

# DESI and the missing-observation problem

Davide Bianchi

ICG Portsmouth

Sexten 02/07/2018

# Based on:

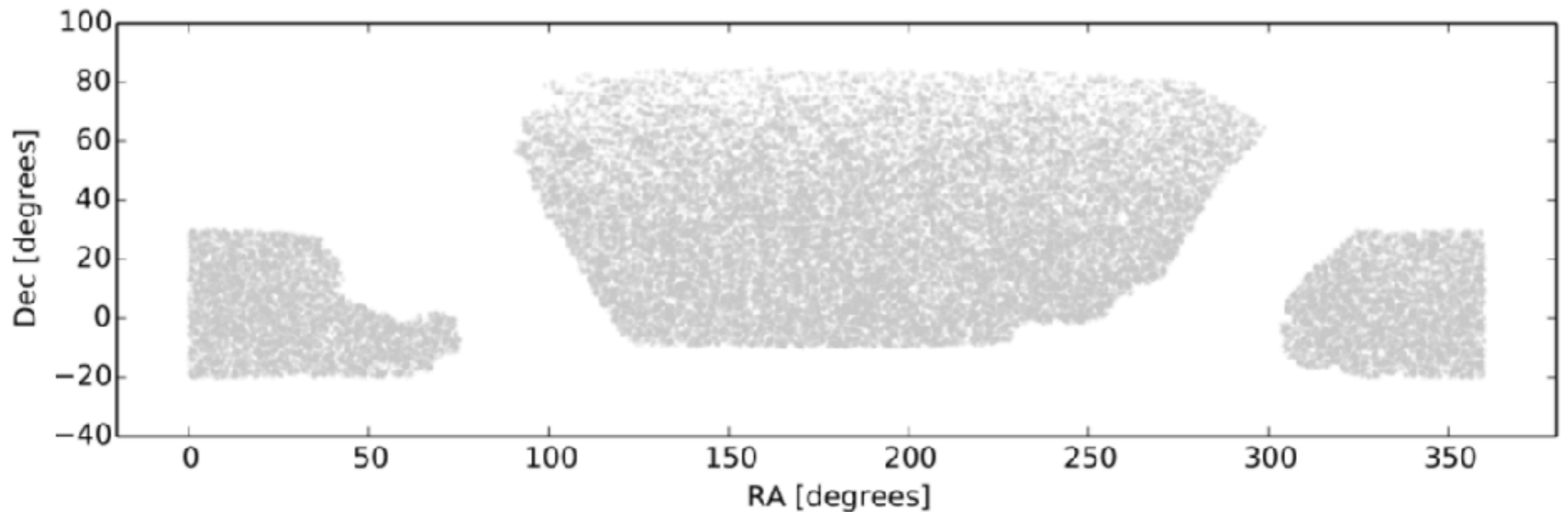
- **DB** & W. J. Percival - arXiv:1703.02070
- W. J. Percival & **DB** - arXiv:1703.02071
- **DB** et al. - arXiv:1805.00951

# The Dark Energy Spectroscopic Instrument (DESI) galaxy survey



Mayall telescope (4m) in Kitt Peak, Arizona

# DESI footprint



total area  $\sim 14,000 \text{ deg}^2$

total number of targets  $\sim 30 \times 10^6$

# DESI targets

Galaxy type	Redshift range	Bands used	Targets per deg <sup>2</sup>	Exposures per deg <sup>2</sup>	Good $z$ 's per deg <sup>2</sup>	Baseline sample
LRG	0.4–1.0	$r, z, W1$	350	580	285	4.0 M
ELG	0.6–1.6	$g, r, z$	2400	1870	1220	17.1 M
QSO (tracers)	$< 2.1$	$g, r, z, W1, W2$	170	170	120	1.7 M
QSO (Ly- $\alpha$ )	$> 2.1$	$g, r, z, W1, W2$	90	250	50	0.7 M
<b>Total in dark time</b>			<b>3010</b>	<b>2870</b>	<b>1675</b>	<b>23.6 M</b>
BGS	0.05–0.4	$r$	700	700	700	9.8 M
<b>Total in bright time</b>			<b>700</b>	<b>700</b>	<b>700</b>	<b>9.8 M</b>

DESI collaboration arXiv:1611.00036

ELG = emission line galaxies

LRG = luminous red galaxies

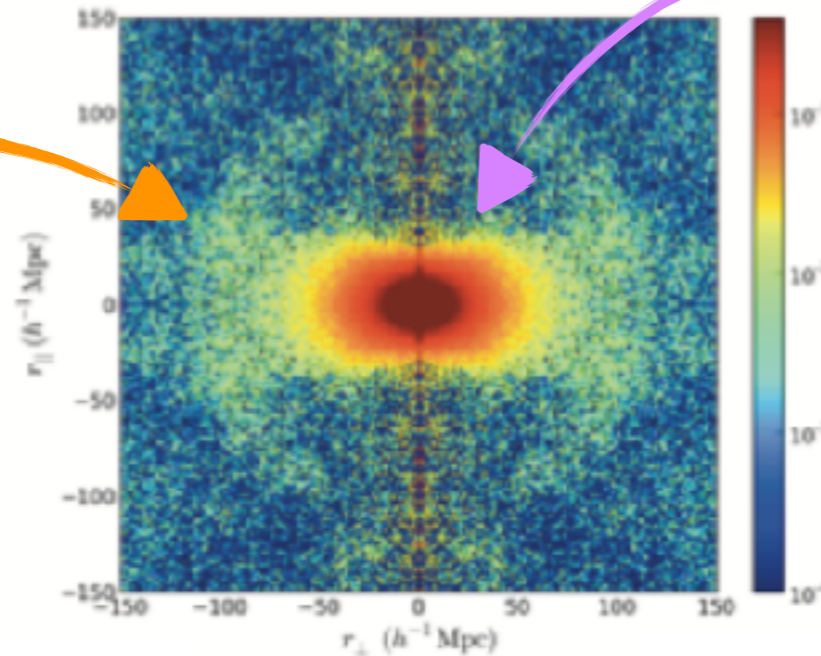
QSO = quasars

BGS = bright galaxy survey

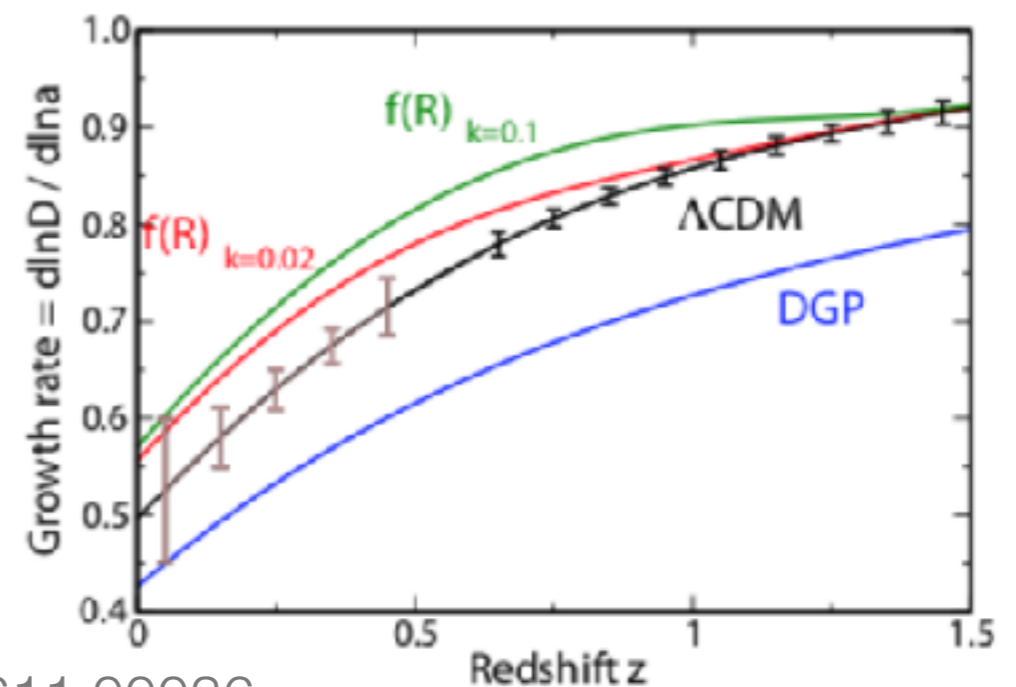
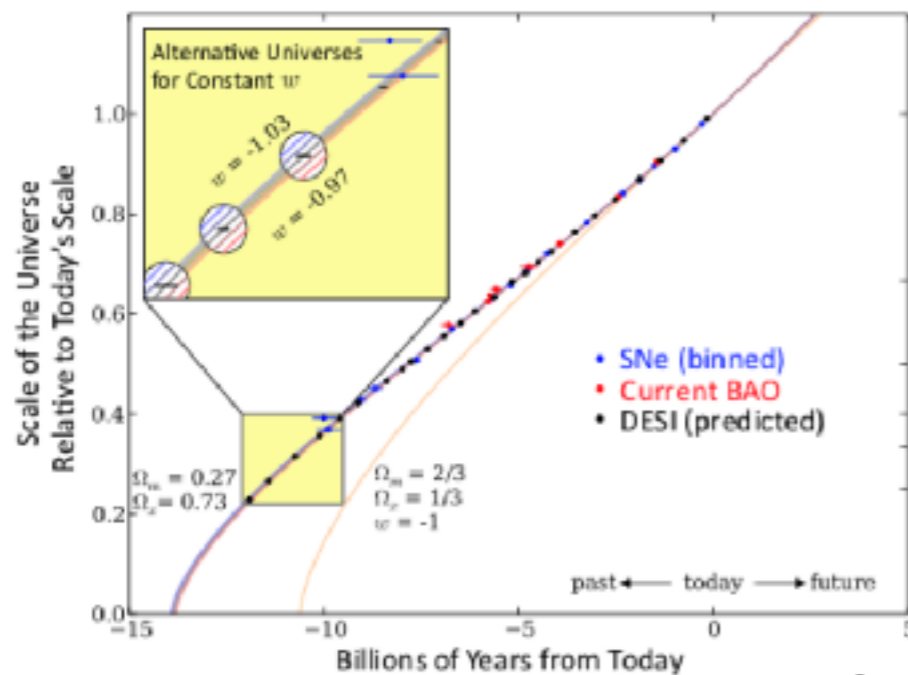
# Why do we want to measure the 3D position of 30M galaxies?

