

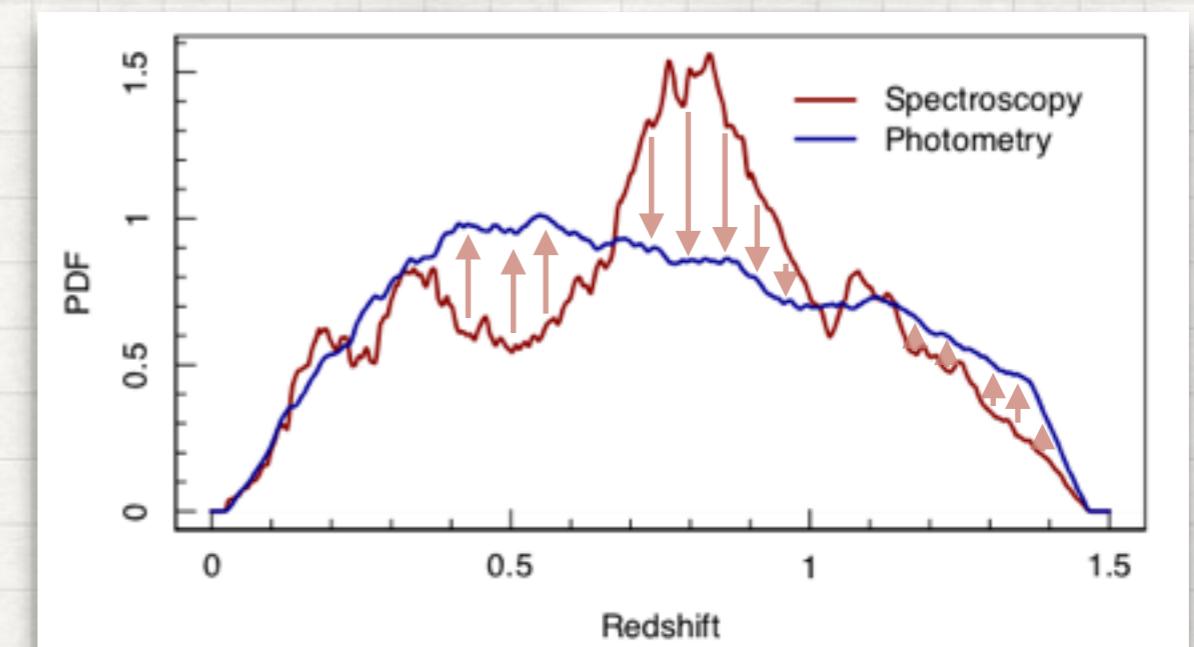
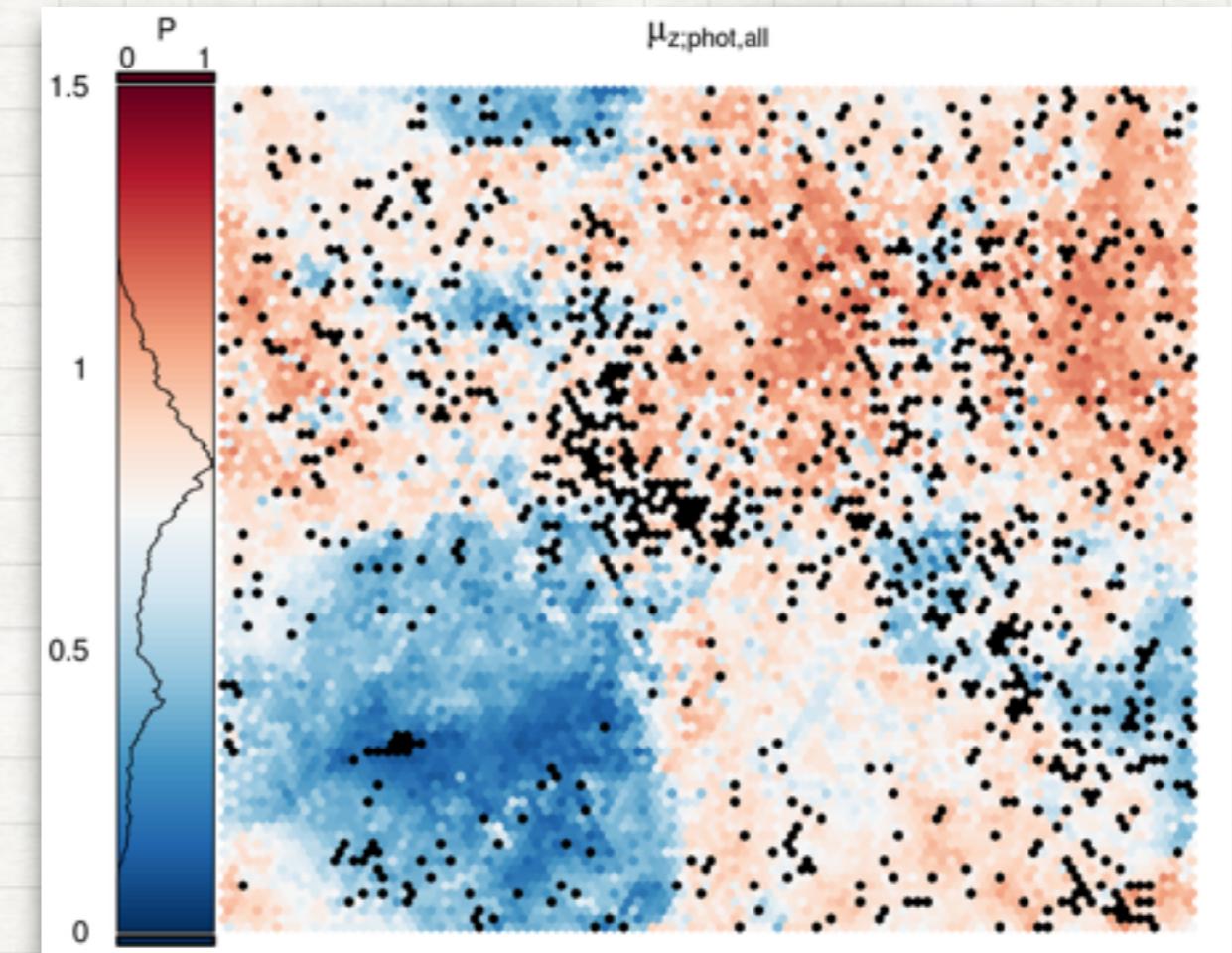
GCCL SEMINAR 06-03-2020

