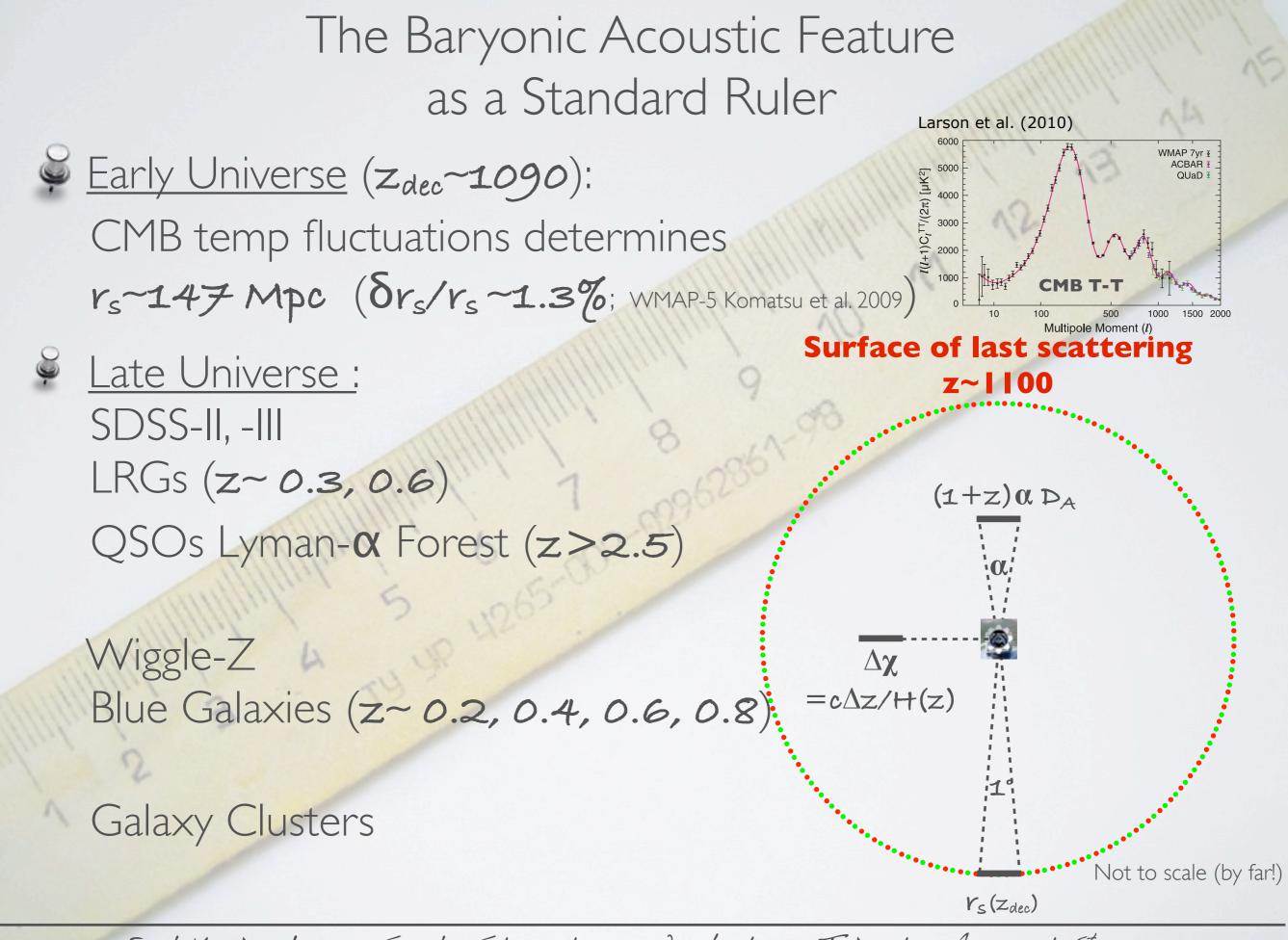


Large Scale Structure Workshop, Trieste, August 1st 2012

The Baryonic Acoustic Feature as a Standard Ruler

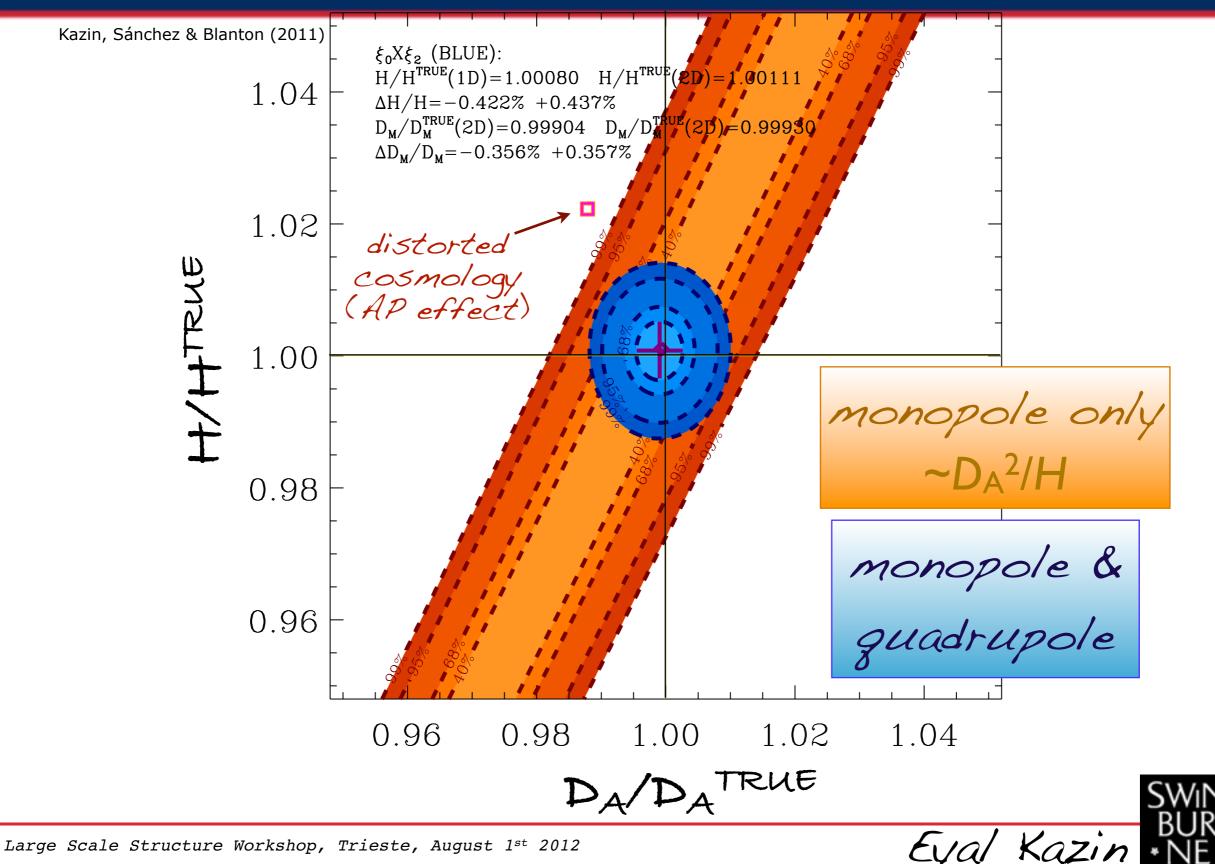


Eyal Kazin, Large-Scale Structure Workshop, Trieste, August 1st 2012



Eyal Kazin, Large-Scale Structure Workshop, Trieste, August 1st 2012

CAASTRO EXAMPLE OF EXCELLENCE The Importance of Anisotropic Clustering



IVERSITY O





I am not going to show BOSS clustering Wedges results today. (but stay tuned ..)

Most of the plots here are from mock catalogues.



Eval Kazir



- z-distortions in practice: a brief practical recap
- There is information in the Hexadecapole $\xi_4(s)$
- In with the new (basis): Clustering Wedges $\xi(\Delta\mu,s)$
- Time Permitting: $N_{RR}(\mu) \neq constant$



Eval Kazi



Redshift Distortions: Dynamical and Geometrical

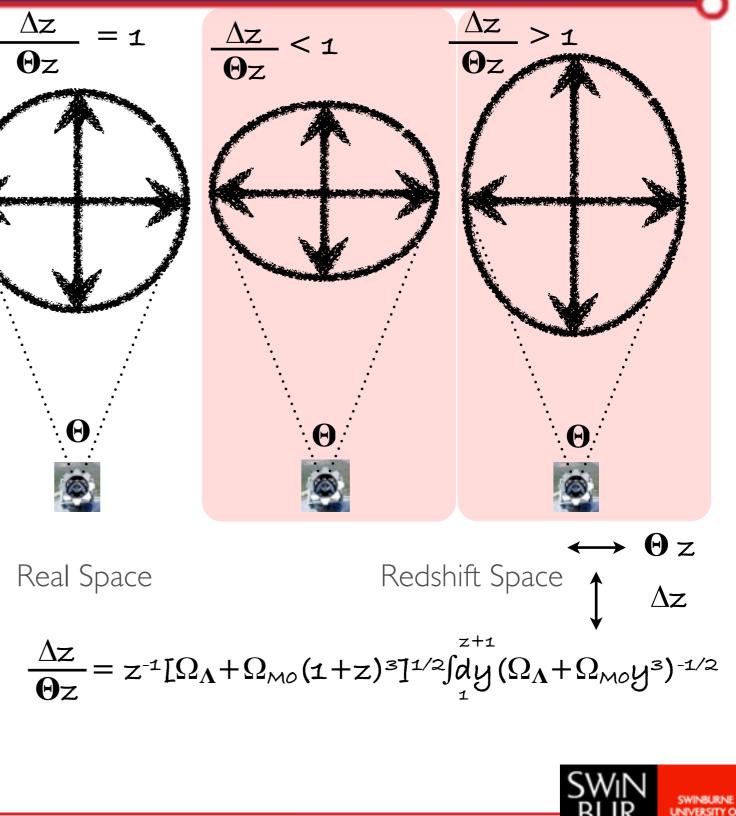
Dynamical: squashing (kaiser 1987), Finger of God comoving distance Zobs $\chi(z) = c \int dz'$ 0 H(z', Ω) Geometrical: AP effect (Alcock&Paczynski 1979) Large Scale Structure Workshop, Trieste, August 1st 2012



The Alcock-Paczynski Effect

In the **anisotropic Baryonic Acoustic Feature H×D**_A $r_{I}=c\Delta z/H(z)$ $r_{2=}(I+z)D_{A}(z)\Theta$ $r_1 = r_2$ $H \times D_A = c \Delta z / (1 + z) / \Theta$ In the **isotropic Baryonic Acoustic Feature** D_A^2/H $d^3s = \alpha d^3s^D$ $\alpha = \left(\frac{H^{\mathcal{D}}}{H}\right)^{1/3} \left(\frac{D_{\mathrm{A}}}{D_{\star}^{\mathcal{D}}}\right)^{2/3}$