2-point correlation function

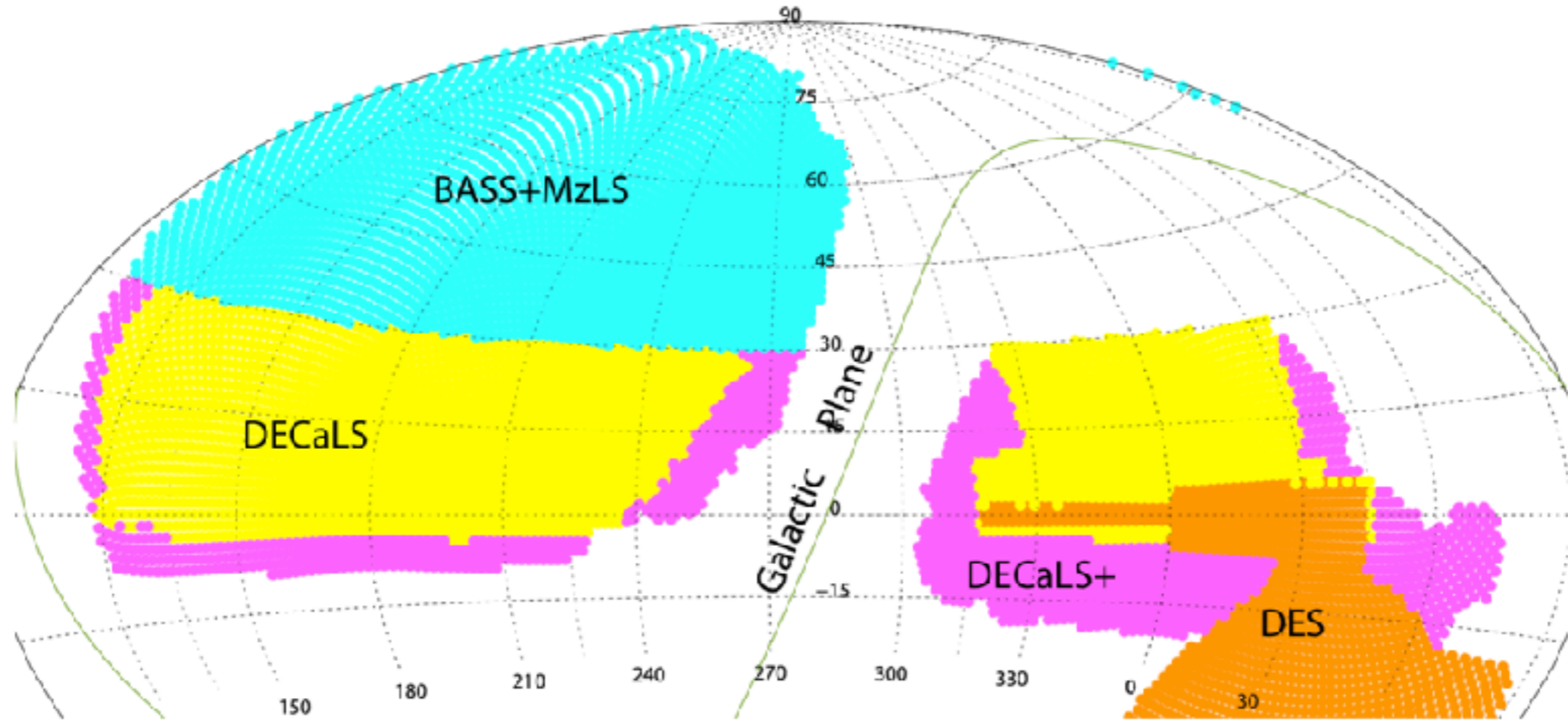
baryon acoustic oscillation (BAO) peak



redshift space distortion (RSD) squashing



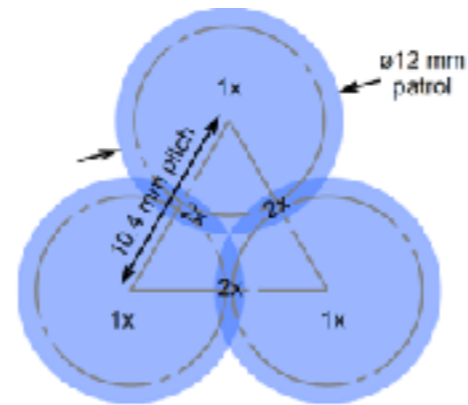
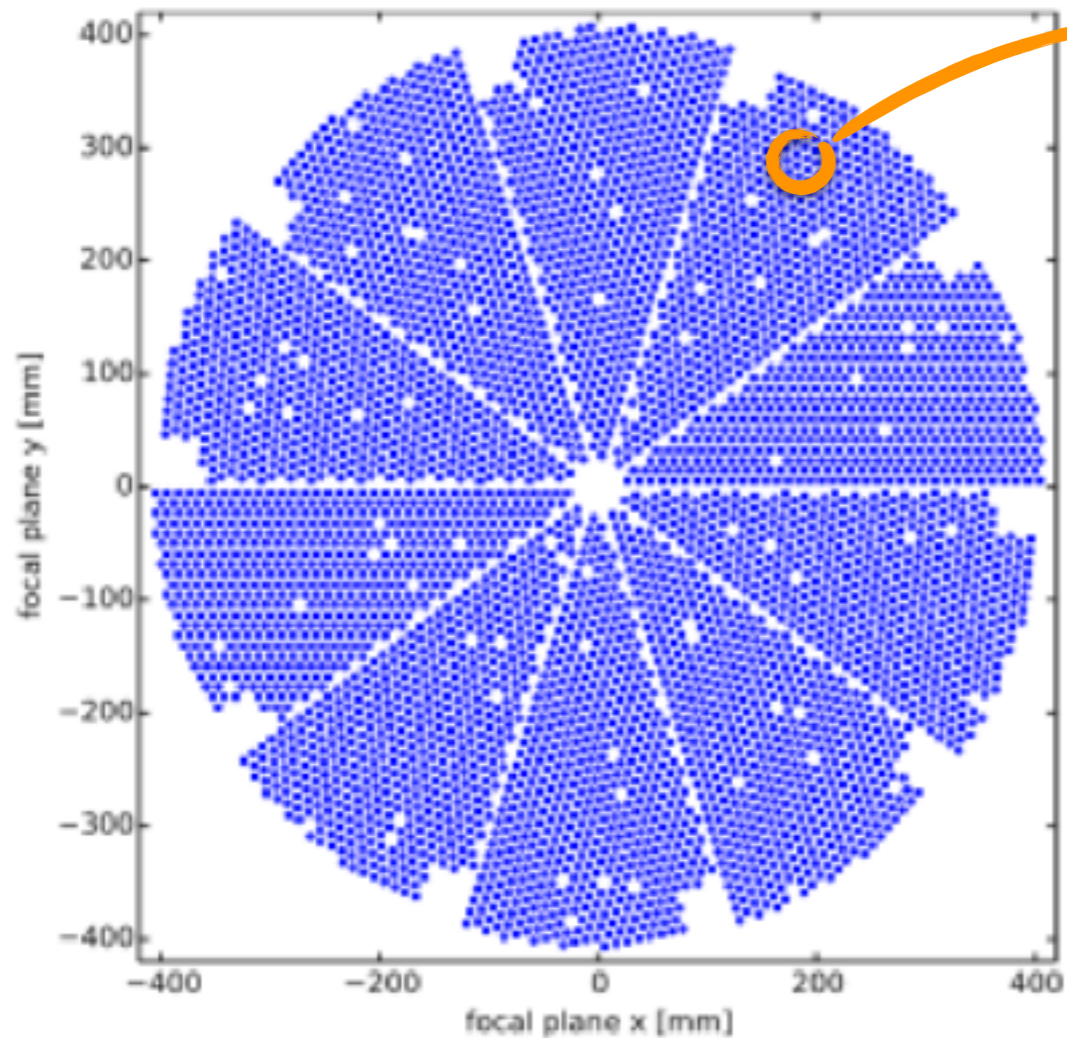
# Parent catalogue of potential targets



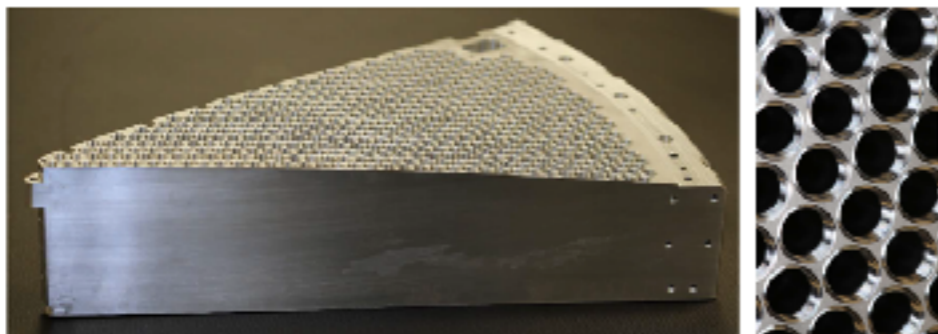
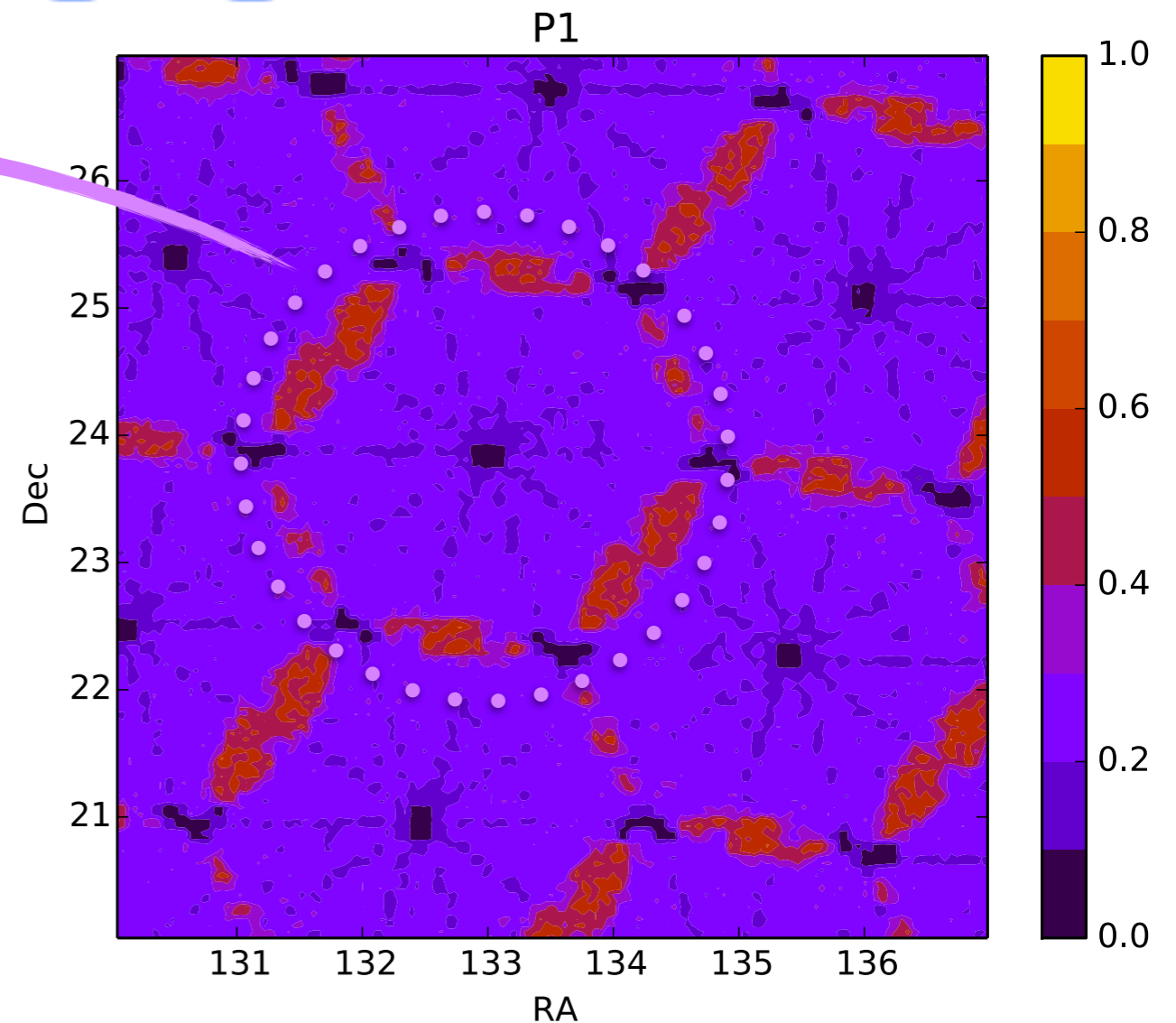
**Figure 3.19:** The primary imaging surveys that will result in targeting data for the DESI project. The footprint at  $\text{DEC} \leq +34^\circ$  will be covered using the Dark Energy Camera (DECam) on the Blanco 4m telescope at Cerro Tololo Inter-American Observatory. The Dark Energy Camera Legacy Survey (DECaLS, in yellow), the Dark Energy Survey (DES, in orange), and the extended DECaLS in the North Galactic Cap (DECaLS+, in purple on left) are underway. A proposal for the remaining extended DECaLS in the South Galactic Cap (DECaLS+, in purple on right) will be submitted. Imaging of the North Galactic Cap region at  $\text{DEC} \geq +34^\circ$  (cyan) will be covered with the 90Prime camera at the Bok 2.3-m telescope in  $g$ - and  $r$ -bands (BASS: the Beijing-Arizona Sky Survey) and with the upgraded MOSAIC-3 camera on the Mayall 4m telescope in  $z$ -band (MzLS: the MOSAIC  $z$ -band Legacy Survey). Both the Bok and Mayall telescopes are located on Kitt Peak National Observatory.