# PHOTOMETRIC REDSHIFT CALIBRATION WITH SELF ORGANISING MAPS

ANGUS H WRIGHT  
RESEARCH FELLOW

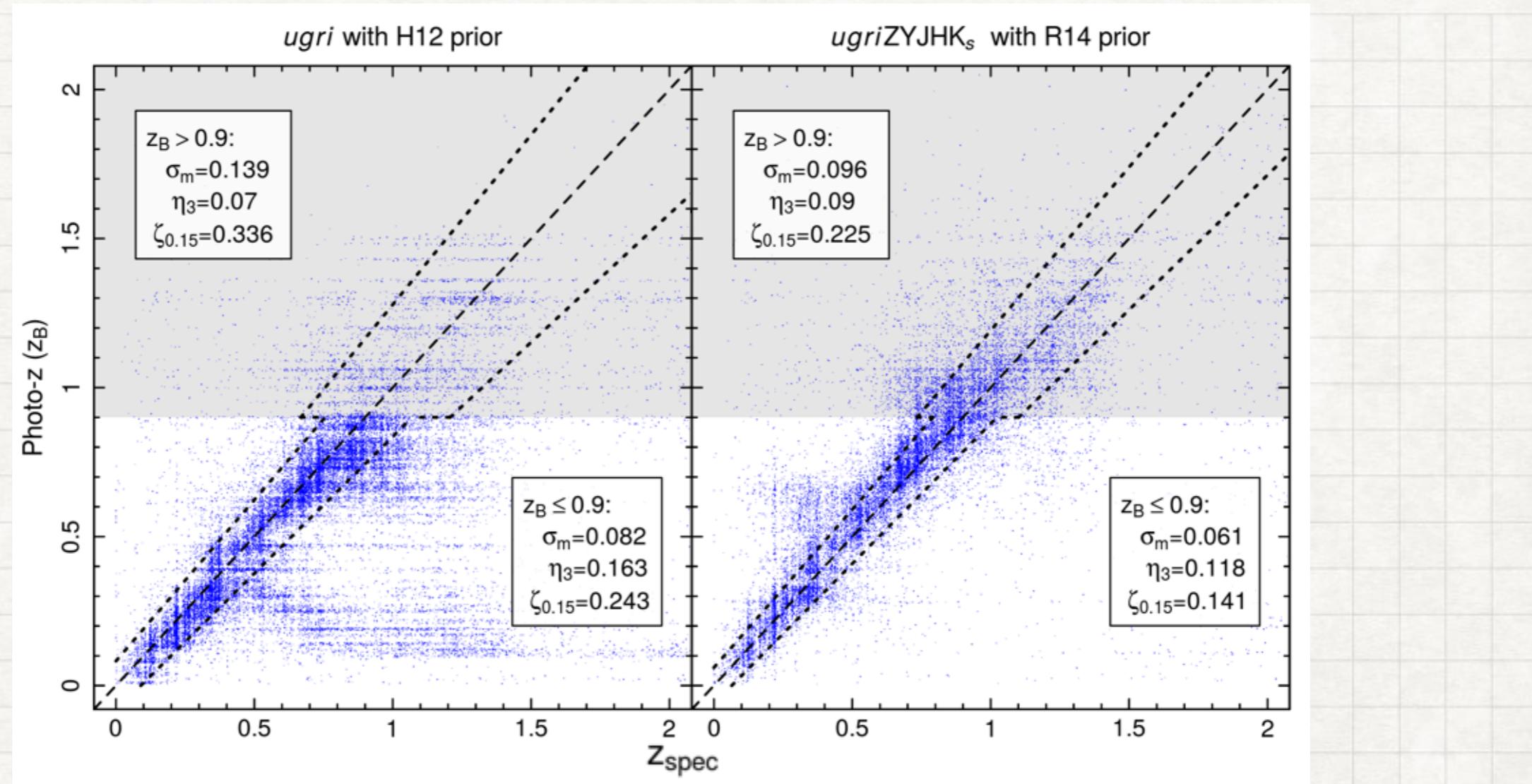
GERMAN CENTRE FOR  
COSMOLOGICAL LENSING (GCCL)

RUHR UNIVERSITÄT BOCHUM, GERMANY



# REDSHIFT CALIBRATION FOR COSMIC SHEAR

## PHOTOMETRIC REDSHIFTS

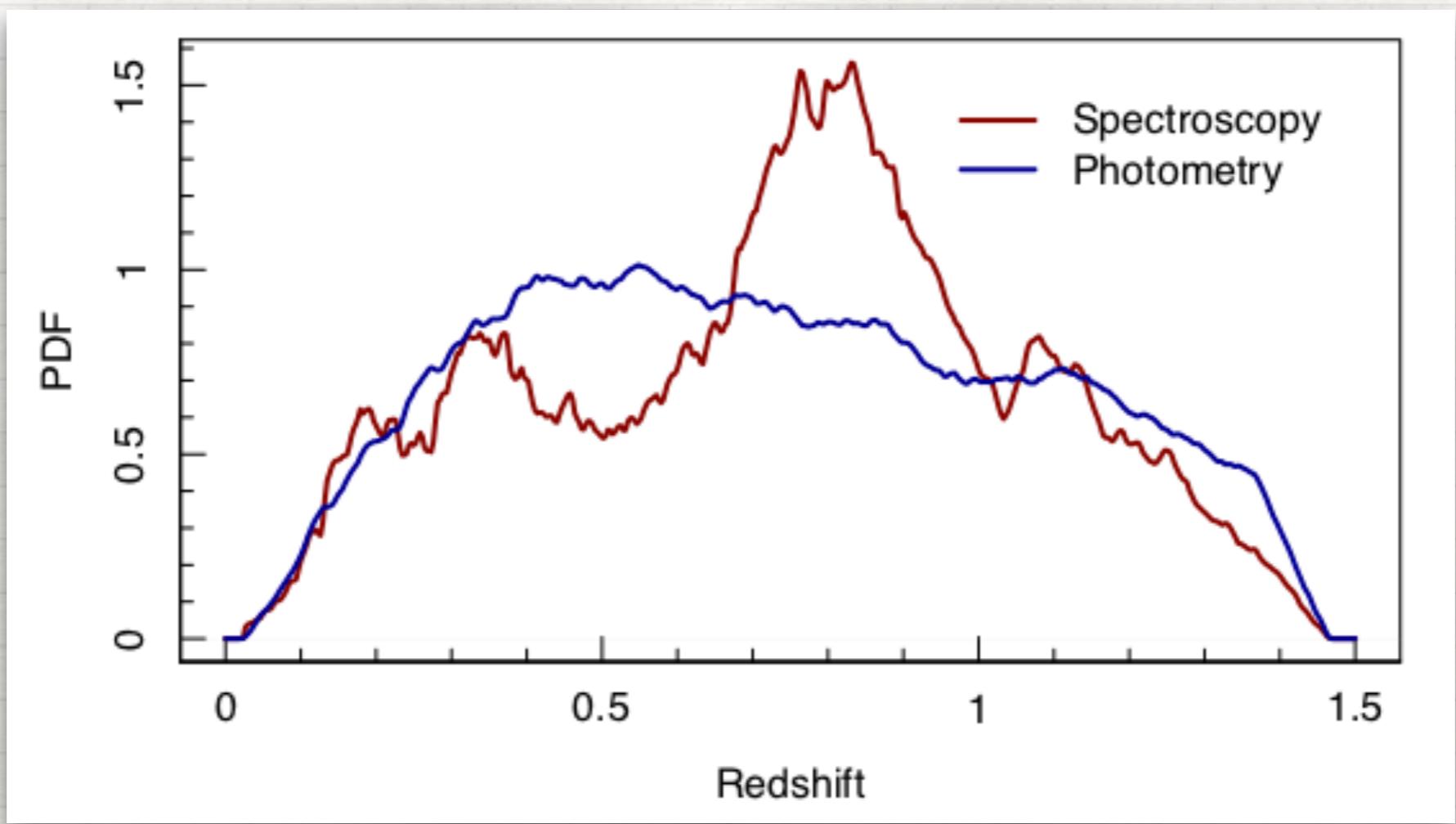


Photometric redshift point estimates correlate with, but are not the same as, true redshift.

↳ These are not sufficiently accurate for cosmic shear.