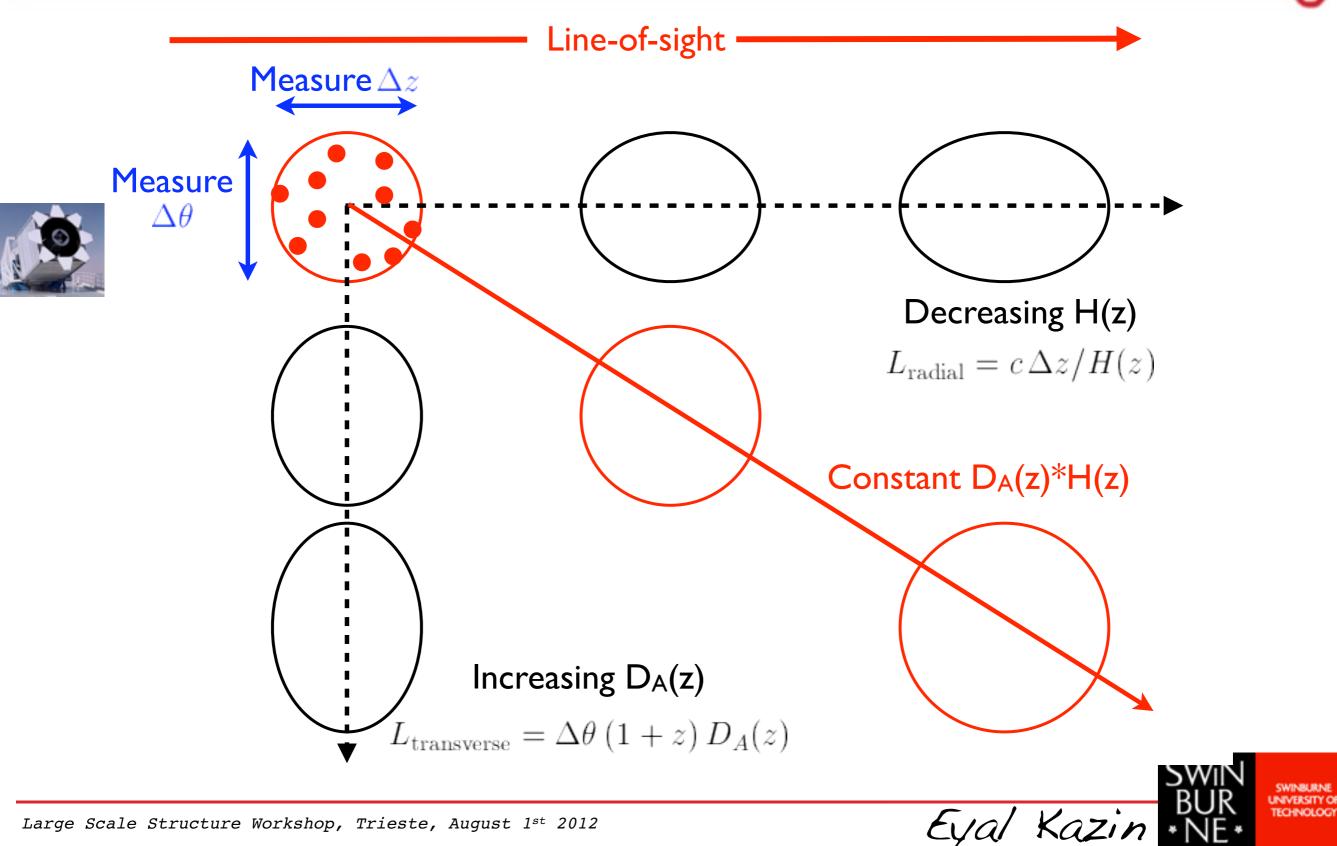
Large Scale Structure Workshop, Trieste, August 1st 2012



azin



The Alcock-Paczynski Effect Plot Credit: Chris Blake





Timely Jargon

Z-distortions: General term for both types

of distortions. Not solely Dynamical!

Dynamical:

Squashing (kaiser 1987), Non-linear etc..

Finger of God (velocity dispersion effect)

Geometrical:

Alcock-Paczynski effect (Alcock&Paczynski 1979)



Eyal Kazin

LasDamas Mock Simulations

public mocks: http://lss.phy.vanderbilt.edu/lasdamas/

McBride et al.; in prep.

Andereas Berlind Michael Busha Jeff Gardner

Cameron McBride

Román Scoccimarro Frank van den Bosch Risa Wechsler



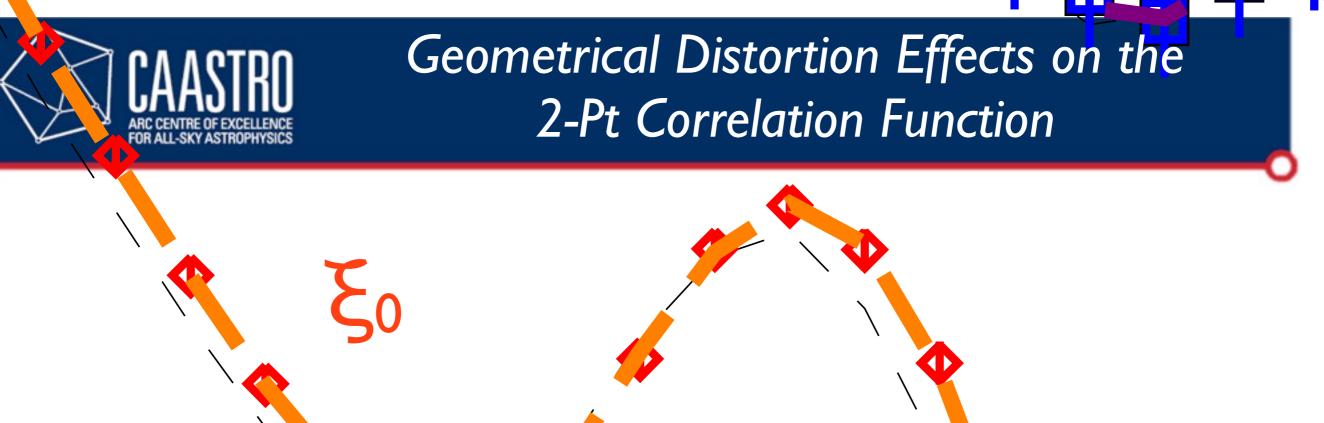
- LArge Suite of DArk MAtter Sims
- Emphasize on many observational effects

Results in most realistic uncertainties of clustering of the SDSS-II LRGs

E(s~ BA feature scale)

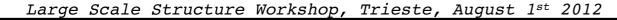
-			SUNC						
, Mock #1	Mock #2	Mock #3	Mock #4	Mock #5	Mock #41	Mock #42	Mock #43	Mock #44	Mock #4
° ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~	• • • • • • • • • • • • • • • • • • • •	**************************************	**************************************	********	** _{***}	►	* ************	• • • • • • • • • • • • • • • • • • • •	° ° ° ° ° ° ° ° ° ° ° ° ° ° ° ° ° ° °
Mock #6	Mock #7	Mock #8	Mock #9	Mock #10	Mock #46	Mock #47	Mock #48	Mock #49	Mock #5
	•••	**************************************	° • • • • • •	°	••••	•		••	°
**************************************				* ************************************	1	× • • • • • • • • • • • • • • • • • • •	**************************************	**** *********************************	
• MOCK #11	Mock #12	Mock #13	Mock #14 . ◇◇	Mock #15	Mock #51	Mock #52	Mock #53	MOCK #54	Mock #5
******	۰ ۰ ۰	****************	****	* ***********	*****	× × × × × × × × × × × × × × × × × × ×	* *******	× • • • • • • • • • • • • • • • • • • •	**************************************
Mock #16	₀ Mock #17 .	Mock #18	Mock #19	Mock #20	Mock #56	mock #01	Mock #58	Mock #59	Mock #6
* * * * * * * * * * * * * * * * * * *	****	• • • • • • • • • • • • • • • • • • • •	· · · · · · · · · · · · · · · · · · ·	**************************************	******	• • • • • • • • • • • • • • • • • • • •	× × × × × × × × × × × × × × × × × × ×	*	• •••••••
Mock #21	Mock #22	Mock #23	Mock #24	Mock #25	Mock #61	Mock #62	Mock #63	Mock #64	Mock #6
**************************************	* *****************	* * * * * * * *	* • • • * * * • • • • • • • • • • • • •	**************************************	* • • • • • • • • • • • • • • • • • • •	* * * * * * * * * * * * * * * * * * * *	· · · · · · · · · · · · · · · · · · ·	· · · · · · · · · · · · · · · · · · ·	**************************************
Mock #26	• Mock #27	ו••• Mock #28	Mock #29	Mock #30	Mock #66	Mock #67	Mock #68	Mock #69	Mock #
° • • • • • • • • • • • • • • • • • • •	° • • • • •	•	\$ •	***********		• •	° • • • • • • • • • • • • • • • • • • •	*.	· · · · · · · · · · · · · · · · · · ·
**** Mock #31	**************************************	× * * * * * * * * * * * * * * * * * * *	**************************************	Mock #35	* <u>*</u> **********************************	* <u>************************************</u>	Mock #73	***** ***** Mock #74	Mock #
	•	Mock #33	•	þ			• •	•	
* • • • • • • • • • • • • • • • • • • •	******	*****************	*****************	******	Nock #76	°° _{°°°°°°°°°°°°°°°°}	***** *****	° °°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°	[*] °°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°
Mock #36	Mock #37	Mock #38	. Mock #39	1 MOCK #40	Mock #76	Mock #77	Mock #78	Mock #79	Mock #
° • • • • • • • • • • • • • • • • • • •	***************	° *****	× × × × × × × × × × × × × × × × × × ×	* *********	*****	••••••••••••••••••••••••••••••••••••••	*********	*****	* *******
Mock #81	Mock #82	Mock #83	Mock #84	Mock #85	Mock #121	1	Mock #123	Mock #124	
••. Î	****	× × × × × × × × × × × × × × × × × × ×	° 	**************************************	° ° • • • • •	• •	• •	** _* **	•
[^]	Mock #87	Mock #88	[°] ************************************	Mock #90	Mock #126		[*] °°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°°	^{**} **********************************	**************************************
Þ		۰ ۲	¢	•		******			P
	**************************************	*************		* ***************	**************************************	*********	*****************	*****	° ° °
Mock #91	Mock #92	Mock #93	. Mock #94	Mock #95	Mock #131	Ļ.	Mock #133	Mock #134	Mock #
° • • • • • • • • • • • • • • • • • • •	**************	****************	*****	*****	* *******	*****************	**************************************	************	*************
Mock #96	Mock #97	Mock #98	Mock #99	Mock #100	Mock #136	1	Mock #138	Mock #139	Mock #
•••••	° • • • • • •	• •••••	^ο	* * * * * * * * * * * * *	******		* * * * * * * * *	• • • • • • • • • •	**************************************
Mock #101	********** Mock #102	Mock #103	Mock #104	Mock #105	Mock #141	**************************************	Mock #143	Mock #144	Mock #
*••**••	° • • • • • • • • • • • • • • • • • • •	°	* ****	* *****	* * * * * * * * * * * * * * * * * * *	• •	• • •	• • • • • • •	••••
Mock #106	Mock #107	Mock #108	Mock #109	Mock #110	Mock #146	Mock #147	Mock #148	Mock #149	Mock #
·	•	° ° ° ° ° .	° • • • •	 	•	° • • • •	•		¢
• • • • • • • • • • • • • • • • • • •	**************************************	Mock #113	**************************************	000 ⁰ 00000000000000000000000000000000	***********************	***	**************************************	**************************************	° °
MOCK #111	• • •	• • • • • • • • • • • • • • • • • • •	MOCK #114.	Mock #115.	Mock #151	Mock #152	Mock #153	Mock #154 .	Mock #
× * * * * * * * * * * * * * * * * * * *	****	*****	** _{**} ***	****	**************************************	***********************	**************************************	******	*******
					1	1			
Mock #116	Mock #117	Mock #118	Mock #119.	Mock #120	Mock #156	Mock #157	 Mock #158 . ♦ 	Mock #159	Mock #1