# Fiber assignment and missing observations

5,000 fibers focal plane area = 7.5 deg<sup>2</sup>



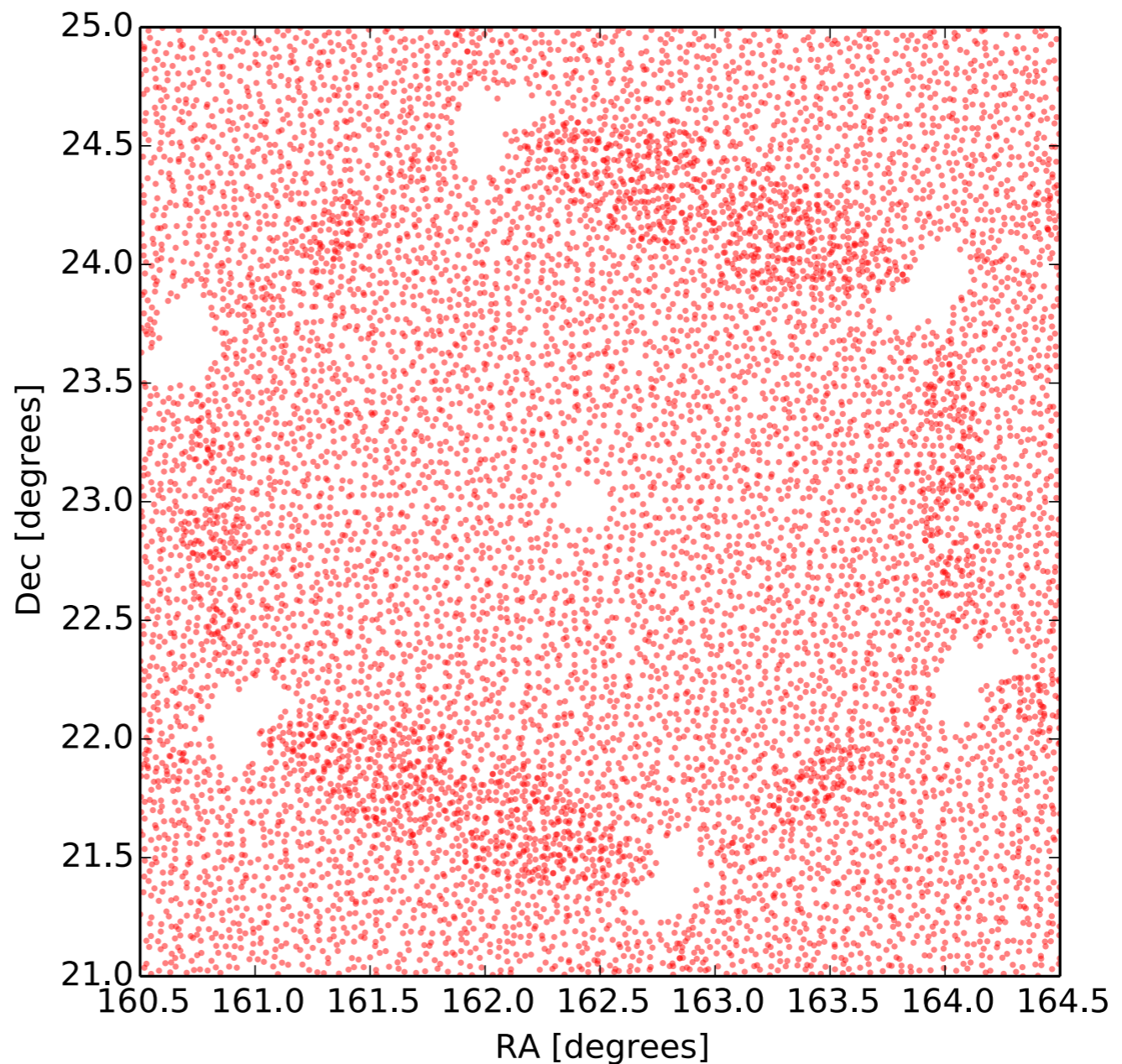
completeness  
after 1 pass of the  
DESI instrument