# DIRECT REDSHIFT CALIBRATION FOR COSMIC SHEAR

## SPECTROSCOPIC CALIBRATION

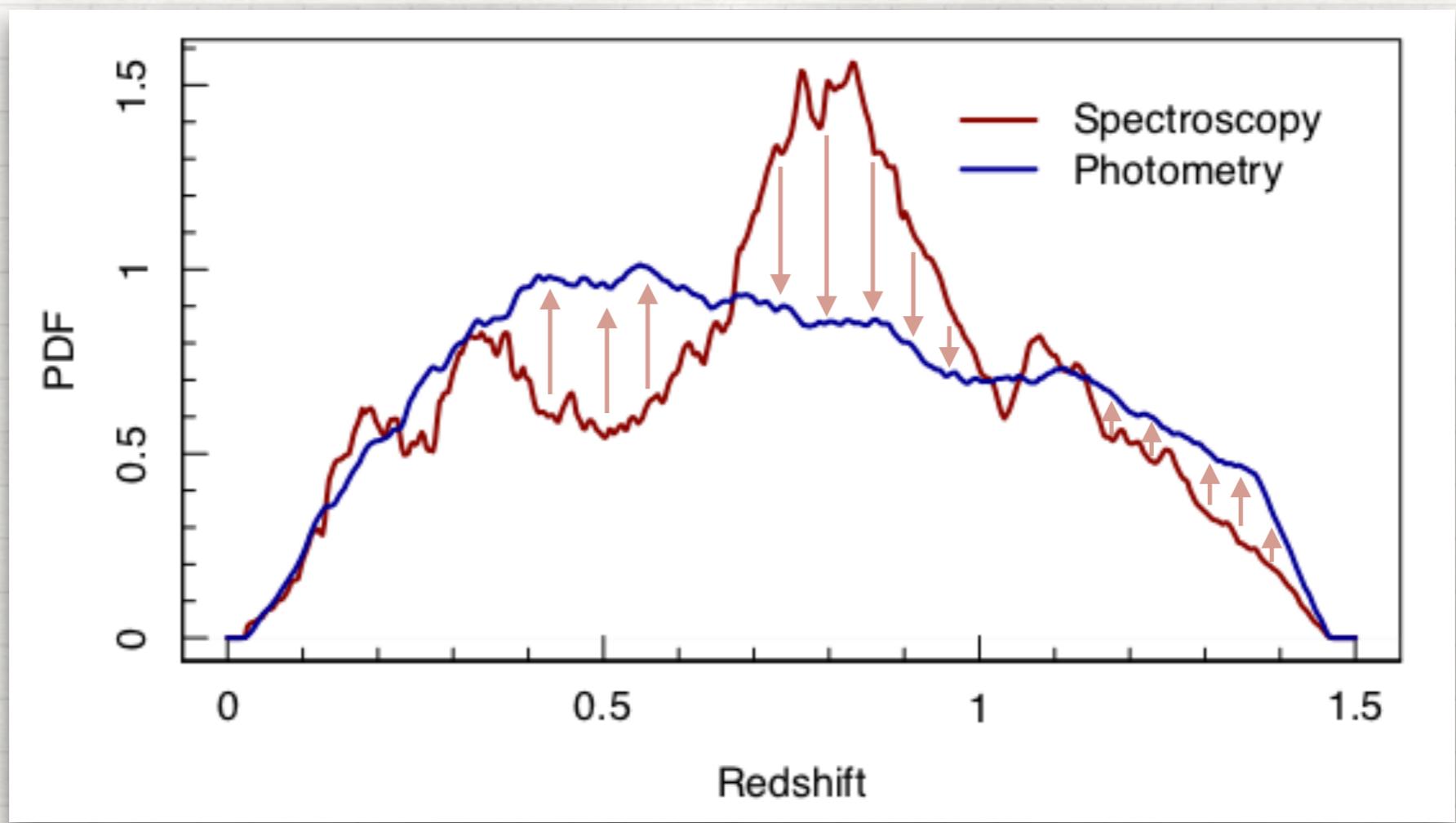


One calibration option is to use (high accuracy) spectroscopic redshifts to construct redshift distributions for the cosmic-shear (i.e. wide-field photometric) galaxies.

- ↳ Requires a method of mapping spectra onto wide-field galaxies
- ↳ Requires a representative spec-z sample *or* a method for determining which photometric sources are matched

# DIRECT REDSHIFT CALIBRATION FOR COSMIC SHEAR

## SPECTROSCOPIC CALIBRATION

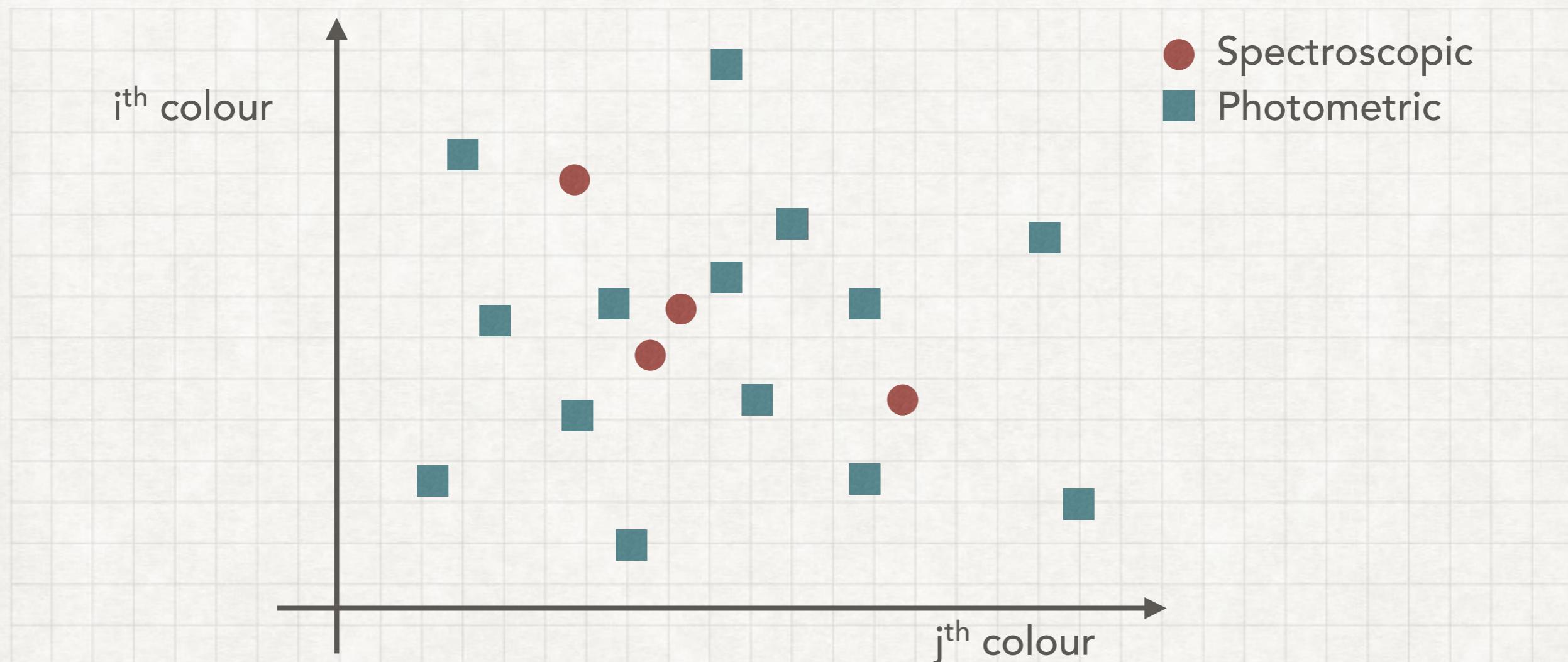


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# KIDS DIRECT REDSHIFT CALIBRATION (DIR)