S [h-1Mpc]



Template (here I use the true mock signal)
``data'' (here I use mock signal affected by AP)

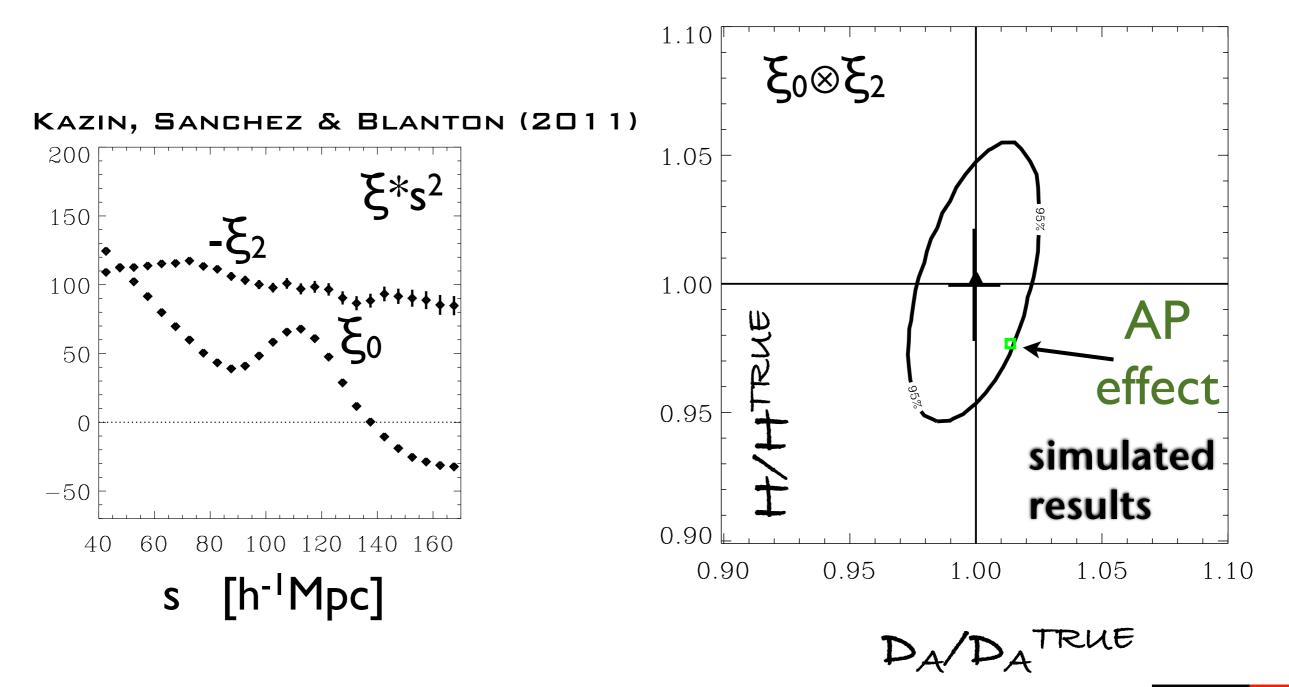
• fit (here I fit Template to ``data" varying H and \hat{D}_{A}



CAASTRO ARC CENTRE OF EXCELLENCE FOR ALL-SKY ASTROPHYSICS	Geometrical Distortion Effects on the Clustering Multipoles
	For the mathematically inclined: Padmanabhan & White (2008)
	$-\xi_2 \sim D_A H$
	$-\frac{1}{2} - \frac{1}{2} - 1$
Large Scale Structure Workshop, Tries	ste, August 1st 2012 Eyal Kazin * NE*



Extra Information from the Hexadecapole

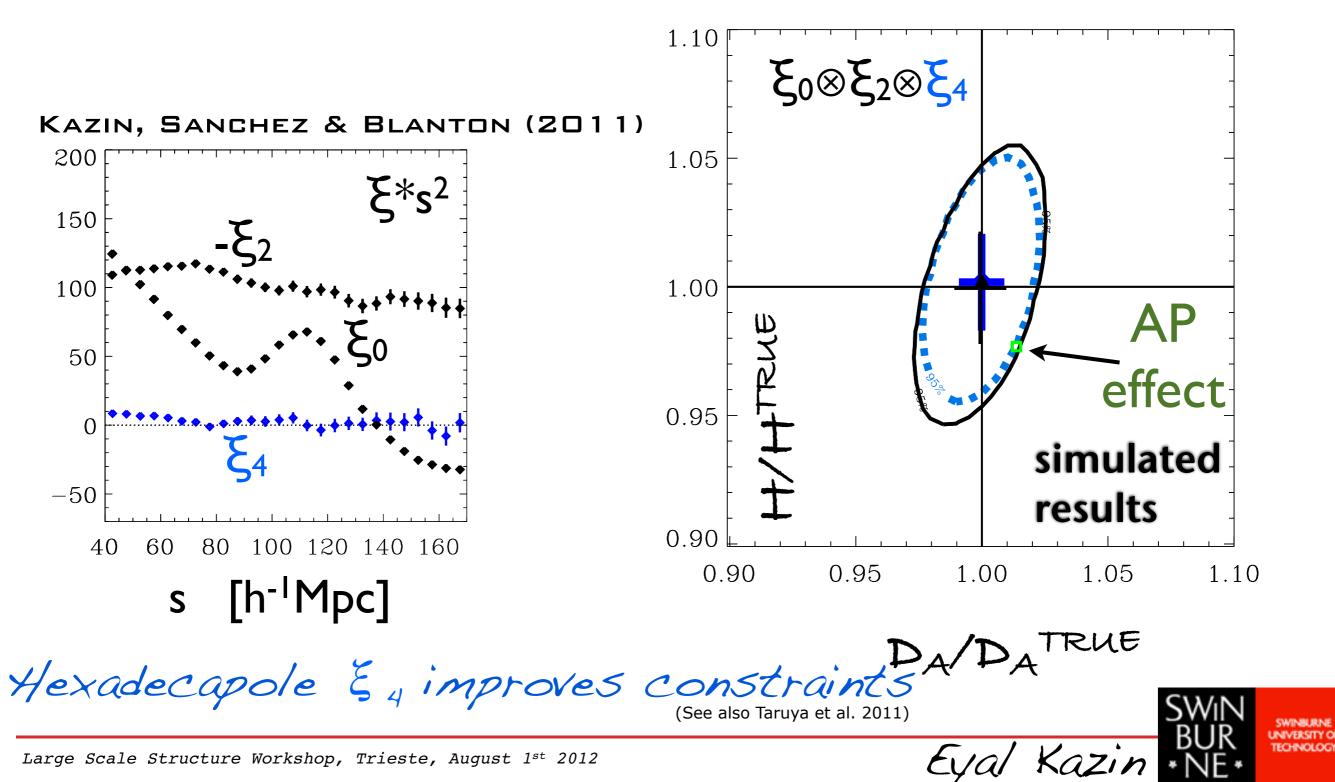




Kazi



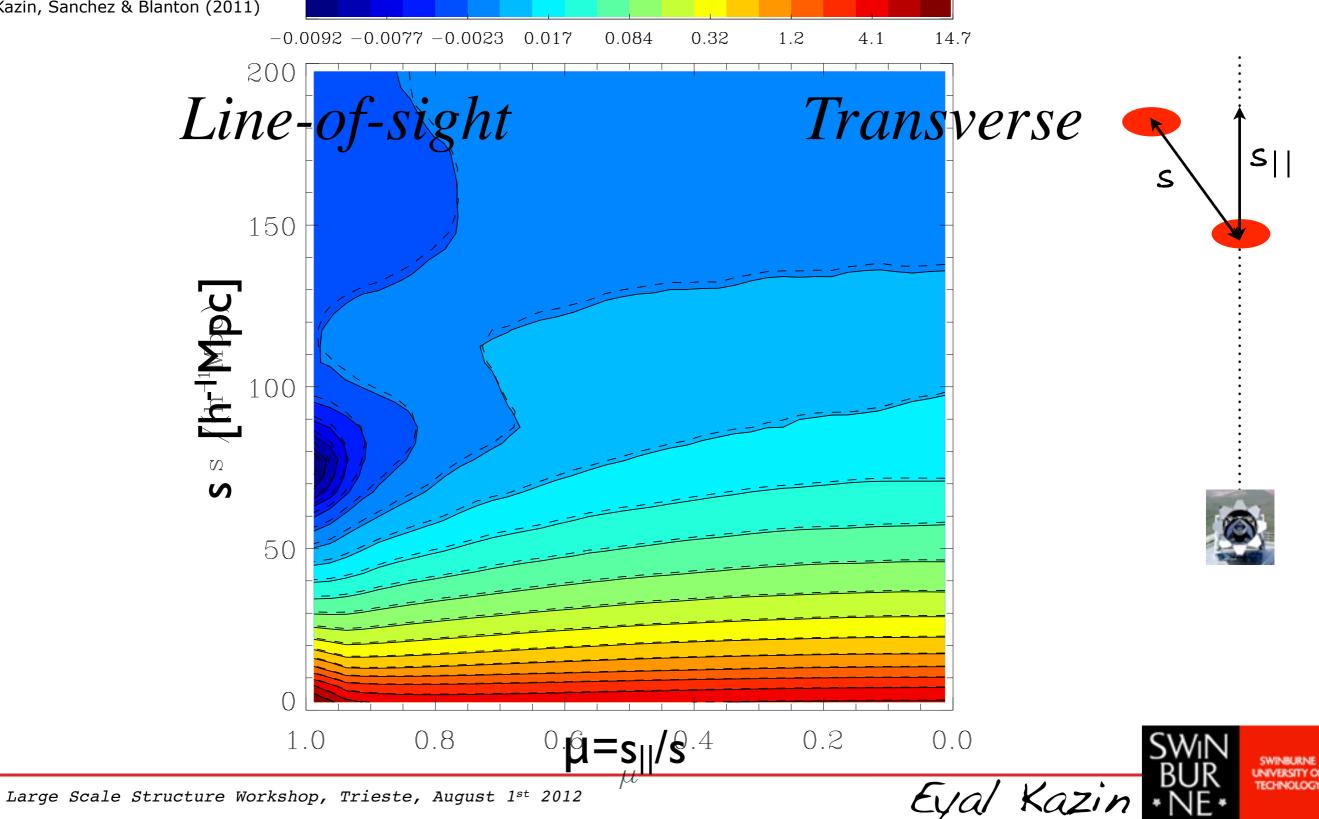
Extra Information from the Hexadecapole





The 2D Clustering Plane $\xi(\mu,s)$

Kazin, Sanchez & Blanton (2011)