# DESI observing strategy

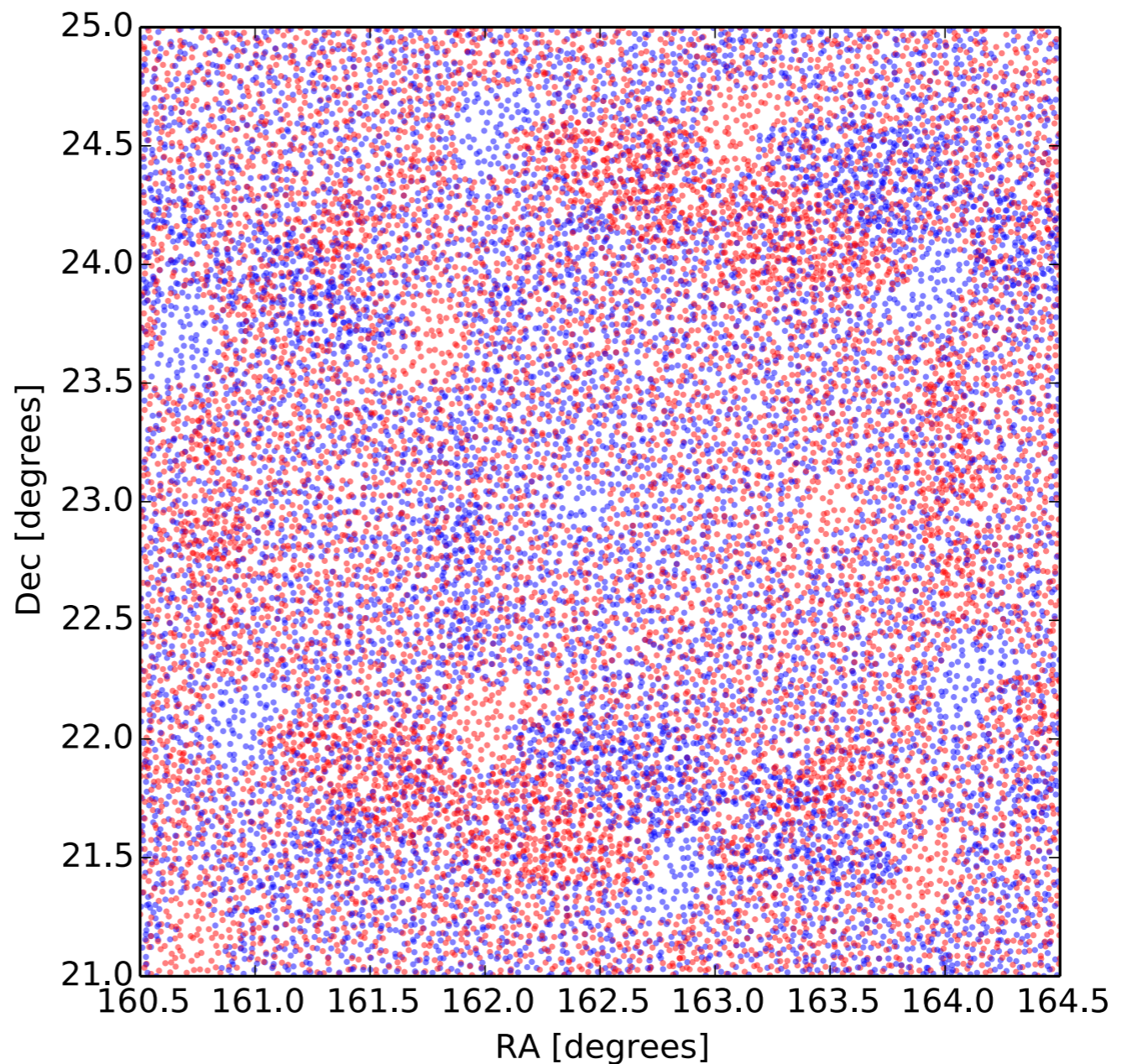
*pass 1*



# DESI observing strategy

*pass 1*

*pass 2*

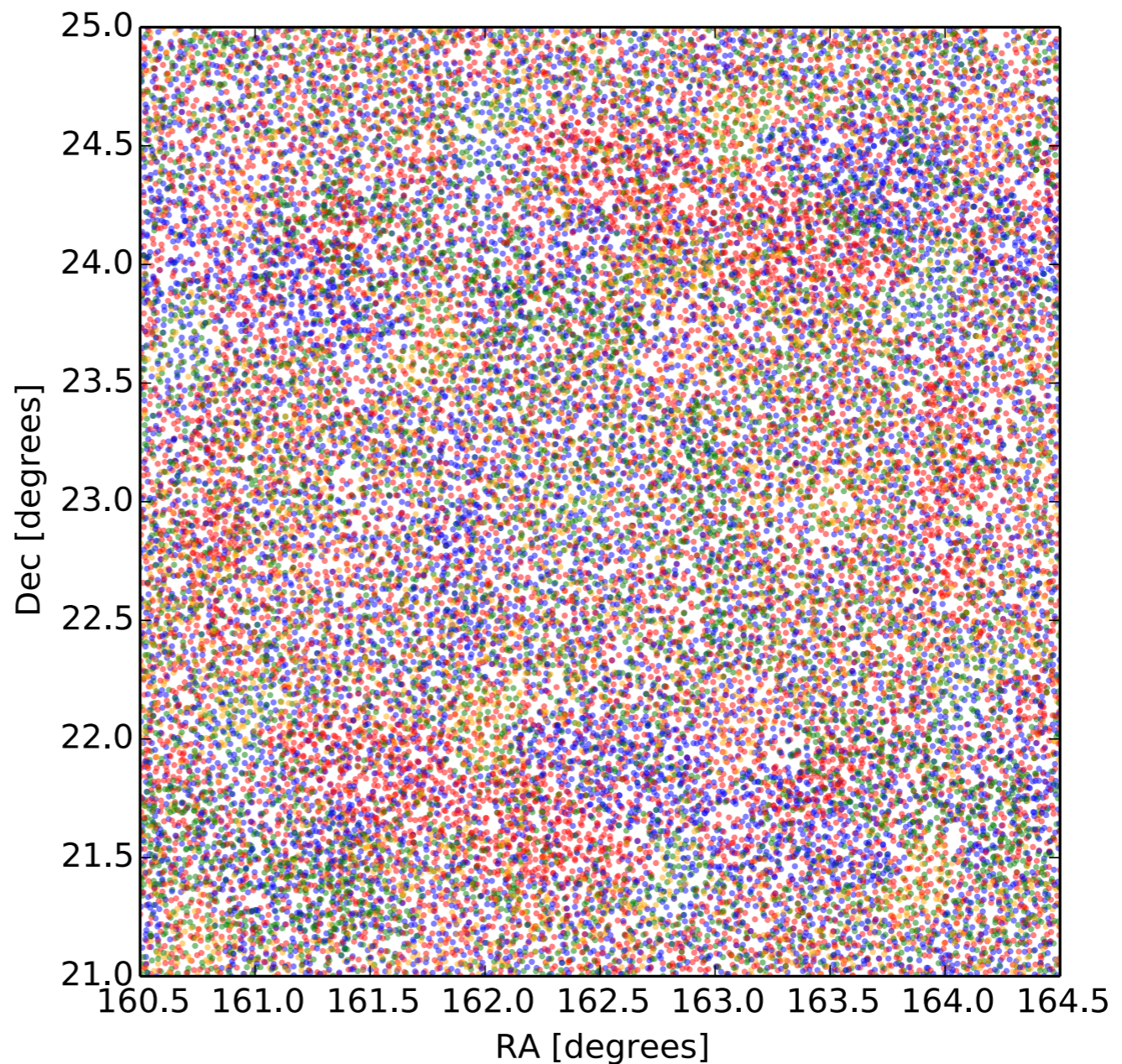


# DESI observing strategy

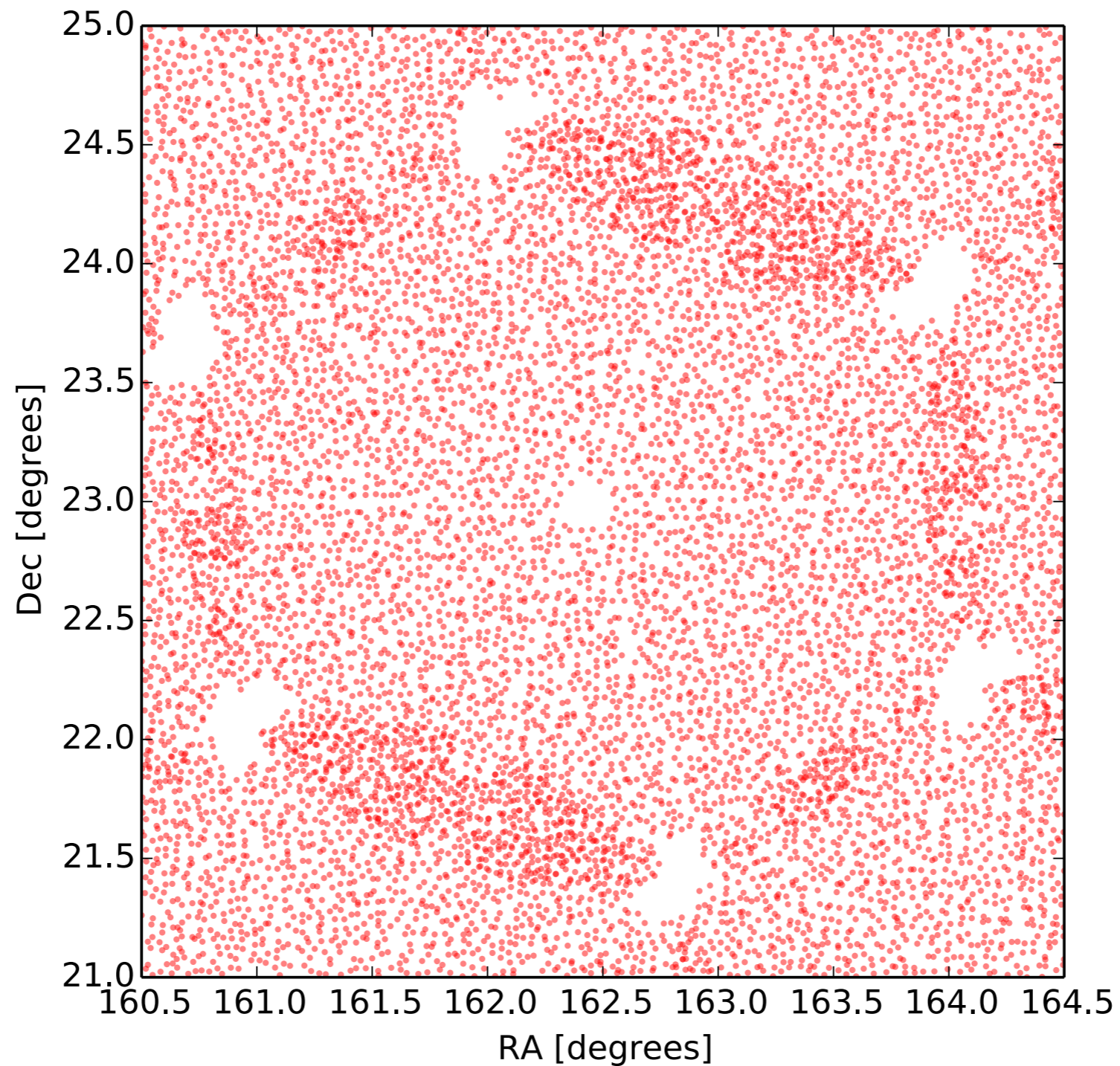
*pass 1*

*pass 2*

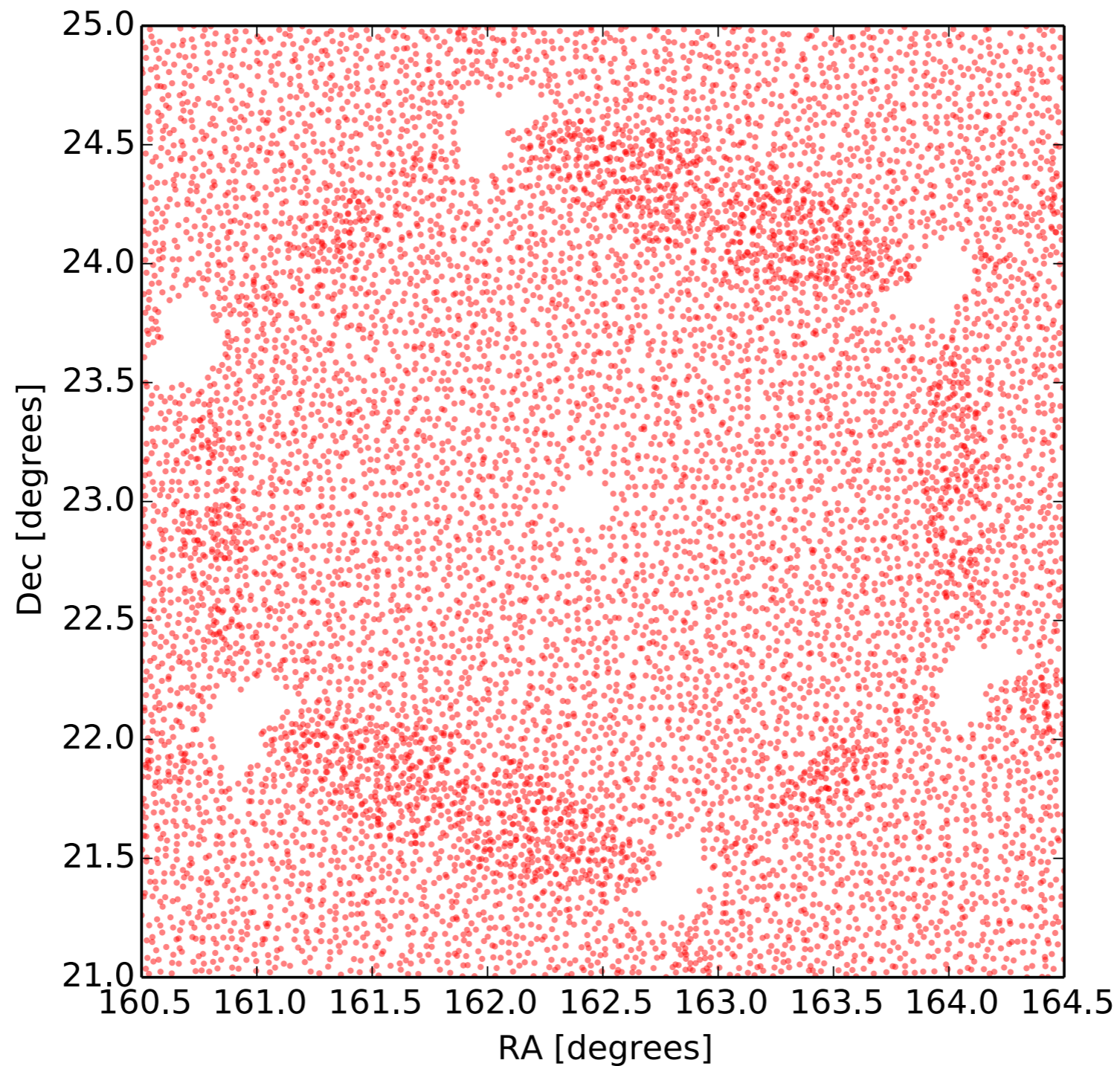
*etc...*



# Targeting realisation 1 (pass 1)



# Targeting realisation 2 (pass 1)



# 2 point correlation function

galaxy-galaxy pairs

$$\xi(\vec{s}) = \frac{DD(\vec{s})}{RR(\vec{s})} - 2\frac{DR(\vec{s})}{RR(\vec{s})} + 1$$

galaxy-random pairs

Landy & Szalay 1993

random-random pairs

# 2 point correlation function

galaxy-galaxy pairs

galaxy-random pairs

$$\xi(\vec{s}) = \frac{DD(\vec{s})}{RR(\vec{s})} - 2\frac{DR(\vec{s})}{RR(\vec{s})} + 1$$

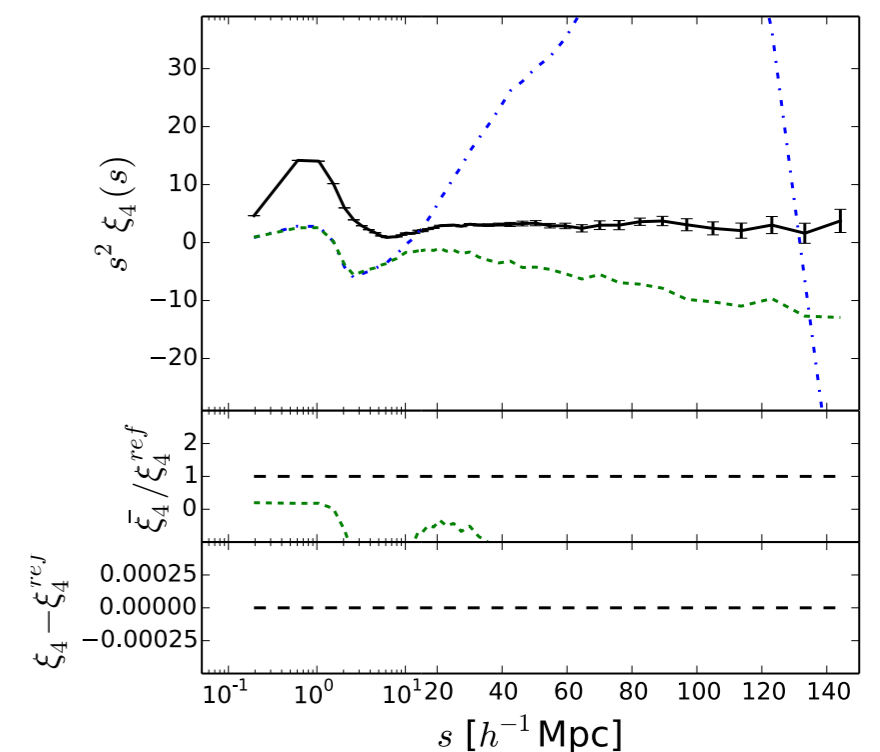
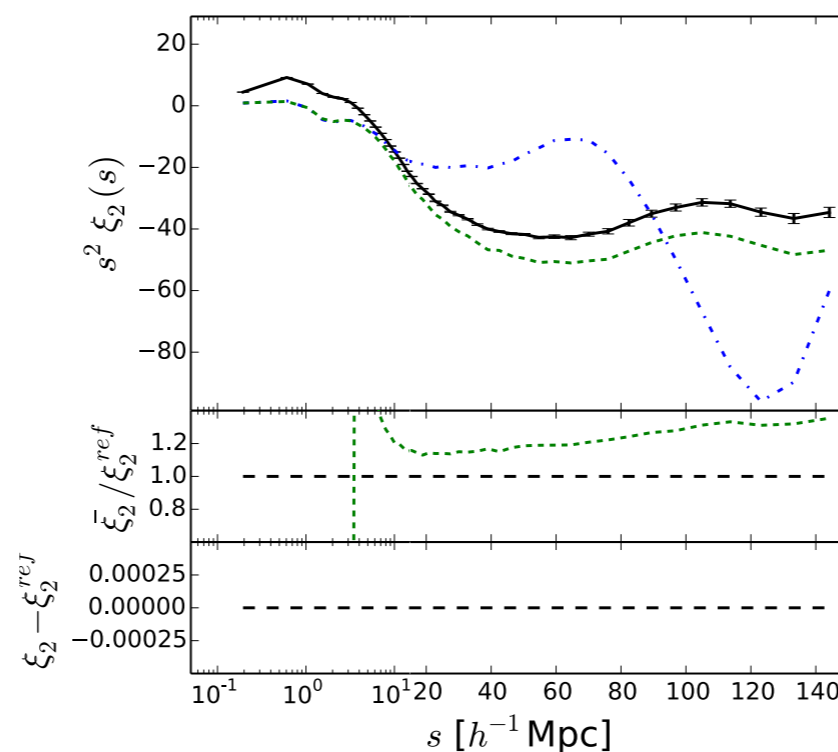
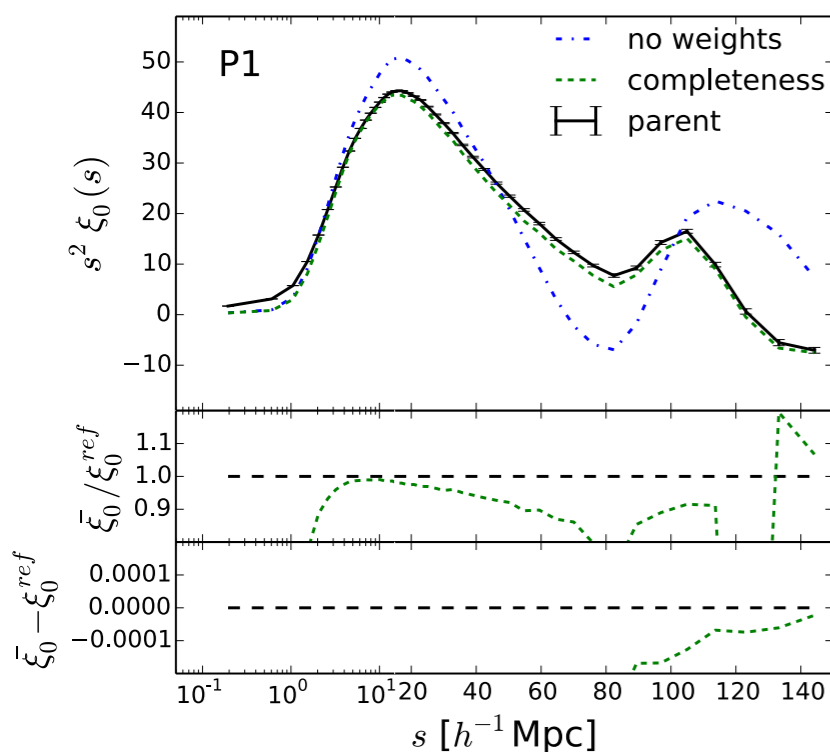
Landy & Szalay 1993

random-random pairs

## Monopole

## Quadrupole

## Hexadecapole



# Countermeasures for missing observations

- Angular upweighting - Hawkins et al. 2003
- Nearest neighbour - Anderson et al. 2012
- Correlation function decomposition - Guo et al. 2012
- Distribution function & top hat function - Hahn et al. 2016
- Target sampling rate - De la Torre et al. 2013, Pezzotta et al. 2017
- Mode subtraction (configuration space) - Burden et al. 2017
- Mode subtraction (Fourier space) - Pinol et al. 2017, Hand et al. 2017
- Pairwise inverse probability (PIP) - Bianchi & Percival 2017, Percival & Bianchi 2017



# Pairwise-Inverse-Probability (PIP) weights: basic idea

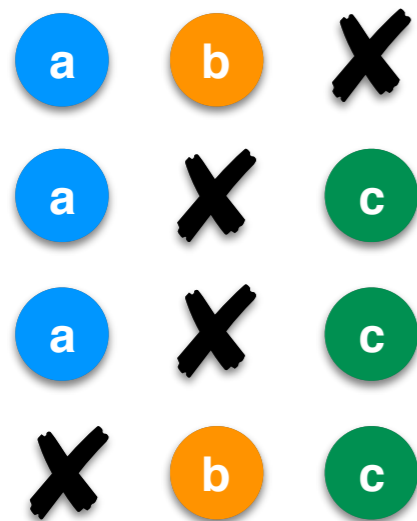


# Pairwise-Inverse-Probability (PIP) weights: basic idea

$$\begin{array}{c} \text{a} \quad \text{b} \quad \text{c} \end{array} \longrightarrow DD = \begin{array}{c} \text{a} \text{b} \\ + \\ \text{a} \text{c} \\ + \\ \text{b} \text{c} \end{array} = 3$$