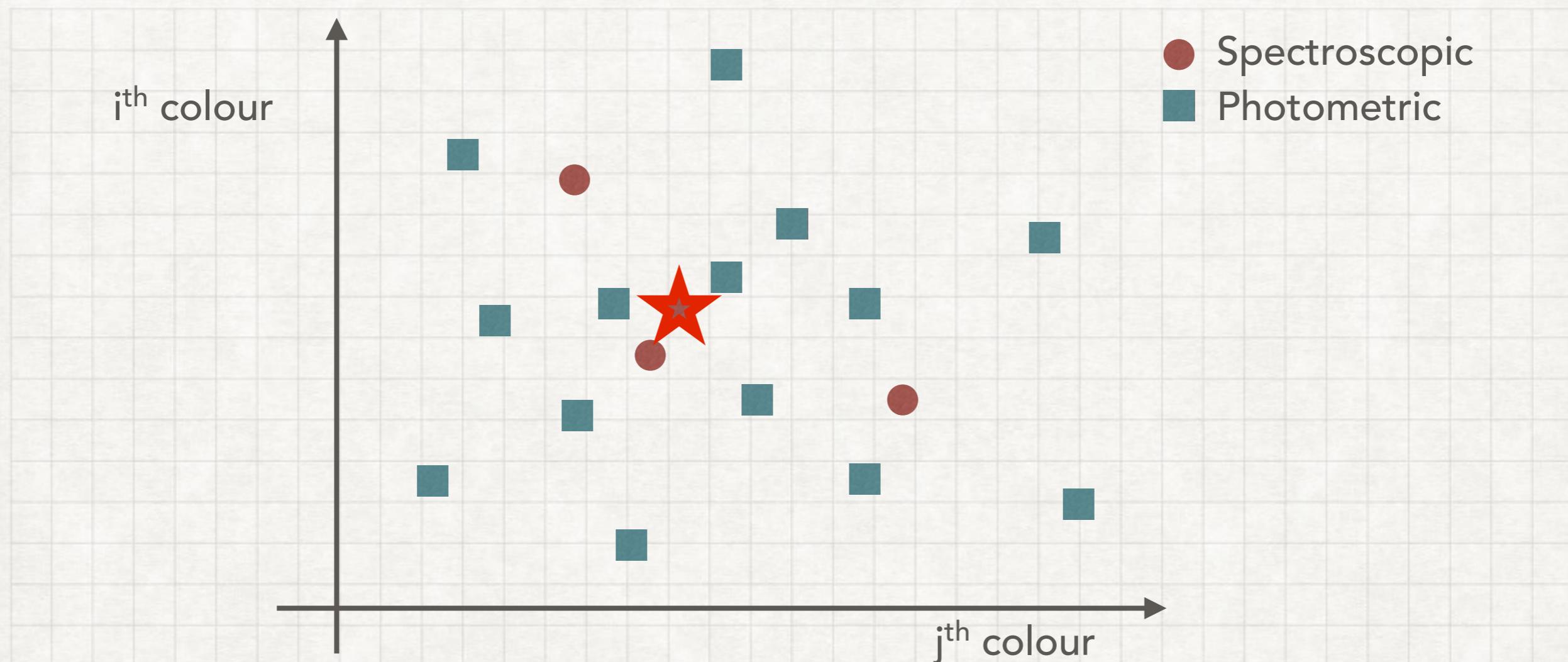
## K-NEAREST-NEIGHBOUR ASSOCIATION



The original implementation of direct redshift calibration used kNN to associate spectroscopic and photometric galaxies  
(Lima et al 2008, Hildebrandt et al 2017, 2018)