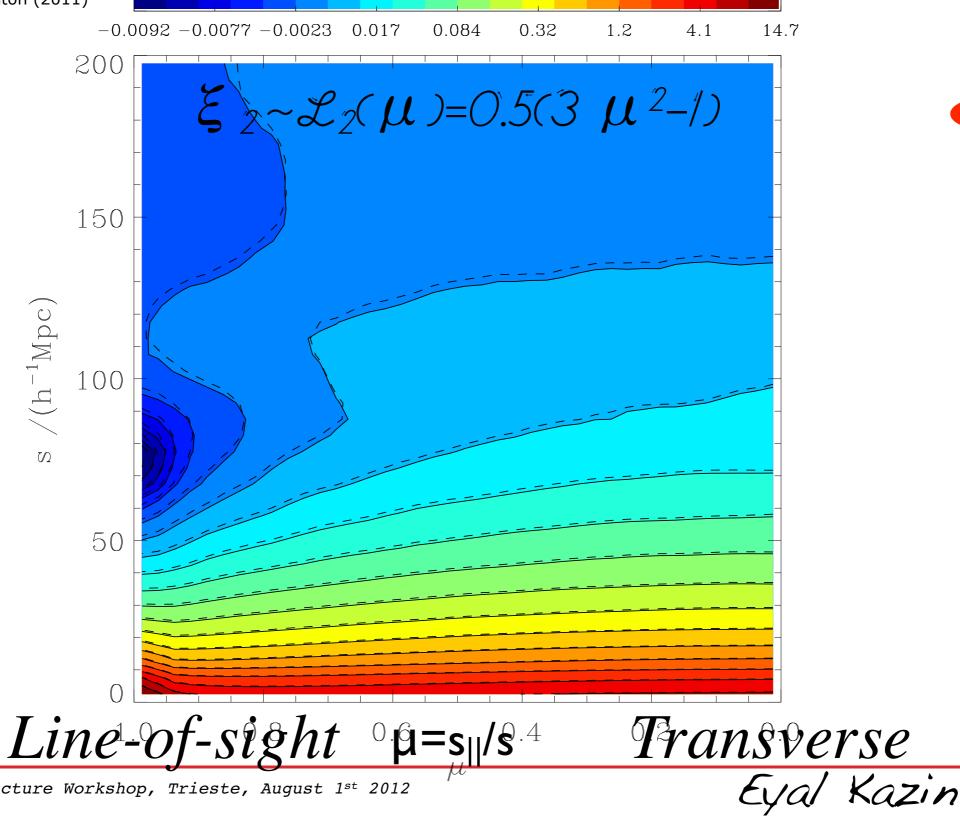


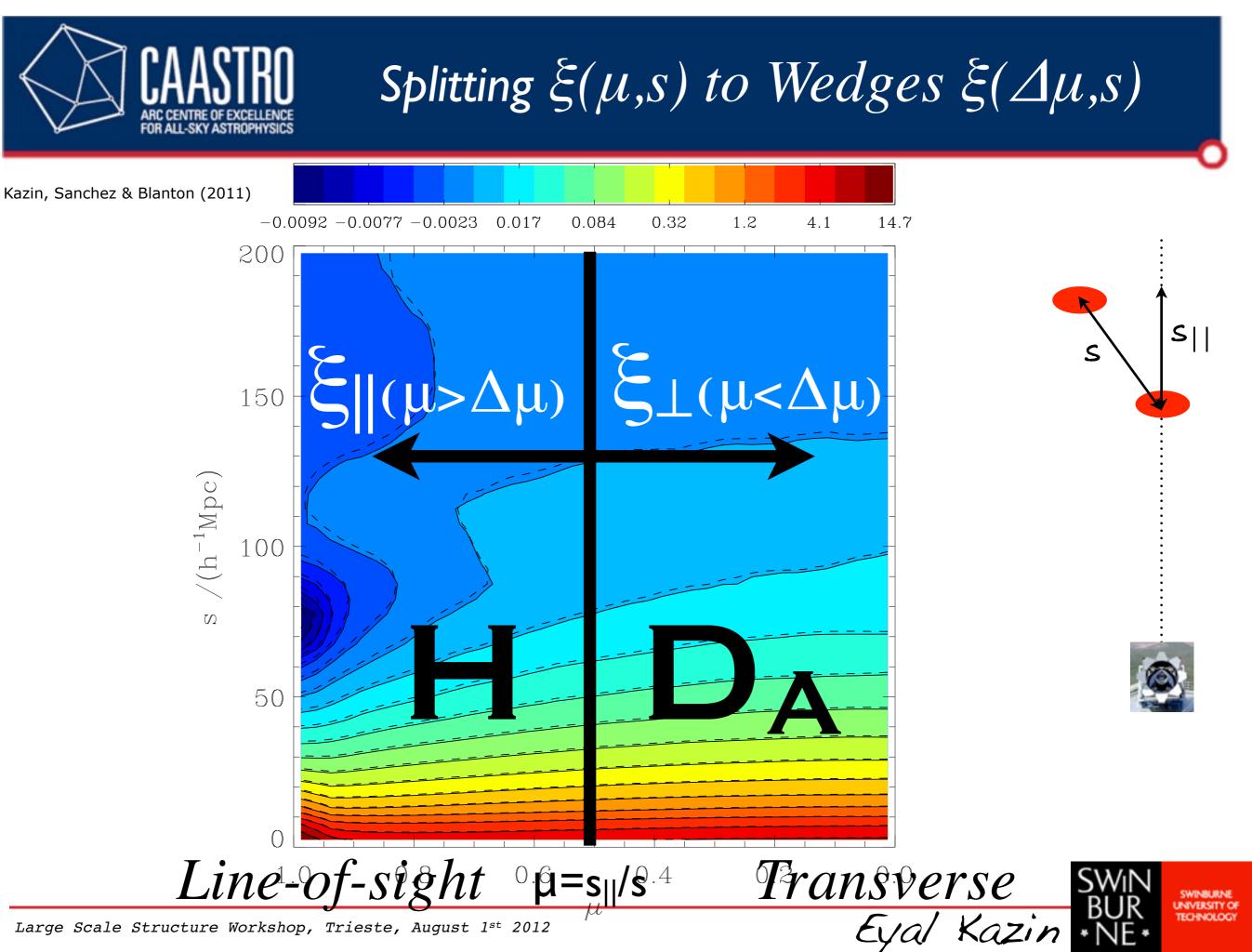
Splitting $\xi(\mu,s)$ to Multipoles $\xi_{\ell}(s)$

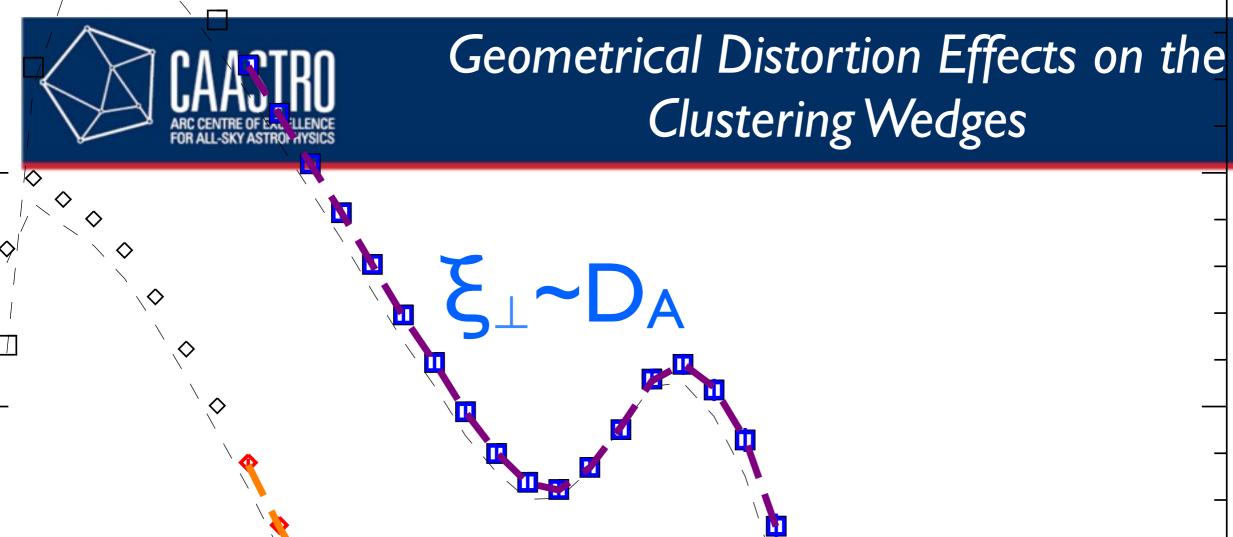
S

NIVERSITY O

Kazin, Sanchez & Blanton (2011)