# Pairwise-Inverse-Probability (PIP) weights: basic idea

$$\begin{matrix} \text{a} & \text{b} & \text{c} \end{matrix} \longrightarrow DD = \begin{matrix} \text{a} & \text{b} \end{matrix} + \begin{matrix} \text{a} & \text{c} \end{matrix} + \begin{matrix} \text{b} & \text{c} \end{matrix} = 3$$



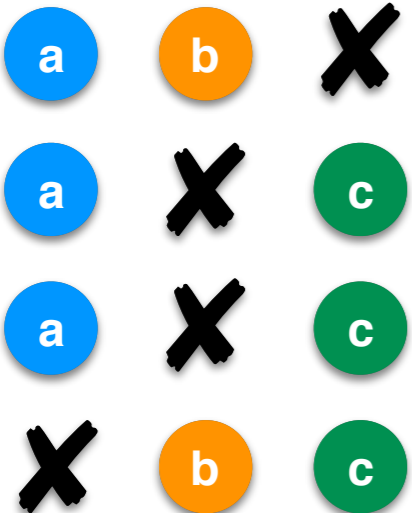
# Pairwise-Inverse-Probability (PIP) weights: basic idea


$$\begin{array}{ccc} \text{a} & \text{b} & \text{c} \end{array} \longrightarrow DD = \begin{array}{cc} \text{a} & \text{b} \end{array} + \begin{array}{cc} \text{a} & \text{c} \end{array} + \begin{array}{cc} \text{b} & \text{c} \end{array} = 3$$

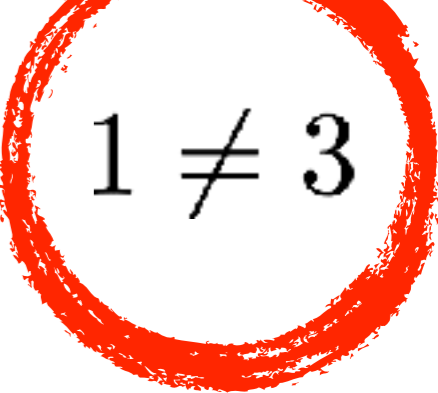
$$\begin{array}{ccc} \text{a} & \text{b} & \times \\ \text{a} & \times & \text{c} \\ \text{a} & \times & \text{c} \\ \times & \text{b} & \text{c} \end{array} \longrightarrow \langle DD \rangle = 1/4 \quad \begin{array}{cc} \text{a} & \text{b} \\ \text{a} & \text{c} \\ \text{a} & \text{c} \\ \text{b} & \text{c} \end{array} = 1$$

# Pairwise-Inverse-Probability (PIP) weights: basic idea

  $\rightarrow DD = \text{a b} + \text{a c} + \text{b c} = 3$

  $\rightarrow \langle DD \rangle = 1/4$

  $= 1$

  $1 \neq 3$

# Pairwise-Inverse-Probability (PIP) weights: basic idea

$$\begin{array}{ccc}
 \text{a} & \text{b} & \text{c}
 \end{array}
 \longrightarrow
 DD = \begin{array}{cc} \text{a} & \text{b} \end{array} + \begin{array}{cc} \text{a} & \text{c} \end{array} + \begin{array}{cc} \text{b} & \text{c} \end{array} = 3$$

$$\begin{array}{ccc}
 \text{a} & \text{b} & \times \\
 \text{a} & \times & \text{c} \\
 \text{a} & \times & \text{c} \\
 \times & \text{b} & \text{c}
 \end{array}
 \longrightarrow
 \langle DD \rangle = 1/4 \left( \begin{array}{cc} 4 & \text{a} & \text{b} & + \\ 2 & \text{a} & \text{c} & + \\ 2 & \text{a} & \text{c} & + \\ 4 & \text{b} & \text{c} & = 3 \end{array} \right)$$

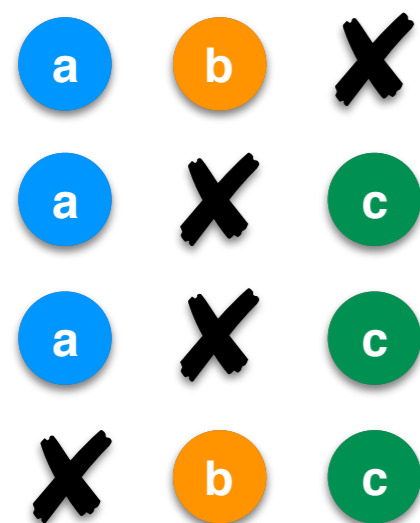
# Pairwise-Inverse-Probability (PIP) weights: basic idea


 $\longrightarrow DD =$ 





 $+$ 

 $+$ 

 $= 3$



$\longrightarrow \langle DD \rangle = 1/4$

$4$  
 $+$   
 $2$  
 $+$   
 $2$  
 $+$   
 $4$  
 $= 3$

$3 = 3$

# Pairwise-Inverse-Probability (PIP) weights: correlation-function estimator

$$\xi(\vec{s}) = \frac{DD(\vec{s})}{RR(\vec{s})} - 2\frac{DR(\vec{s})}{RR(\vec{s})} + 1 \quad \text{Landy \& Szalay 1993}$$



# Pairwise-Inverse-Probability (PIP) weights: correlation-function estimator

$$\xi(\vec{s}) = \frac{DD(\vec{s})}{RR(\vec{s})} - 2 \frac{DR(\vec{s})}{RR(\vec{s})} + 1 \quad \text{Landy \& Szalay 1993}$$

$$DD(\vec{s}) = \sum_{\vec{x}_m - \vec{x}_n \approx \vec{s}} w_{mn} \frac{DD_a^{(p)}(\theta)}{DD_a(\theta)}$$

$$DD_a(\theta) = \sum_{\vec{u}_m \cdot \vec{u}_n \approx \cos(\theta)} w_{mn}$$

$$w_{mn} = \frac{1}{p_{mn}} = \frac{1}{p_m p_n (1 + c_{mn})}$$

# Pairwise-Inverse-Probability (PIP) weights: correlation-function estimator

$$\xi(\vec{s}) = \frac{DD(\vec{s})}{RR(\vec{s})} - 2 \frac{DR(\vec{s})}{RR(\vec{s})} + 1 \quad \text{Landy \& Szalay 1993}$$

$$DD(\vec{s}) = \sum_{\vec{x}_m - \vec{x}_n \approx \vec{s}} w_{mn} \frac{DD_a^{(p)}(\theta)}{DD_a(\theta)}$$

$$DD_a(\theta) = \sum_{\vec{u}_m \cdot \vec{u}_n \approx \cos(\theta)} w_{mn}$$

$$w_{mn} = \frac{1}{p_{mn}} = \frac{1}{p_m p_n (1 + c_{mn})}$$

$$DR(\vec{s}) = \sum_{\vec{x}_m - \vec{y}_n \approx \vec{s}} w_m \frac{DR_a^{(p)}(\theta)}{DR_a(\theta)}$$

$$DR_a(\theta) = \sum_{\vec{u}_m \cdot \vec{v}_n \approx \cos(\theta)} w_m$$

$$w_m = \frac{1}{p_m}$$

# Practical implementation: bitwise weights

$$N_{bits} = 8$$

	1	2	3	4	5	6	7	8
m	✓	✓	✗	✓	✗	✗	✓	✗

# Practical implementation: bitwise weights

$$N_{bits} = 8$$

	1	2	3	4	5	6	7	8
m	✓	✓	✗	✓	✗	✗	✓	✗

$$w_m^{(b)} = \{1, 1, 0, 1, 0, 0, 1, 0\}$$

$$\text{popcnt} [w_m^{(b)}] = 4$$

$$p_m = \frac{4}{N_{bits}} = \frac{1}{2}$$

$$w_m = \frac{1}{p_m} = 2$$

# Practical implementation: bitwise weights

$$N_{bits} = 8$$

	1	2	3	4	5	6	7	8
m	✓	✓	✗	✓	✗	✗	✓	✗
n	✗	✓	✓	✗	✓	✓	✓	✓

$$w_m^{(b)} = \{1, 1, 0, 1, 0, 0, 1, 0\}$$

$$\text{popcnt} [w_m^{(b)}] = 4$$

$$p_m = \frac{4}{N_{bits}} = \frac{1}{2}$$

$$w_m = \frac{1}{p_m} = 2$$

$$w_n^{(b)} = \{0, 1, 1, 0, 1, 1, 1, 1\}$$

$$\text{popcnt} [w_n^{(b)}] = 6$$

$$p_n = \frac{6}{N_{bits}} = \frac{3}{4}$$

$$w_n = \frac{1}{p_n} = \frac{4}{3}$$

# Practical implementation: bitwise weights

$$N_{bits} = 8$$

	1	2	3	4	5	6	7	8
m	✓	✓	✗	✓	✗	✗	✓	✗
n	✗	✓	✓	✗	✓	✓	✓	✓

$$w_m^{(b)} = \{1, 1, 0, 1, 0, 0, 1, 0\}$$

$$\text{popcnt} [w_m^{(b)}] = 4$$

$$p_m = \frac{4}{N_{bits}} = \frac{1}{2}$$

$$w_m = \frac{1}{p_m} = 2$$

$$w_n^{(b)} = \{0, 1, 1, 0, 1, 1, 1, 1\}$$

$$\text{popcnt} [w_n^{(b)}] = 6$$

$$p_n = \frac{6}{N_{bits}} = \frac{3}{4}$$

$$w_n = \frac{1}{p_n} = \frac{4}{3}$$

$$w_{mn} = w_m w_n = \frac{8}{3}$$

# Practical implementation: bitwise weights

$$N_{bits} = 8$$

	1	2	3	4	5	6	7	8
m	✓	✓	✗	✓	✗	✗	✓	✗
n	✗	✓	✓	✗	✓	✓	✓	✓

$$w_m^{(b)} = \{1, 1, 0, 1, 0, 0, 1, 0\}$$

$$w_n^{(b)} = \{0, 1, 1, 0, 1, 1, 1, 1\}$$

# Practical implementation: bitwise weights

$$N_{bits} = 8$$

	1	2	3	4	5	6	7	8
m	✓	✓	✗	✓	✗	✗	✓	✗
n	✗	✓	✓	✗	✓	✓	✓	✓

$$w_m^{(b)} = \{1, 1, 0, 1, 0, 0, 1, 0\}$$

$$w_n^{(b)} = \{0, 1, 1, 0, 1, 1, 1, 1\}$$

$$w_{mn}^{(b)} \equiv w_m^{(b)} \text{ and } w_n^{(b)} = \{0, 1, 0, 0, 0, 0, 1, 0\}$$

$$\text{popcnt} \left[ w_{mn}^{(b)} \right] = 2 \quad p_{mn} = \frac{2}{N_{bits}} = \frac{1}{4}$$



# Practical implementation: bitwise weights

$$N_{bits} = 8$$

	1	2	3	4	5	6	7	8
m	✓	✓	✗	✓	✗	✗	✓	✗
n	✗	✓	✓	✗	✓	✓	✓	✓

$$w_m^{(b)} = \{1, 1, 0, 1, 0, 0, 1, 0\}$$

$$w_n^{(b)} = \{0, 1, 1, 0, 1, 1, 1, 1\}$$

$$w_{mn}^{(b)} \equiv w_m^{(b)} \text{ and } w_n^{(b)} = \{0, 1, 0, 0, 0, 0, 1, 0\}$$

$$\text{popcnt} \left[ w_{mn}^{(b)} \right] = 2 \quad p_{mn} = \frac{2}{N_{bits}} = \frac{1}{4}$$

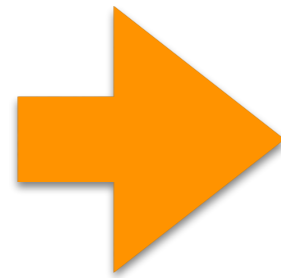
$$w_{mn} = \frac{1}{p_{mn}} = 4$$

# Practical implementation: summary

The pair weights are computed (as usual) while doing pair counts via a simple function of individual weights

*Standard  
approach*

$$w_{mn} = w_m w_n$$



*Bitwise weights*

$$w_{mn} = \frac{N_{bits}}{\text{popcnt} \left[ w_m^{(b)} \text{ and } w_n^{(b)} \right]}$$