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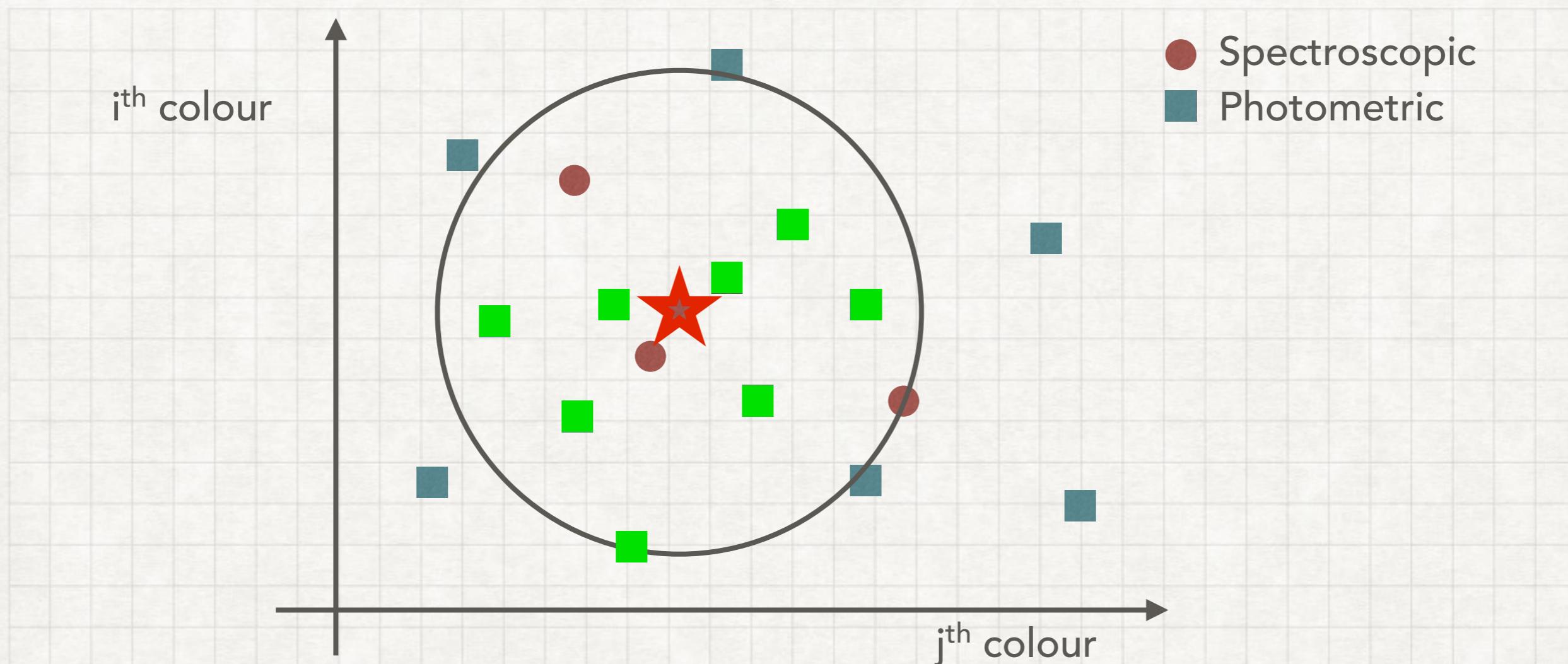
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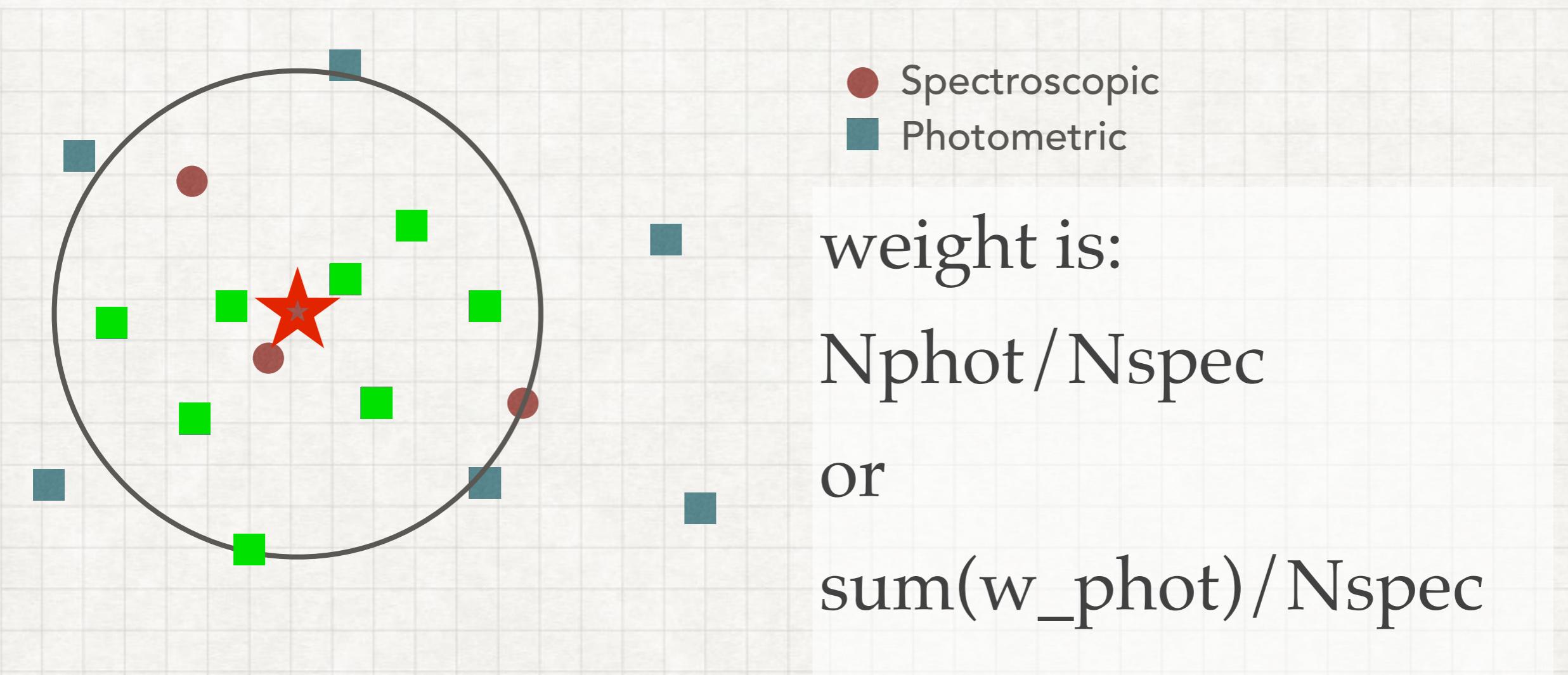
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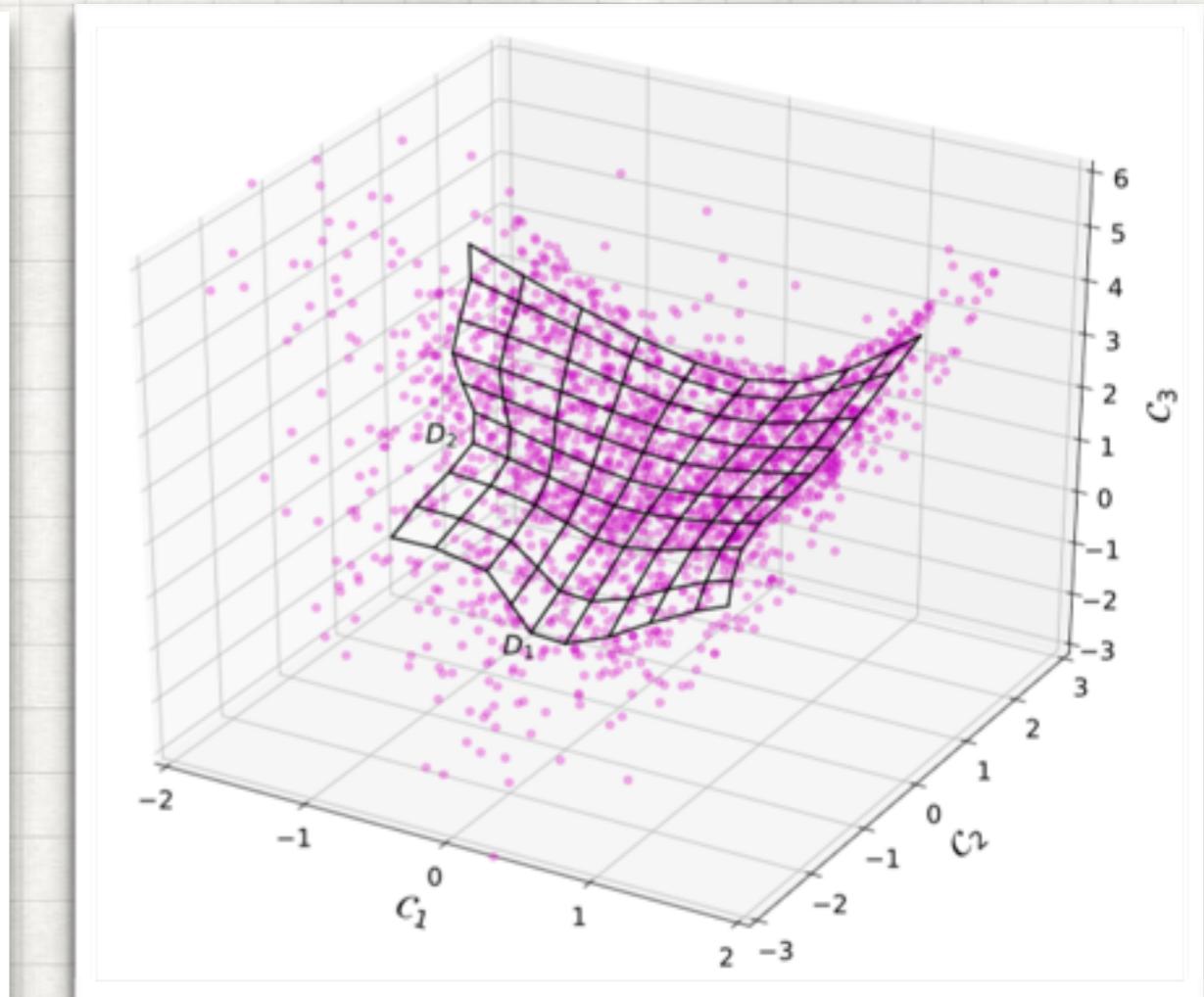
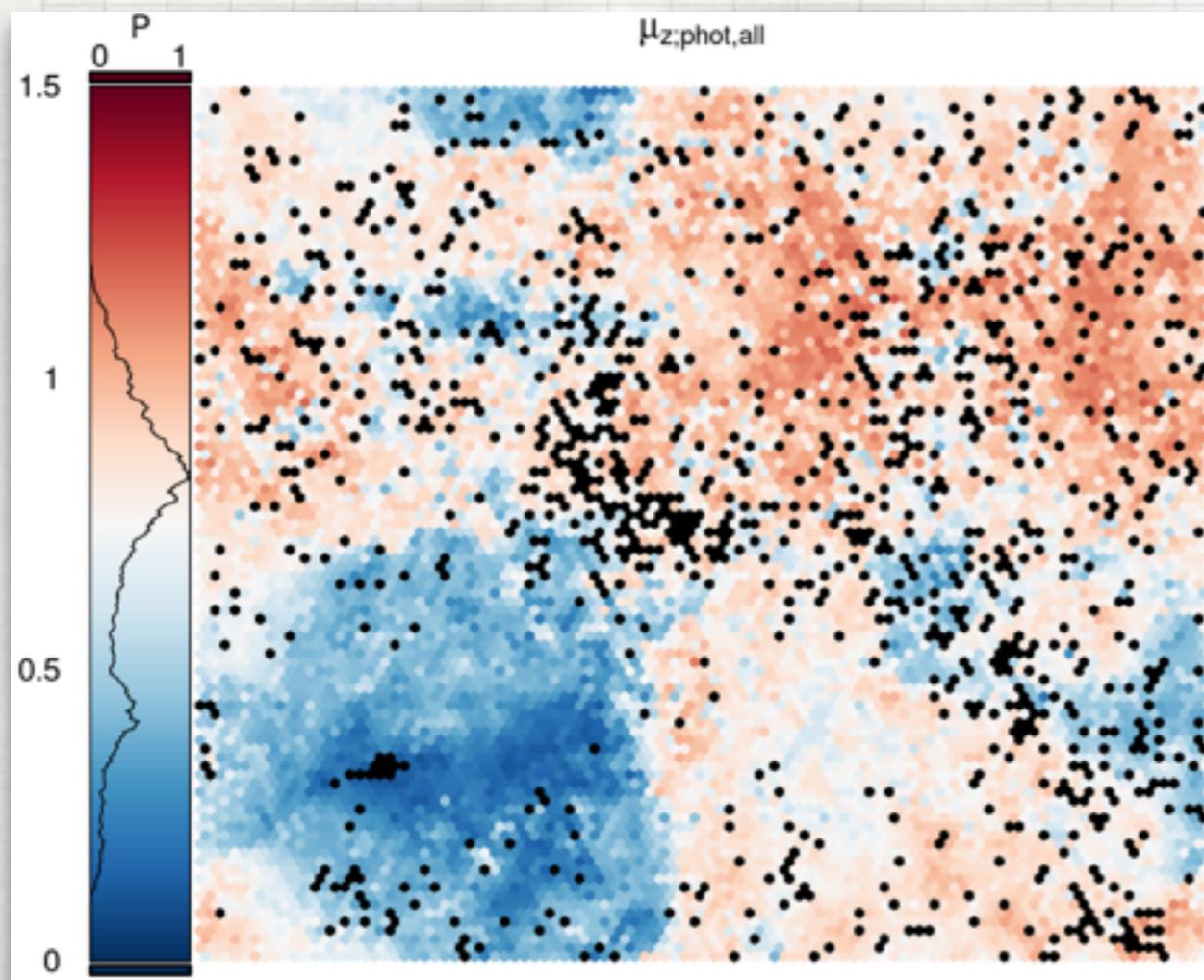