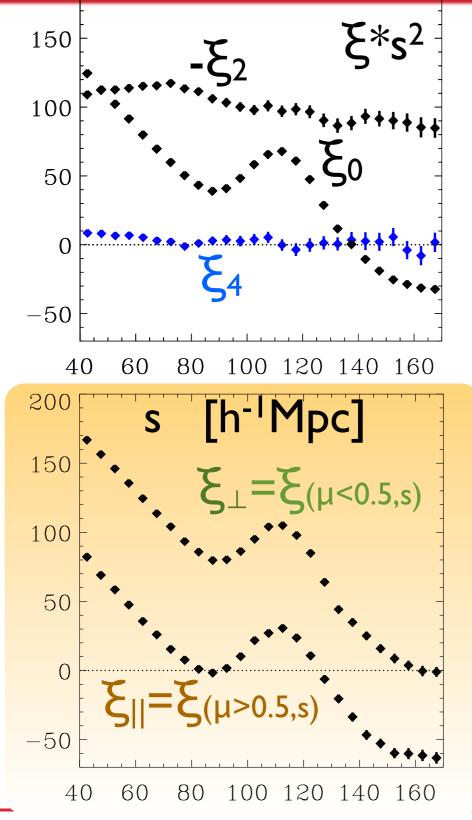
Kazin

Eyal

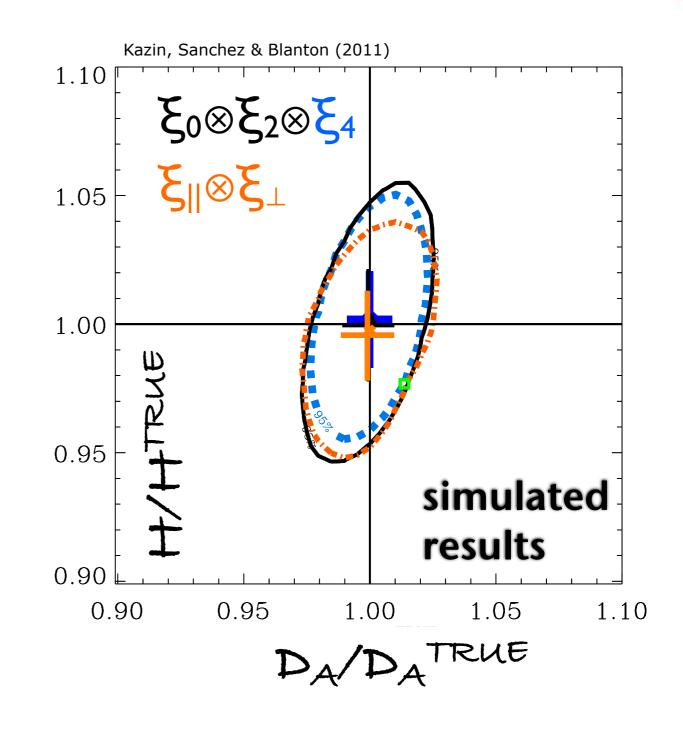
╌┤┍



Wedges H, D_A Performance

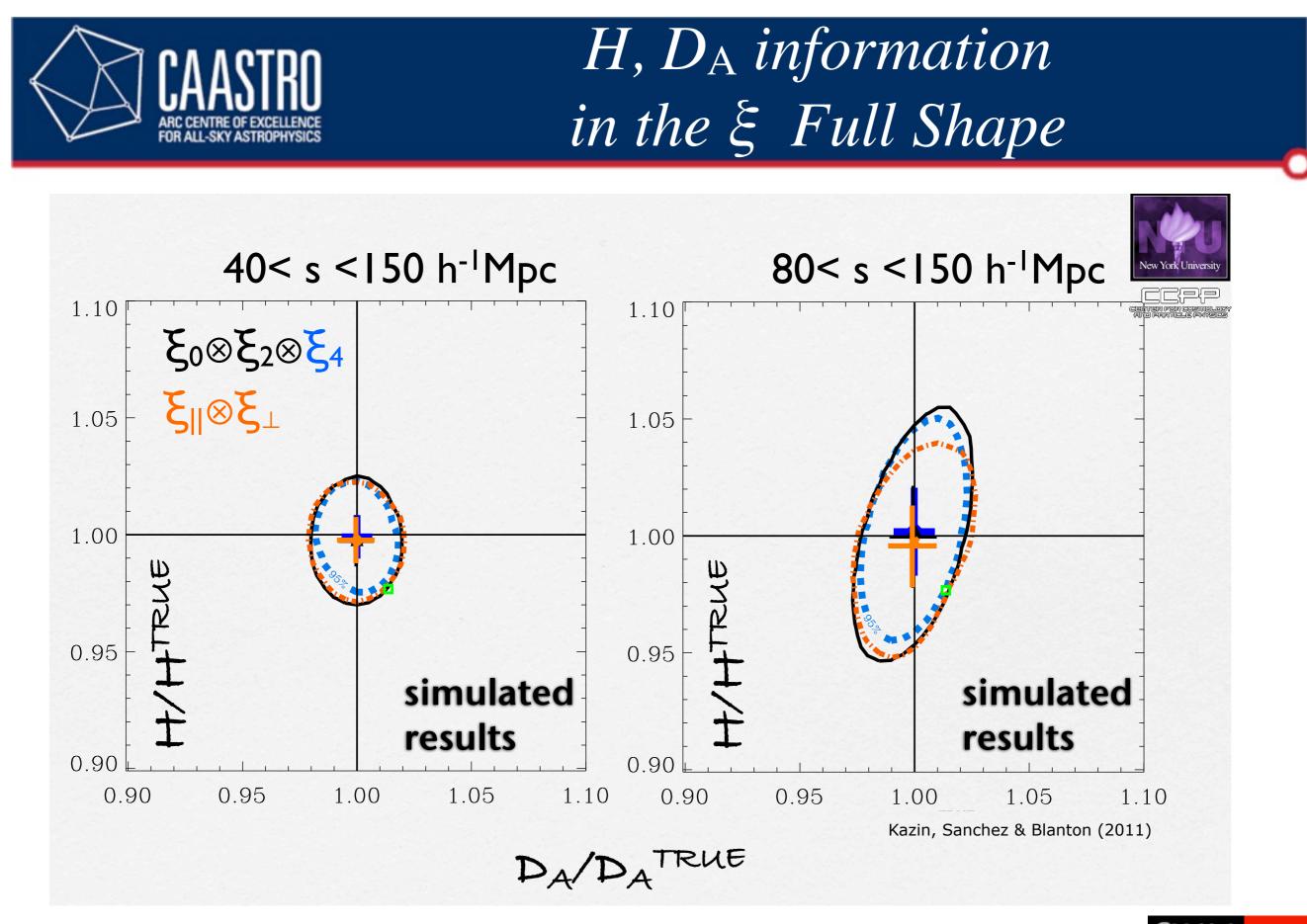


Large Scale Structure Workshop, Trieste, August 1st 2012



Eyal Kazir

SWIN BUR * NE *





Eyal Kazir



Wedges: Basic Equations

Definition:
$$\xi(\Delta\mu, s) \equiv \frac{\int_{\mu_{\min}}^{\mu_{\max}} \xi(\mu', s) d\mu'}{\int_{\mu_{\min}}^{\mu_{\max}} d\mu'}$$

Basis Transform From Multipoles:

$$\xi(\Delta\mu, s) = \xi_0 + \frac{1}{2} \left(\frac{\mu_{\text{max}}^3 - \mu_{\text{min}}^3}{\mu_{\text{max}} - \mu_{\text{min}}} - 1 \right) \xi_2$$

For $\Delta\mu=0.5$:

 $\begin{pmatrix} \xi_{\parallel} \\ \xi_{\perp} \end{pmatrix} = \begin{pmatrix} 1 & \frac{3}{8} \\ 1 & -\frac{3}{8} \end{pmatrix} \begin{pmatrix} \xi_{0} \\ \xi_{2} \end{pmatrix}$ $\underset{\text{ugust 1st 2012}}{\text{Eyal Kazin Supp}}$ Large Scale Structure Workshop, Trieste, August 1st 2012



$$\xi_{||}^{\mathcal{D}}(s) = \xi_{||} \left(\frac{H^{\mathcal{D}}}{H}s\right) + \mathcal{C}_{||}(\epsilon),$$

$$\xi_{\perp}^{\mathcal{D}}(s) = \xi_{\perp} \left(\frac{D_{\mathrm{A}}}{D_{\mathrm{A}}^{\mathcal{D}}} s \right) + \mathcal{C}_{\perp}(\epsilon),$$

Inter-mixing terms (not a pretty sight ...):

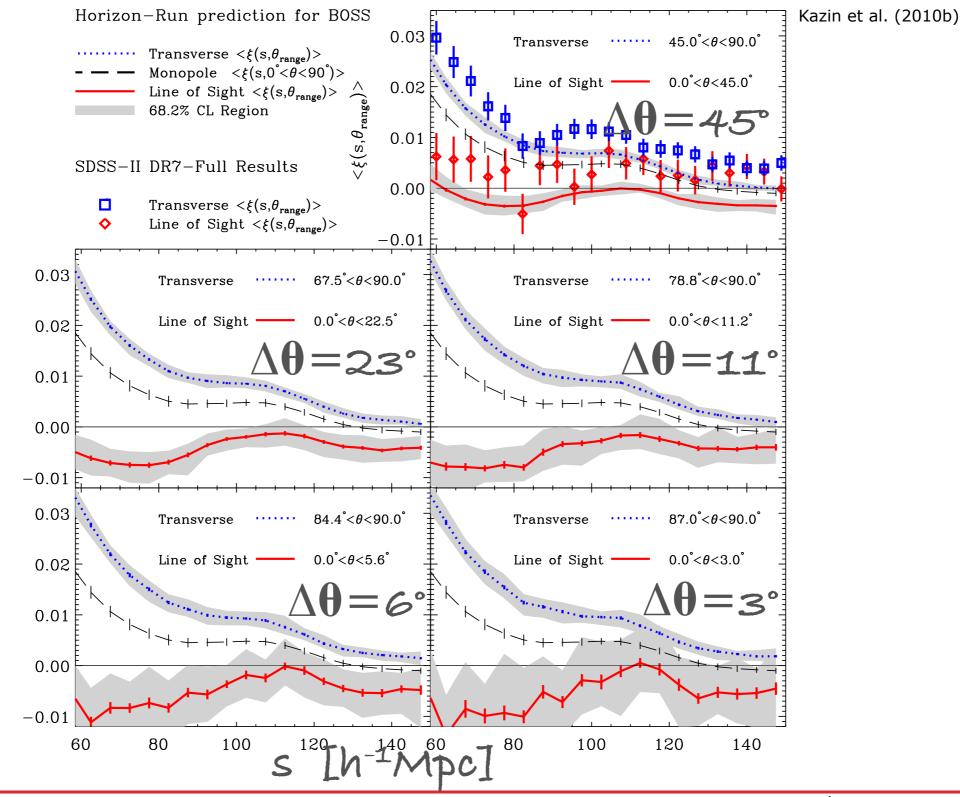
$$\mathcal{C}_{||}(\epsilon,\alpha) = \epsilon \left(-\frac{5}{4} \frac{d\xi_0(s)}{d\ln(s)} - \frac{19}{140} \frac{d\xi_2(s)}{d\ln(s)} + \frac{213}{140} \xi_2(\alpha s) \right)$$
$$\mathcal{C}_{\perp}(\epsilon,\alpha) = \epsilon \left(\frac{1}{4} \frac{d\xi_0(s)}{d\ln(s)} - \frac{53}{280} \frac{d\xi_2(s)}{d\ln(s)} + \frac{123}{140} \xi_2(\alpha s) \right)$$