# “Why angular upweighting?”

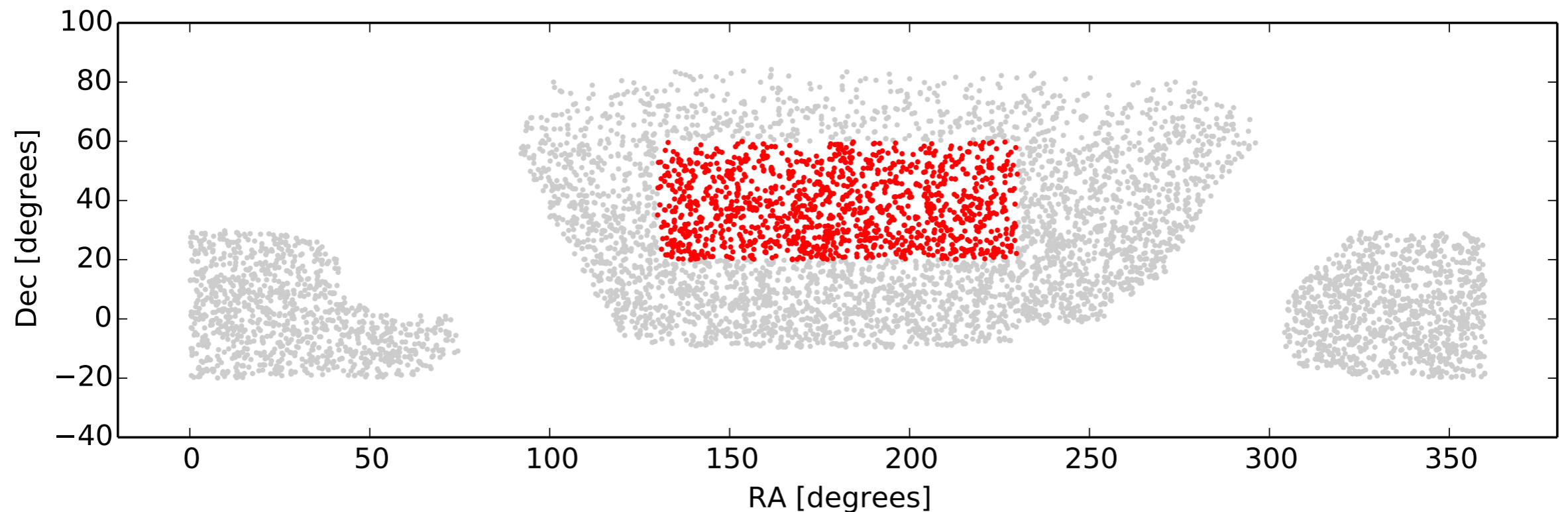
- Natural way to **minimise the variance**, see W. J. Percival & **DB** - arXiv:1703.02071
- Allows to relax the assumption that **no pair has zero probability of being observed** (**not true**, e.g., for BOSS and DESI) to the weaker assumption that **all the zero-probability pairs have the same clustering properties of the observed pairs** (**true**, e.g., for BOSS and DESI)

# Algorithm

1. Evaluate PIP weights by running the targeting algorithm many times ( $N \sim 1000?$ )
2. Compute the angular pair counts DD and DR of the full parent sample
3. Compute the angular pair counts DD and DR of the observed sample using PIP weights
4. Compute the 3D correlation function of the observed sample via PIP weights + ang. upweighting (i.e. ratios between 2 and 3)