The original implementation of direct redshift calibration used kNN to associate spectroscopic and photometric galaxies  
(Lima et al 2008, Hildebrandt et al 2017, 2018)

# KIDS SOM REDSHIFT CALIBRATION (SOM)

## A NOVEL APPROACH

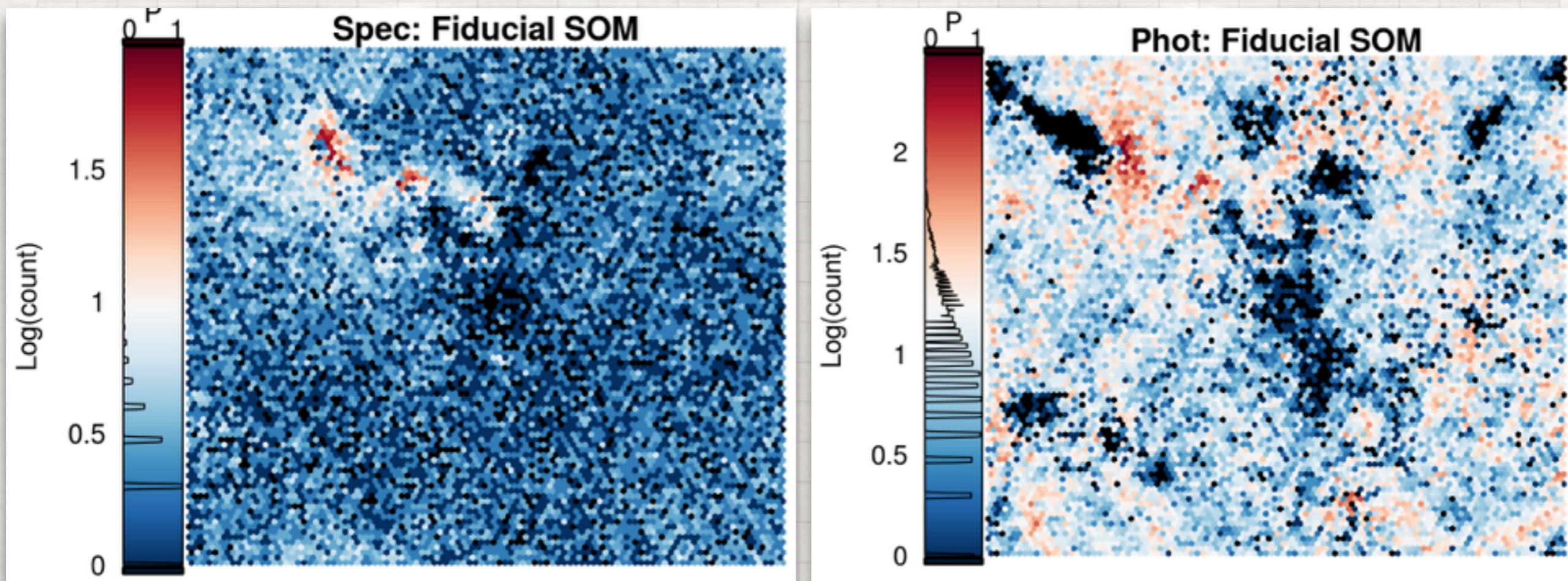
Self organising maps present another method of associating galaxies in distinct catalogues. The weighting mathematics remains the same, but the form of the association changes.



# REDSHIFT CALIBRATION WITH SELF-ORGANISING MAPS

## A NOVEL APPROACH

Associations are based on the self-similarity between the sources in respect to the colour-colour manifold. Allows variable Nspec.  
Number of neighbours choice is converted to SOM pixel choice.