NIVERSITY O

Eyal Kazi



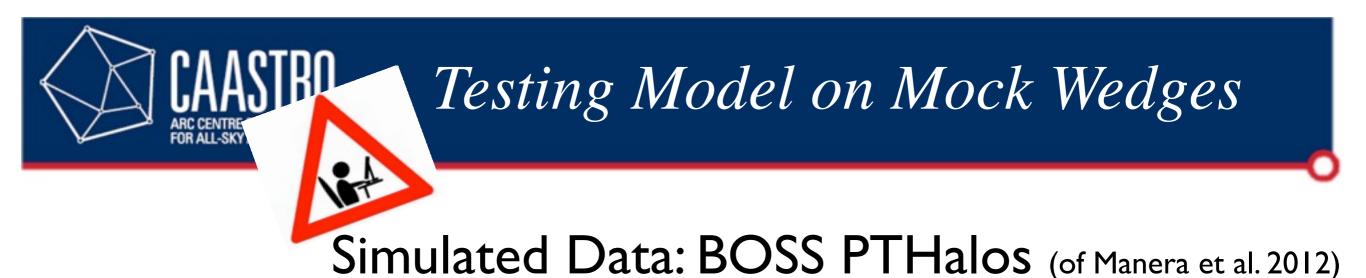
BOSS 2014 Predicted Wedges $\xi(\Delta\mu,s)$



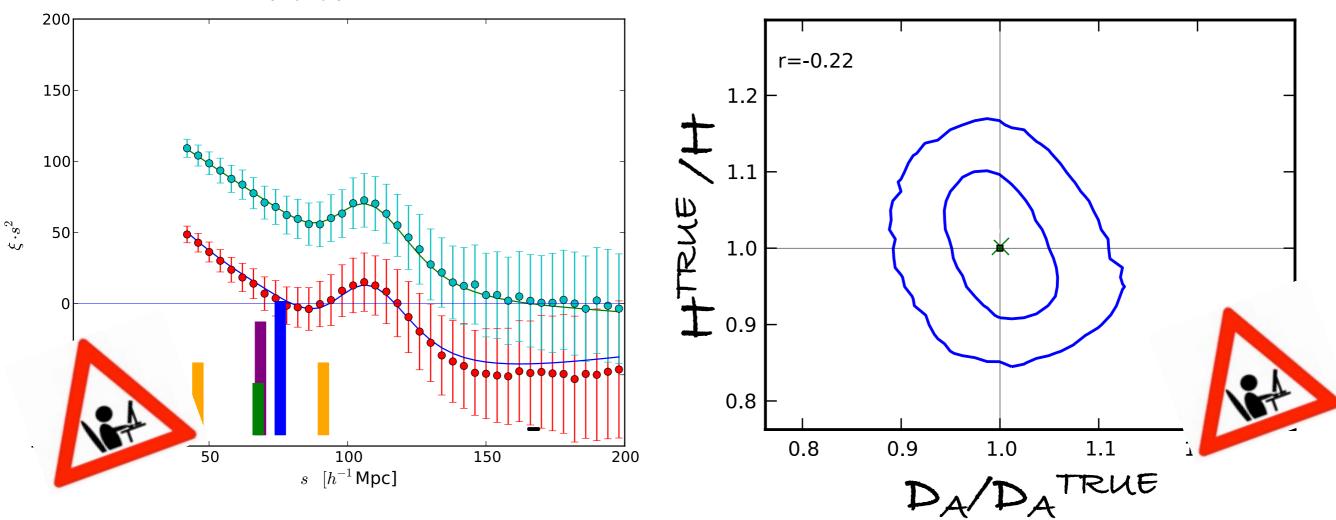
Large Scale Structure Workshop, Trieste, August 1st 2012

Eyal Kazir





Kazin, Sánchez & the SDSS (in prep.)



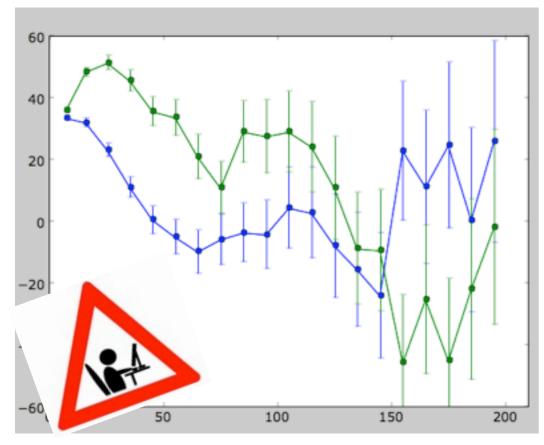
NIVERSITY O

Eyal Kazir



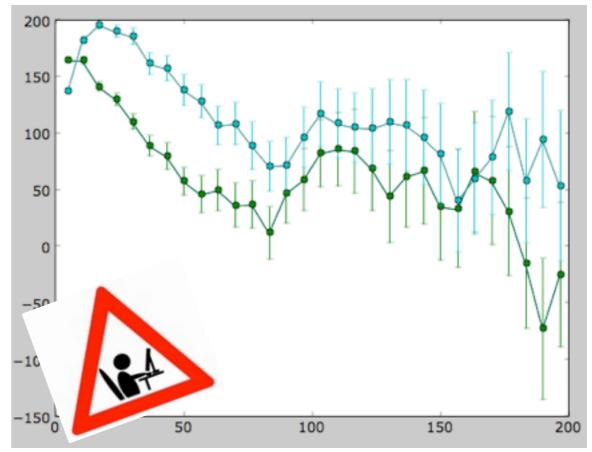
Clustering Wedges in the Data

Davis, Kazin & the WiggleZ (in prep.)



WiggleZ (0.2<z<1) (bias ~ 1)

SDSS-II LRGs (0.16<z<0.44) (bias ~ 2.2)



Eyal Kazir

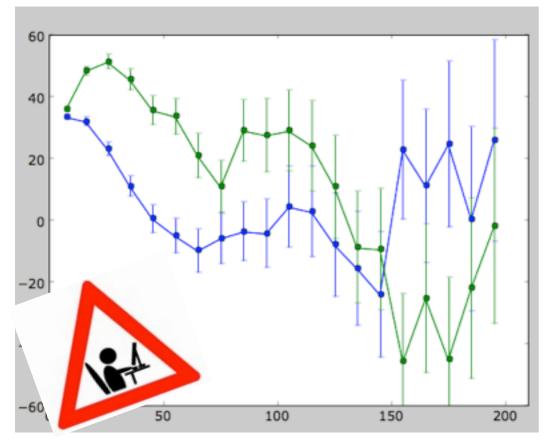
Sánchez & Kazin (in prep.)





Clustering Wedges in the Data

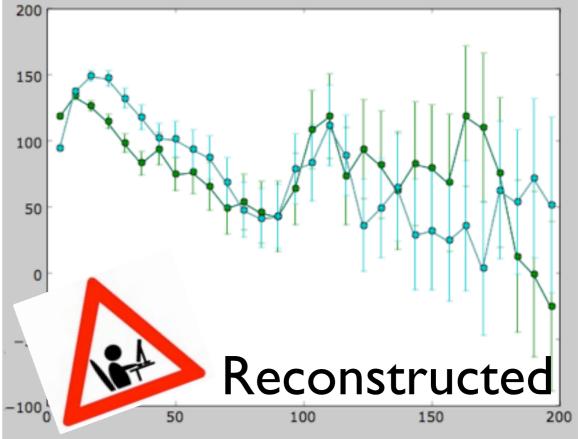
Davis, Kazin & the WiggleZ (in prep.)



WiggleZ (0.2<z<1) (bias ~ 1)

SDSS-II LRGs (0.16<z<0.44) (bias ~ 2.2)

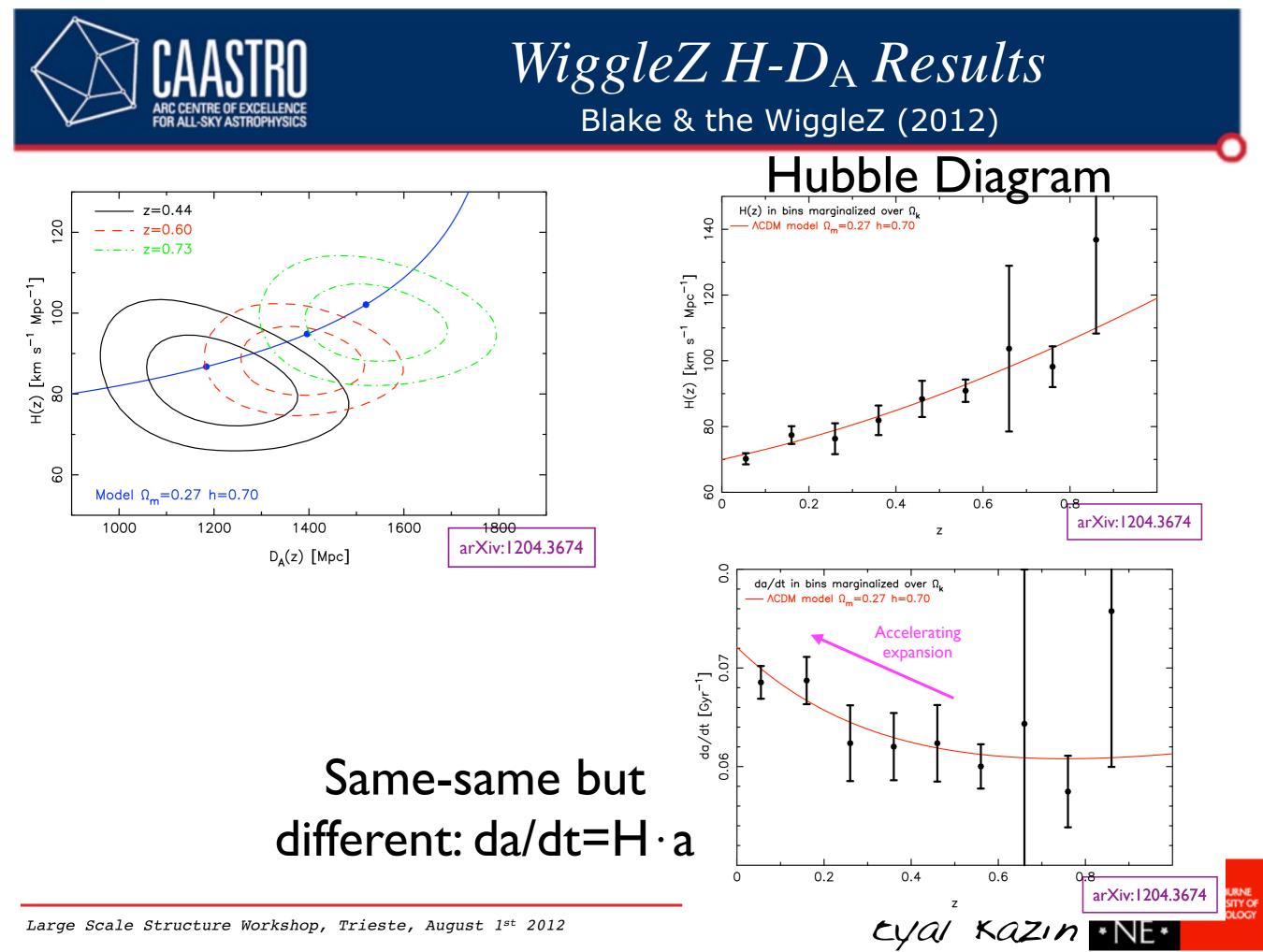
Credit for reconstructed data: Nikhil Padmanabhan



Eyal Kazir

Sánchez & Kazin (in prep.)