# Comparison to mocks:

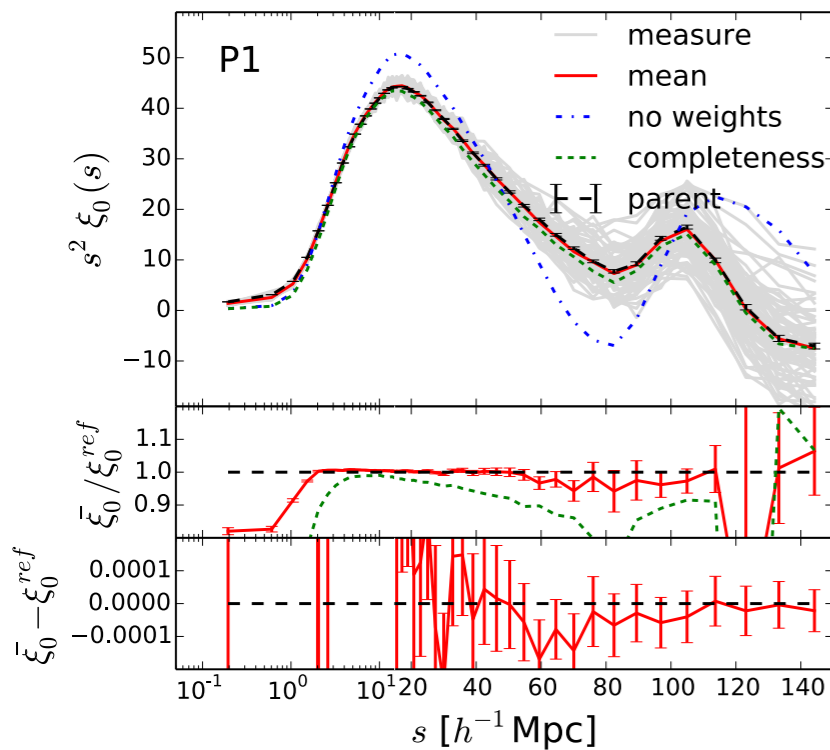
## DESJ footprint vs **our sample**



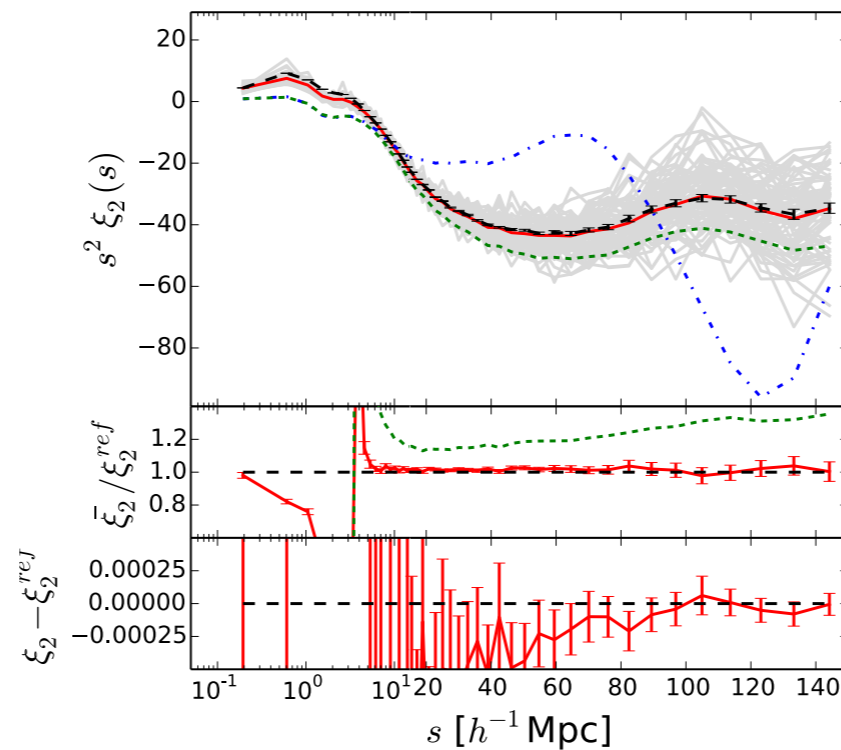
# Comparison to DESI mocks: pass 1

completeness = 25%

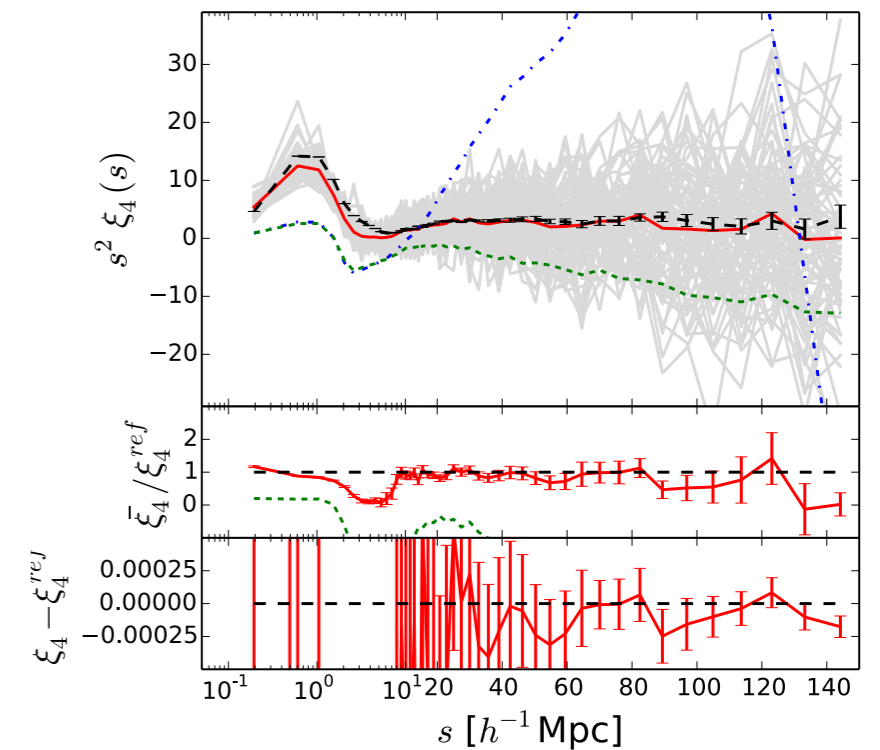
## Monopole



## Quadrupole



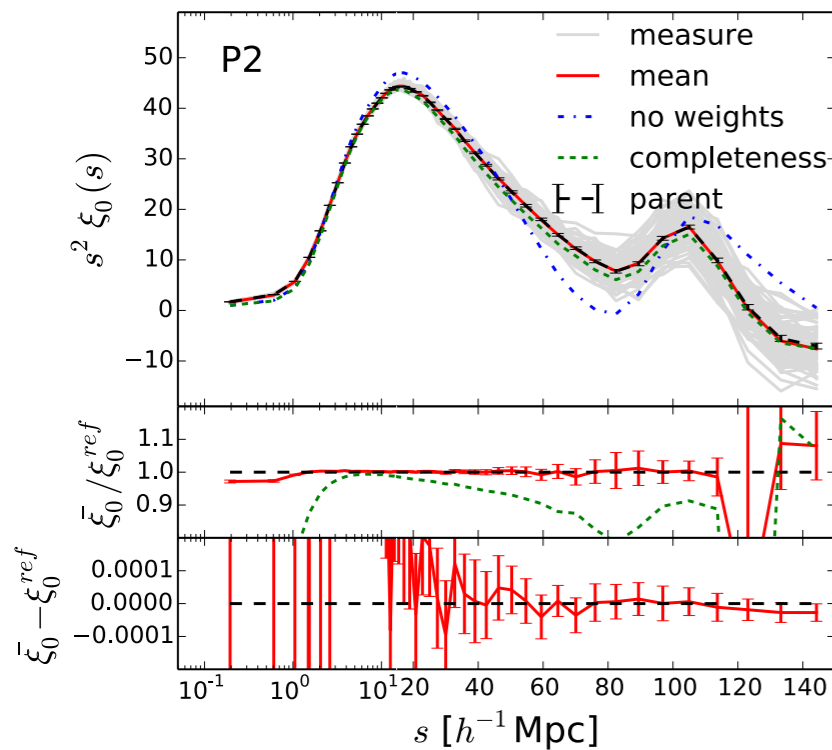
## Hexadecapole



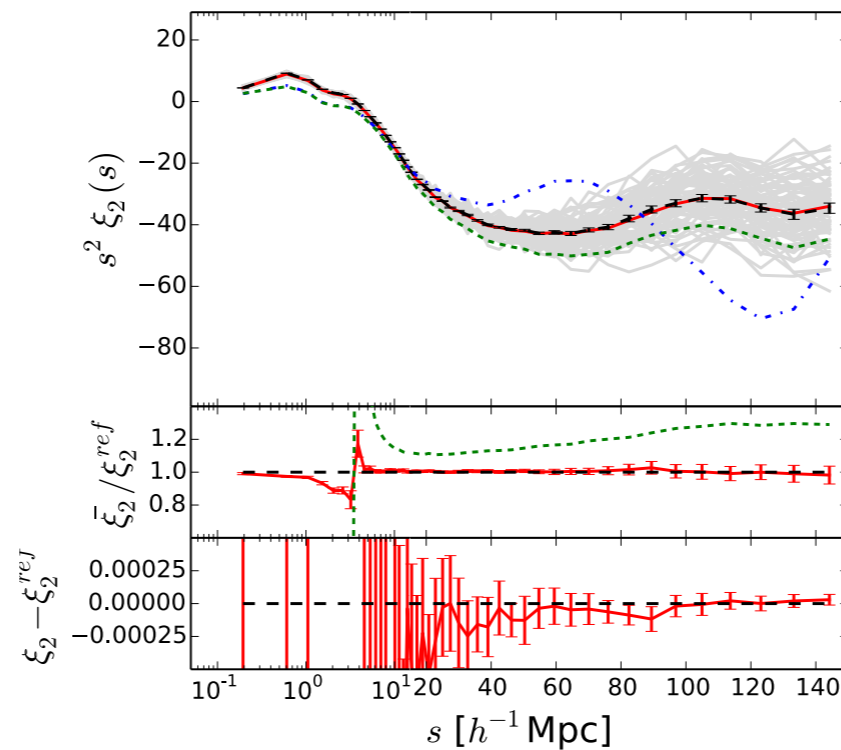
# Comparison to DESI mocks: pass 2

completeness = 48%

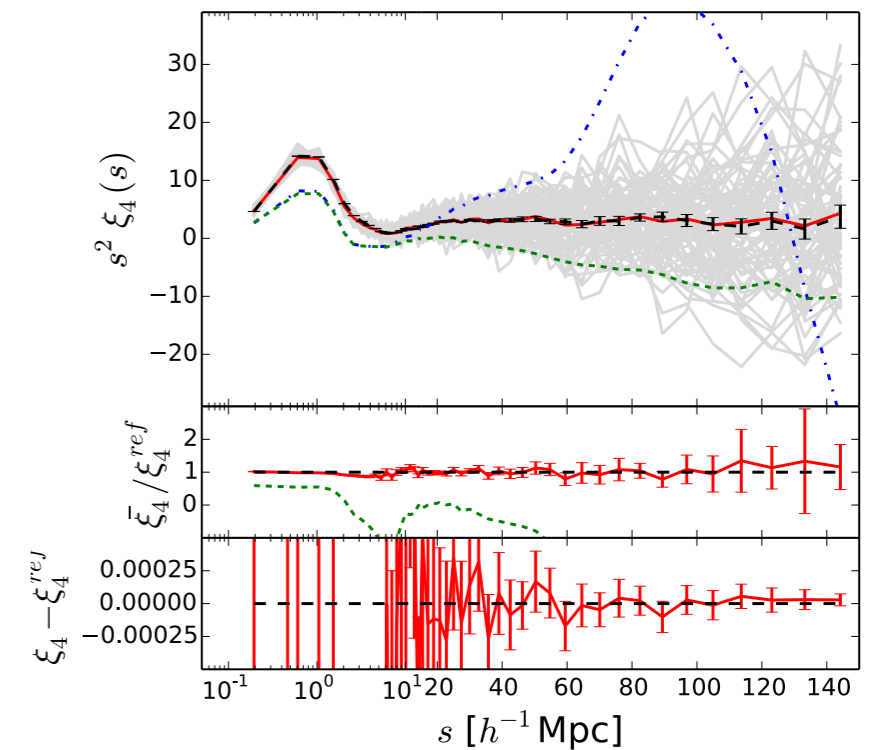
## Monopole



## Quadrupole



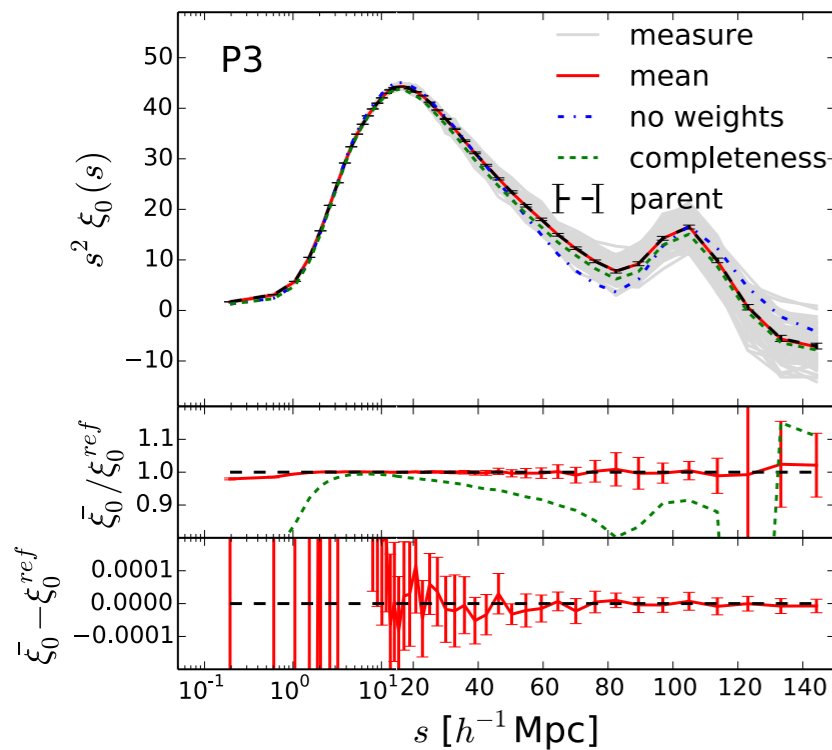
## Hexadecapole



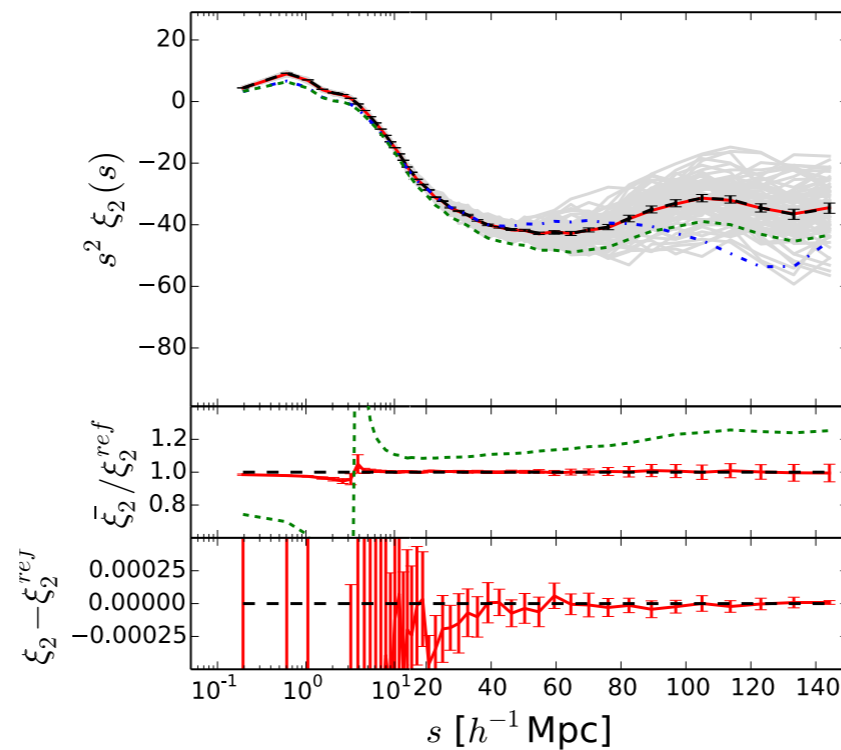
# Comparison to DESI mocks: pass 3

completeness = 67%

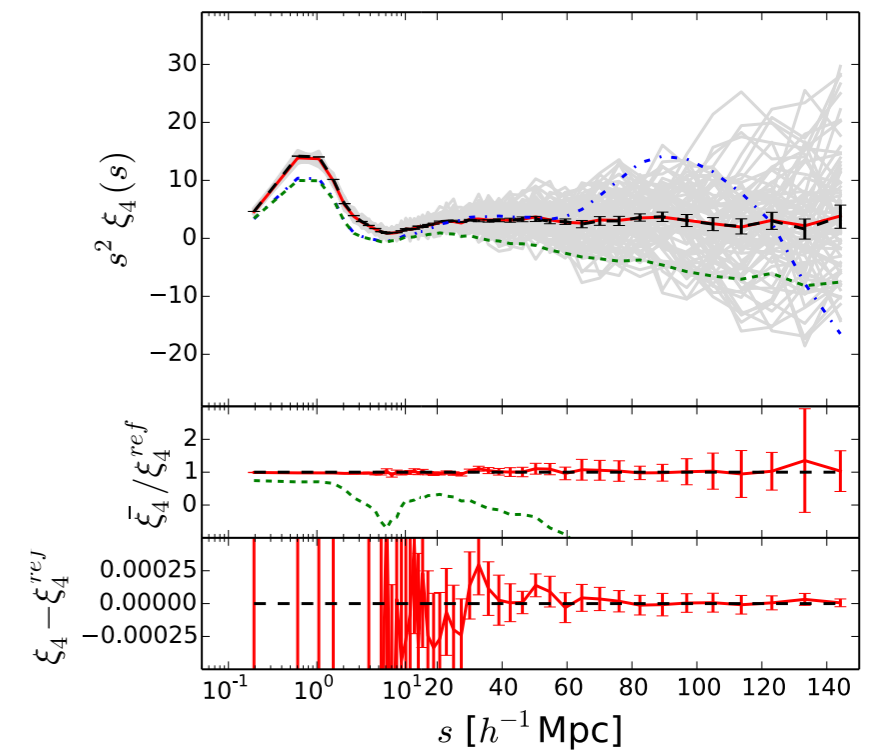
## Monopole



## Quadrupole



## Hexadecapole

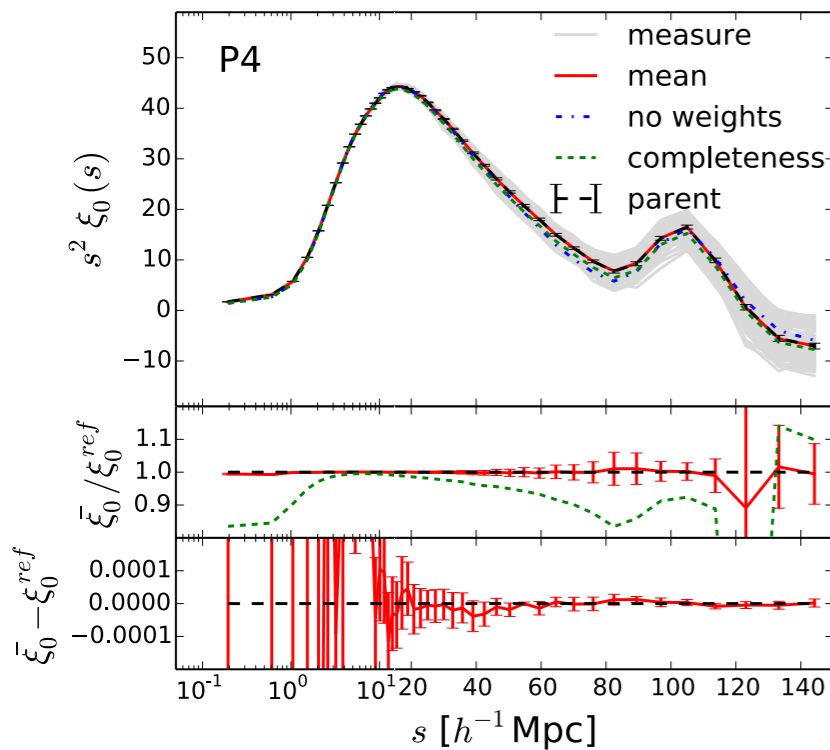




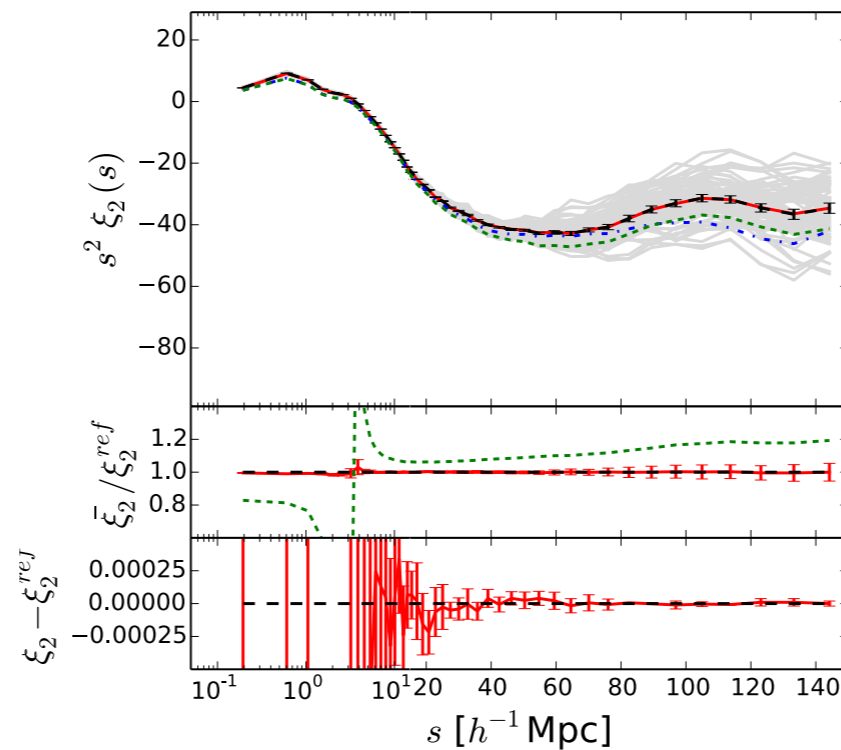
# Comparison to DESI mocks: pass 4

completeness = 81%