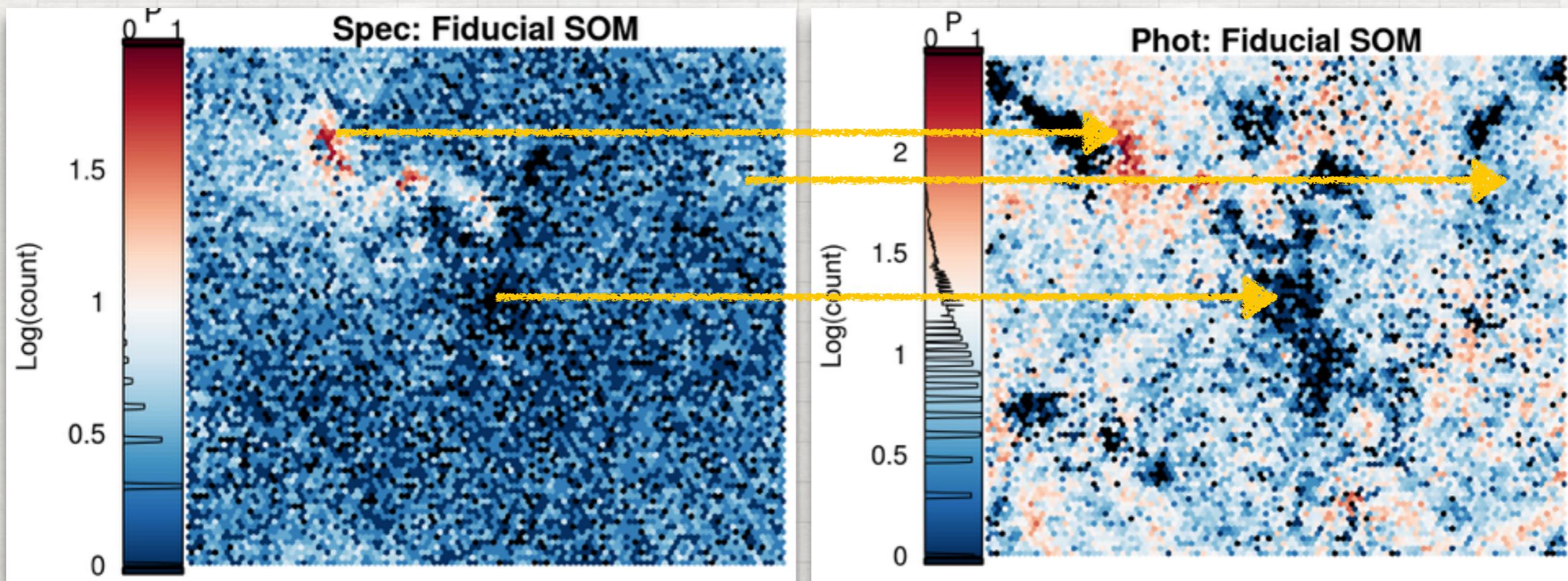


Direct knowledge of cells with shear sources and no spec-z

# REDSHIFT CALIBRATION WITH SELF-ORGANISING MAPS