- ξ(Δμ,s) wedges more practical than than 2D ξ(μ,s)
 plane because:
 - Higher S/N
 - Much cheaper (=easier) covariance matrix
- Compared to multipoles $\xi_{\ell}(s)$ in constraining H, D_{A}, f :
 - Is one basis better than the other?
 - Are two peaks more useful than one?
 - to be continued ...







slides contain t might not be individuals periodic box.

The following information that appropriate for that live inside a

Eval Kazi,







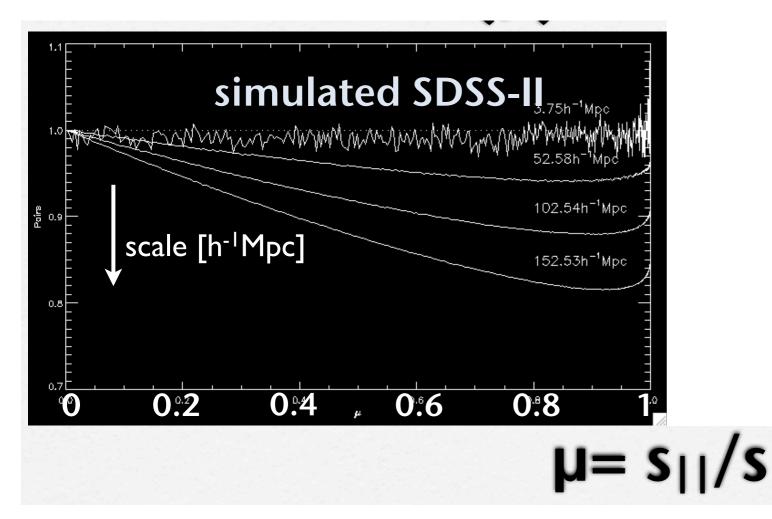
The following slides contain information that might not be appropriate for individuals that live inside a periodic box.



Eval Kazi







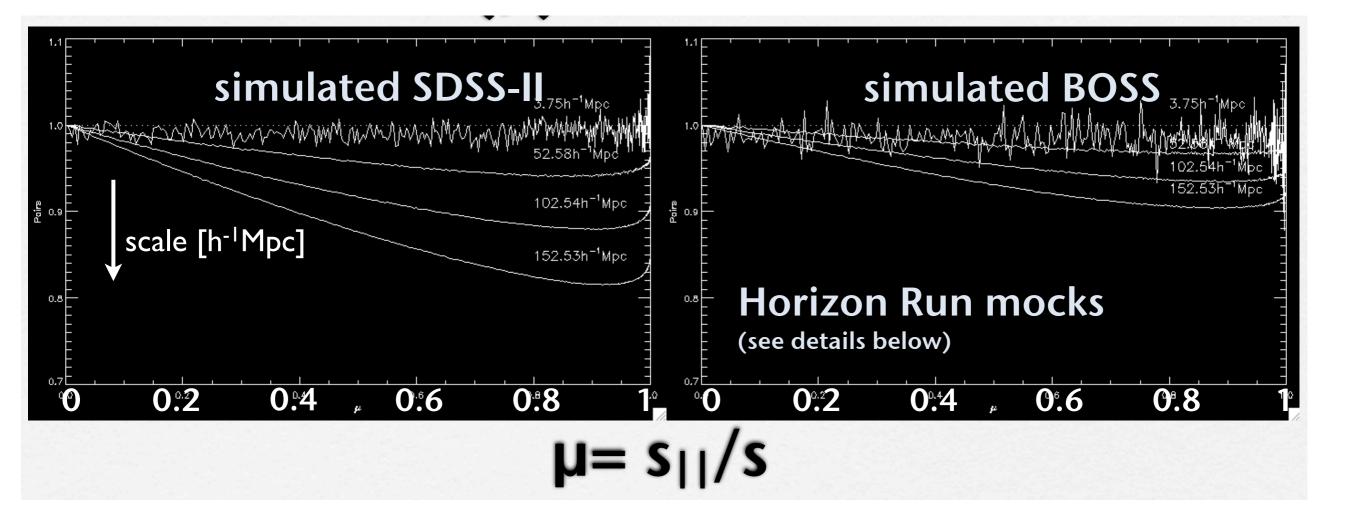
LasDamas (SDSS-II geometry) 0.16<z<0.44 volume limited 8000 deg², SDSS sky-coverage



Eyal Kazir







LasDamas (SDSS-II geometry) 0.16<z<0.44 volume limited 8000 deg², SDSS sky-coverage Horizon Run (~BOSSish 2014) 0.16 < z < 0.6 volume limited $10,300 \text{ deg}^2(\pi \text{ str})$ BOX

Kazir



SWINBURNE UNIVERSITY OF TECHNOLOGY



BOSS Advertisement

Now in your nearest browser!

More than 800,000 Spectra in over 3,000 deg²

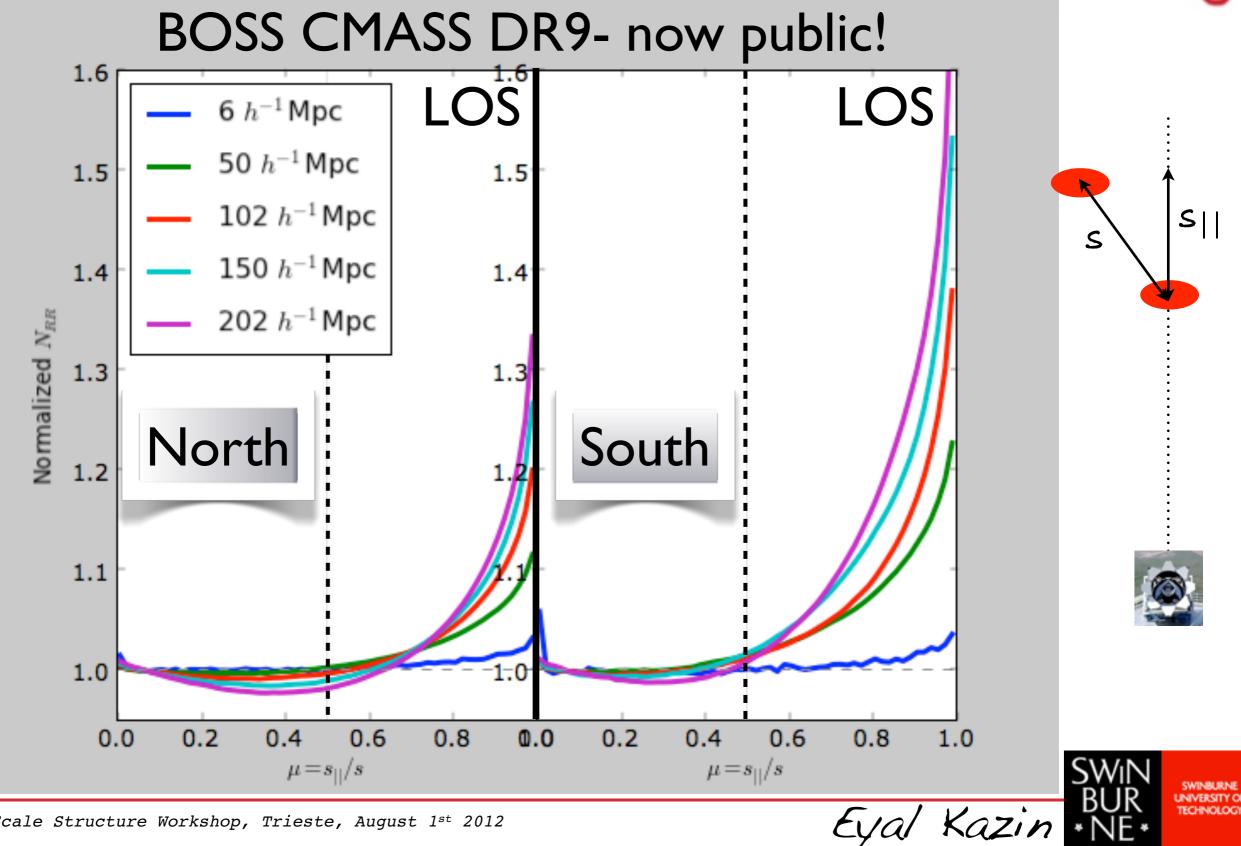
http://sdss3.org/dr9/



Eval Kazi



 $N_{RR}(\mu) \neq constant$





ξ Estimators: Direct vs Integrated

$$\mathcal{P}_{0} = 1$$

$$\mathcal{P}_{2} = \frac{1}{2} \left(3\mu^{2} - 1 \right)$$

[up to (2*l*+1)/2]

Eyal Kazin

$$\xi_{\ell} \equiv \int_{-1}^{+1} \mathrm{d}\mu \mathcal{P}_{\ell}(\mu) \xi(\mu, s) = \int_{-1}^{+1} \mathrm{d}\mu \mathcal{P}_{\ell}(\mu) \frac{DD(\mu, s) - RR(\mu, s)}{RR(\mu, s)}$$





ξ Estimators: Direct vs Integrated

$$\mathcal{P}_0 = 1$$

 $\mathcal{P}_2 = \frac{1}{2} (3\mu^2 - 1)$
[up to (2 ℓ +1)/2]