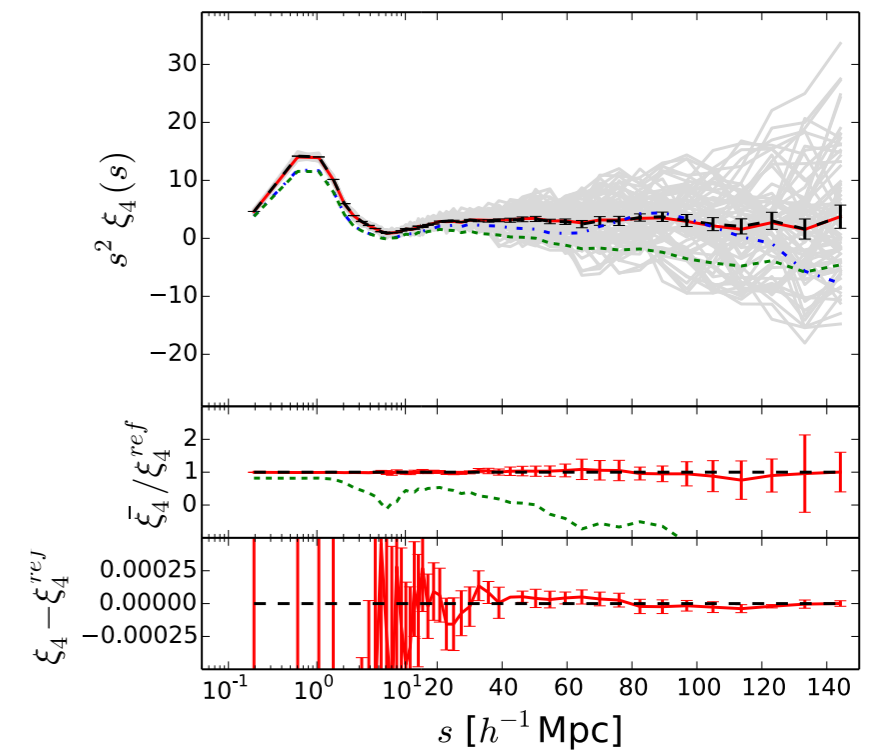
## Monopole



## Quadrupole

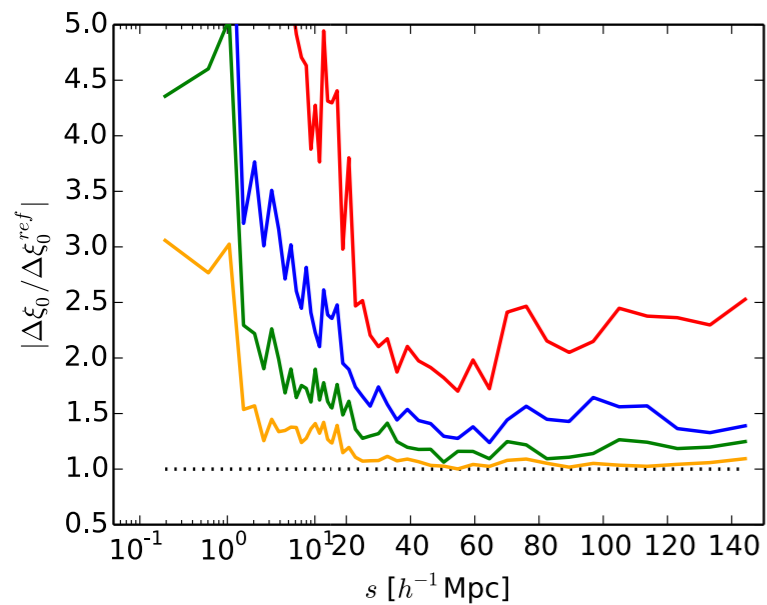


## Hexadecapole

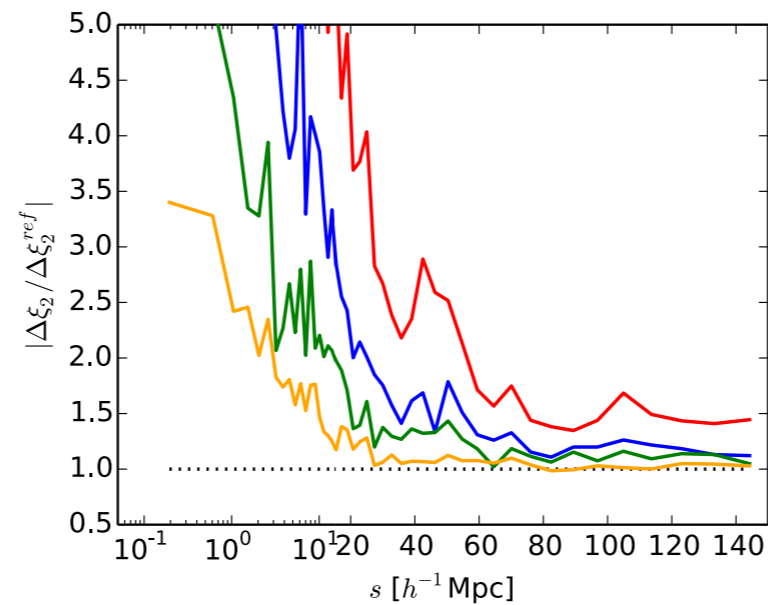


# Error vs cosmic variance

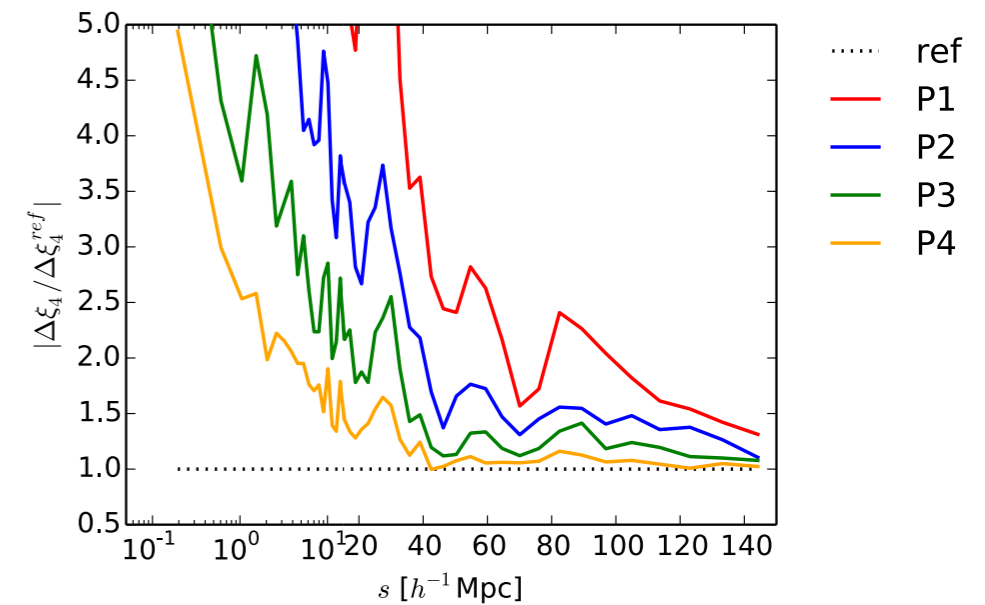
## Monopole



## Quadrupole

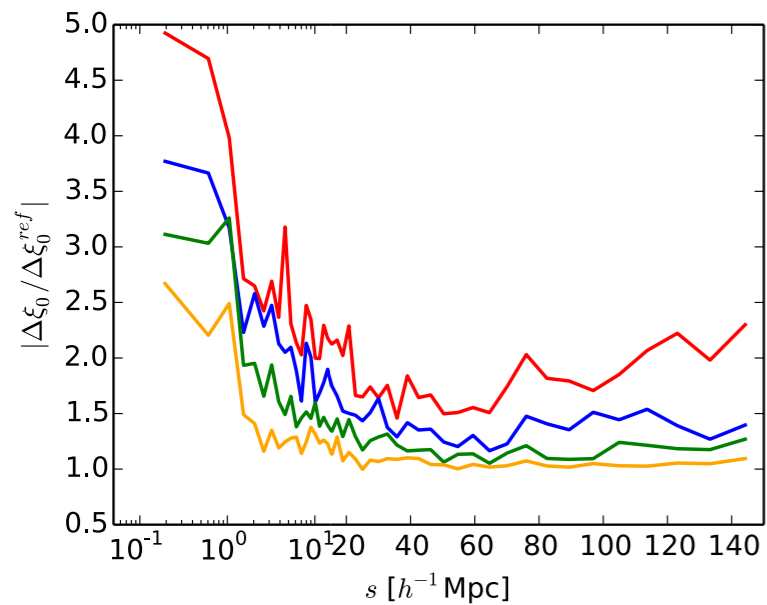


## Hexadecapole

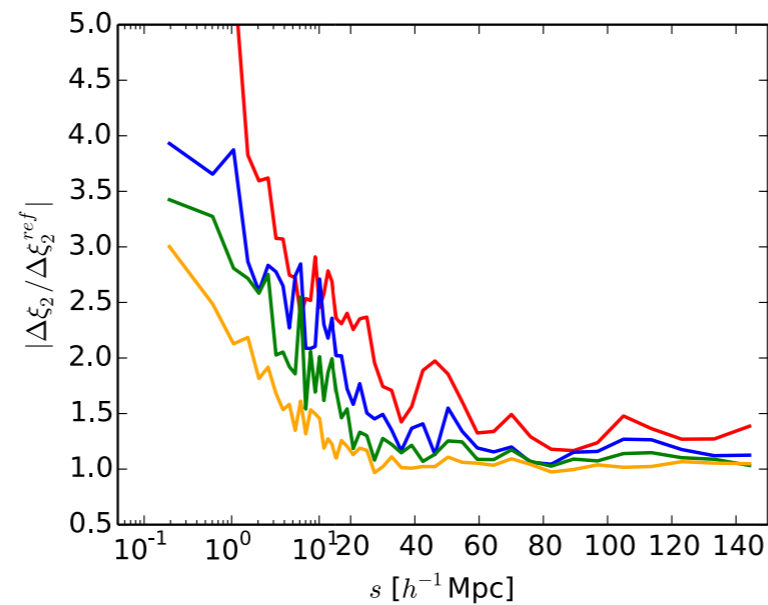


# Error vs uniform dilution

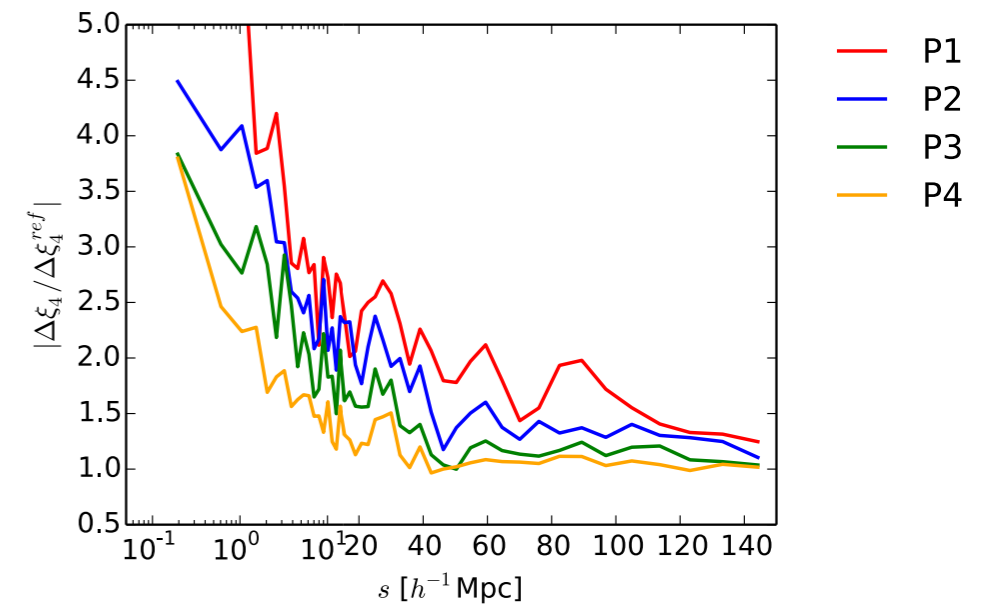
## Monopole



## Quadrupole



## Hexadecapole



# Summary

- We developed a theoretically-motivated weighting scheme capable to recover unbiased estimates of the galaxy clustering for a large class of targeting algorithm
- We provided a practical implementation of the algorithm
- We showed that this approach works for the DESI survey

(Main) Things to do:

- Fourier space
- Reconstruction