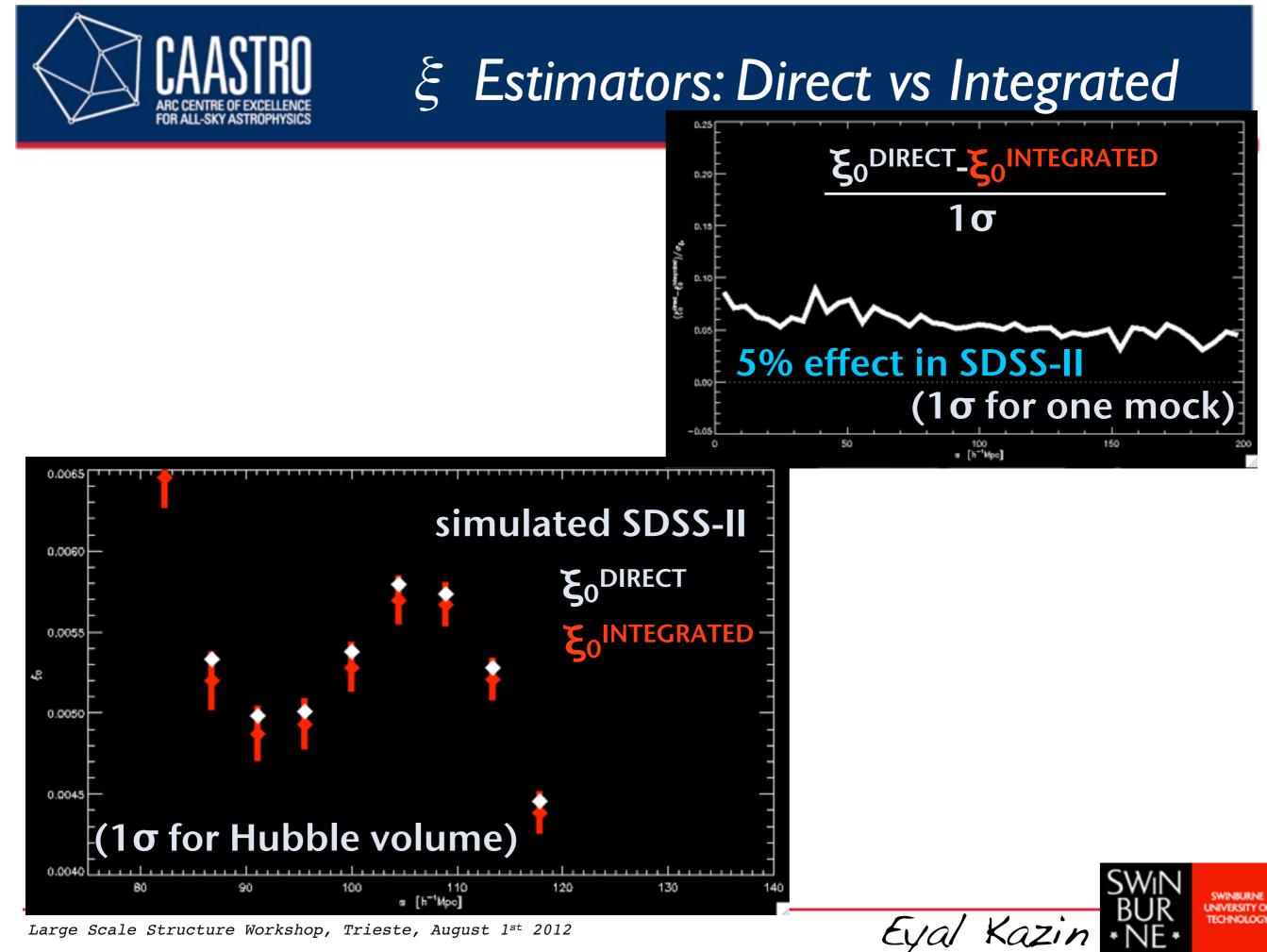
Eyal Kazin

$$\xi_{\ell} \equiv \int_{-1}^{+1} \mathrm{d}\mu \mathcal{P}_{\ell}(\mu) \xi(\mu, s) = \int_{-1}^{+1} \mathrm{d}\mu \mathcal{P}_{\ell}(\mu) \frac{DD(\mu, s) - RR(\mu, s)}{RR(\mu, s)}$$

$$\xi_0(s) = \frac{DD(s) - RR(s)}{RR(s)} \neq \int_{-1}^{+1} d\mu \frac{DD(\mu, s) - RR(\mu, s)}{RR(\mu, s)}$$

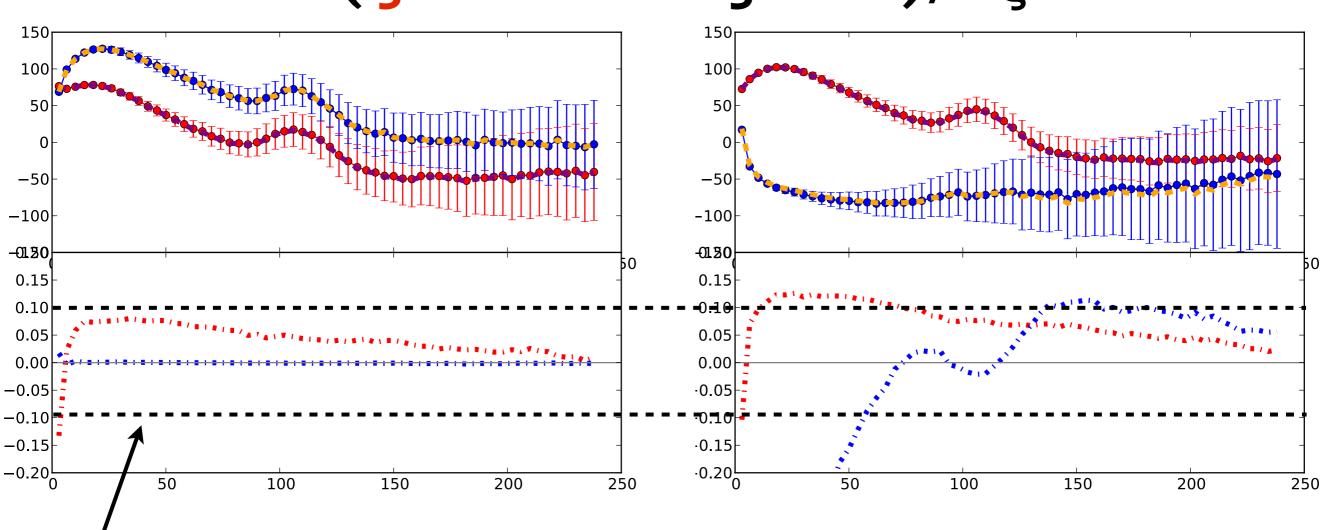
$$\xi_{0}(s) = \frac{DD(s) - RR(s)}{RR(s)} = \int_{-1}^{+1} d\mu \frac{DD(\mu, s) - RR(\mu, s)}{RR(\mu, s)} \cdot \frac{RR(\mu, s)}{RR(s)}$$

SWINBURNE UNIVERSITY OF TECHNOLOGY



ξ Estimators: Direct vs Integrated Investigating 610 BOSS Mocks

 $(\xi^{\text{INTEGRATED}} - \xi^{\text{DIRECT}})/\sigma_{\xi}$



10% of 1 σξ line: indicating wedges are less sensitive to method of estimator



Kazin

SWINBURNE UNIVERSITY OF TECHNOLOGY



 ξ Estimators: Direct vs Integrated

But, don't you degrade



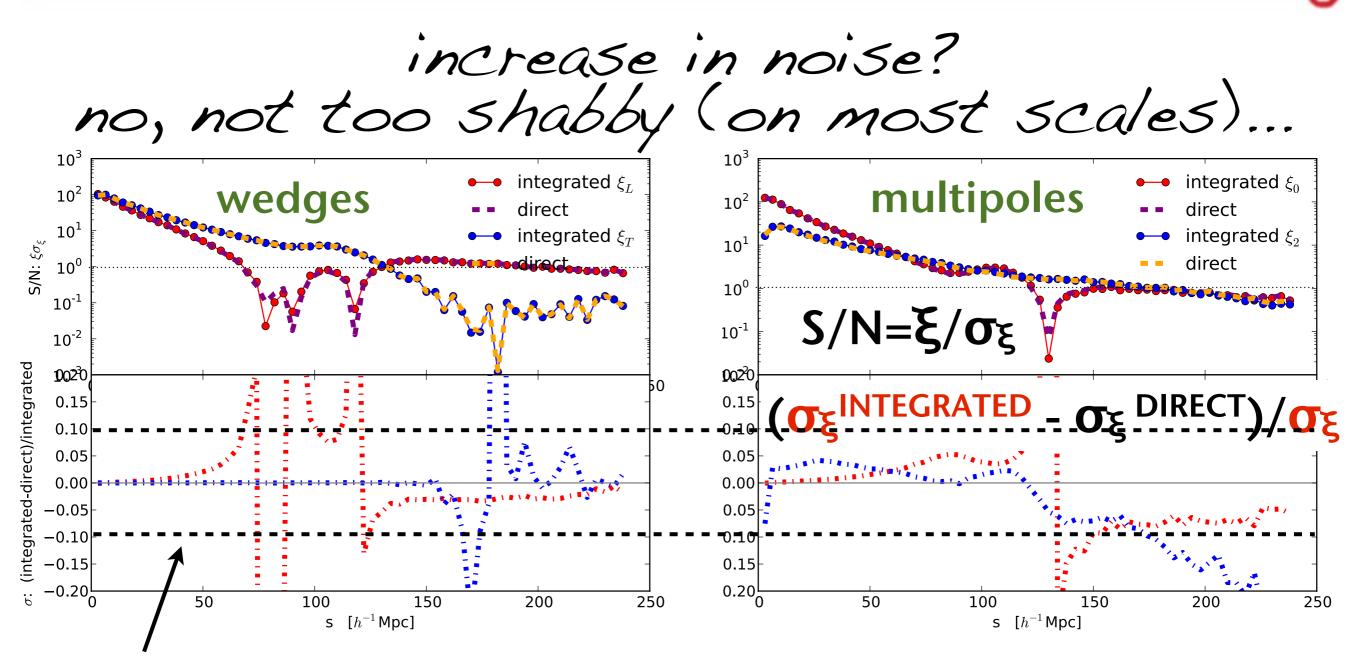
when integrating

over noisy data bins?



Eyal Kazir

ξ Estimators: Direct vs Integrated Investigating 610 BOSS Mocks



10% of I σξ line: indicating wedges are less sensitive to method of estimator

Large Scale Structure Workshop, Trieste, August 1st 2012

Eyal Kazin



SWINBURNE UNIVERSITY OF TECHNOLOGY



- $\xi(\Delta\mu,s)$ wedges more practical than than 2D $\xi(\mu,s)$ plane because:
 - Higher S/N
 - Much cheaper (=easier) covariance matrix
- Comparing ξ(Δμ) wedges to ξ_ℓ(s) multipoles in constraining H,D_A,f





- ξ(Δμ,s) wedges more practical than than 2D ξ(μ,s)
 plane because:
 - Higher S/N
 - Much cheaper (=easier) covariance matrix
- Comparing ξ(Δµ) wedges to ξ_ℓ(s) multipoles in constraining H,D_A,f







- ξ(Δμ,s) wedges more practical than than 2D ξ(μ,s)
 plane because:
 - Higher S/N
 - Much cheaper (=easier) covariance matrix









- ξ(Δμ,s) wedges more practical than than 2D ξ(μ,s)
 plane because:
 - Higher S/N
 - Much cheaper (=easier) covariance matrix
- Compared to multipoles $\xi_{\ell}(s)$ in constraining H, D_{A}, f :
 - Is one basis better than the other?
 - Are two strong peaks more useful than one?
 - to be continued ...