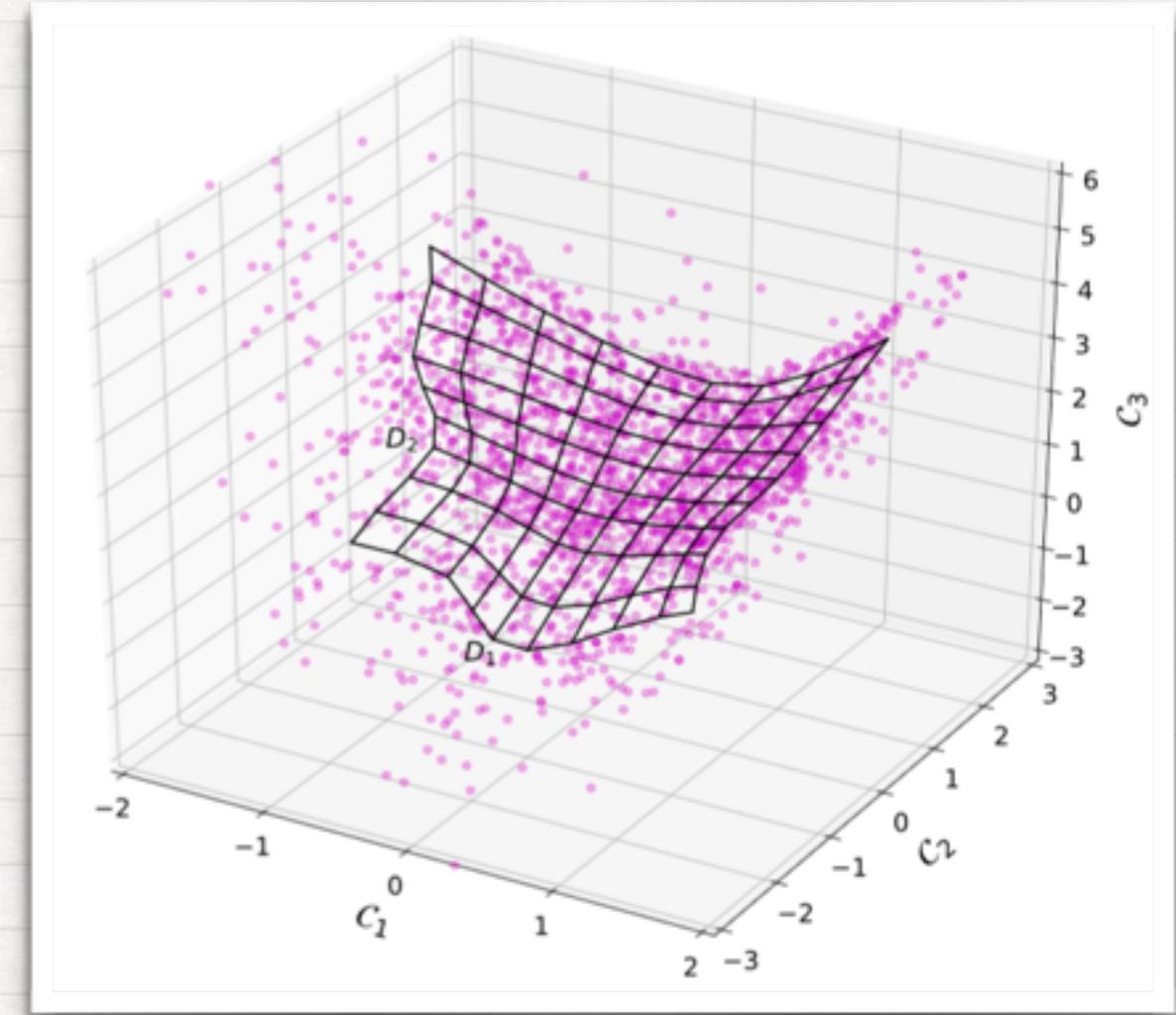
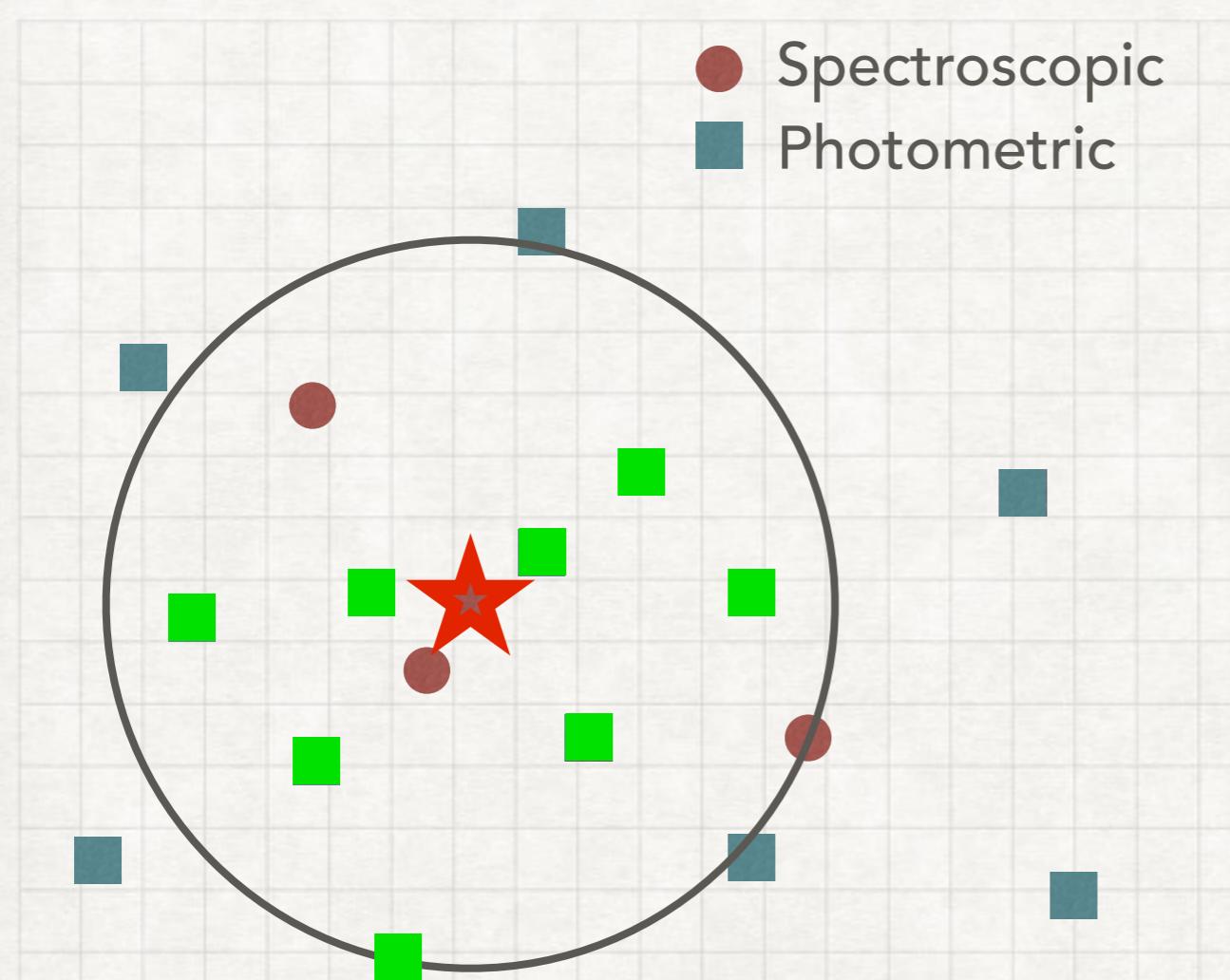
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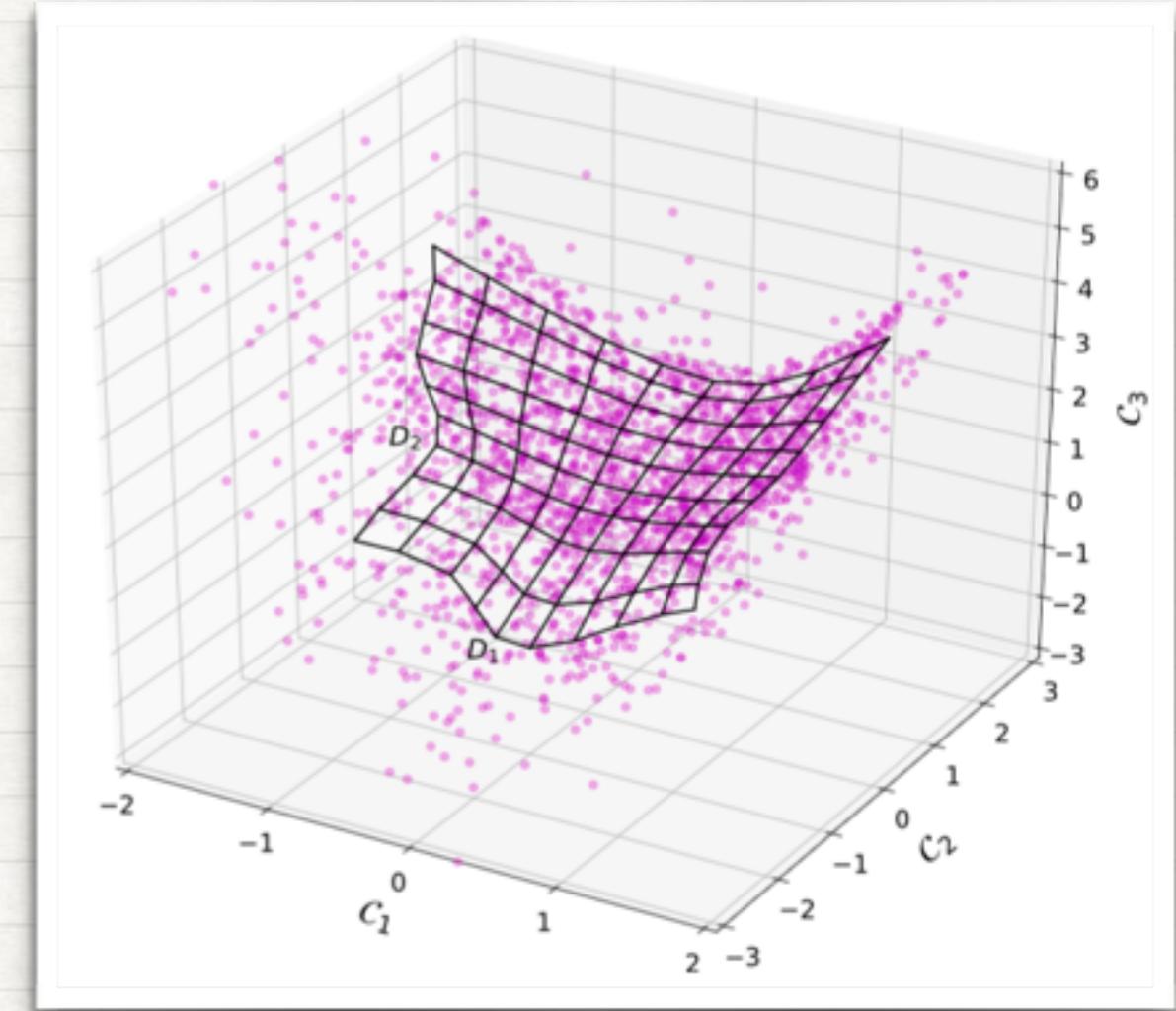
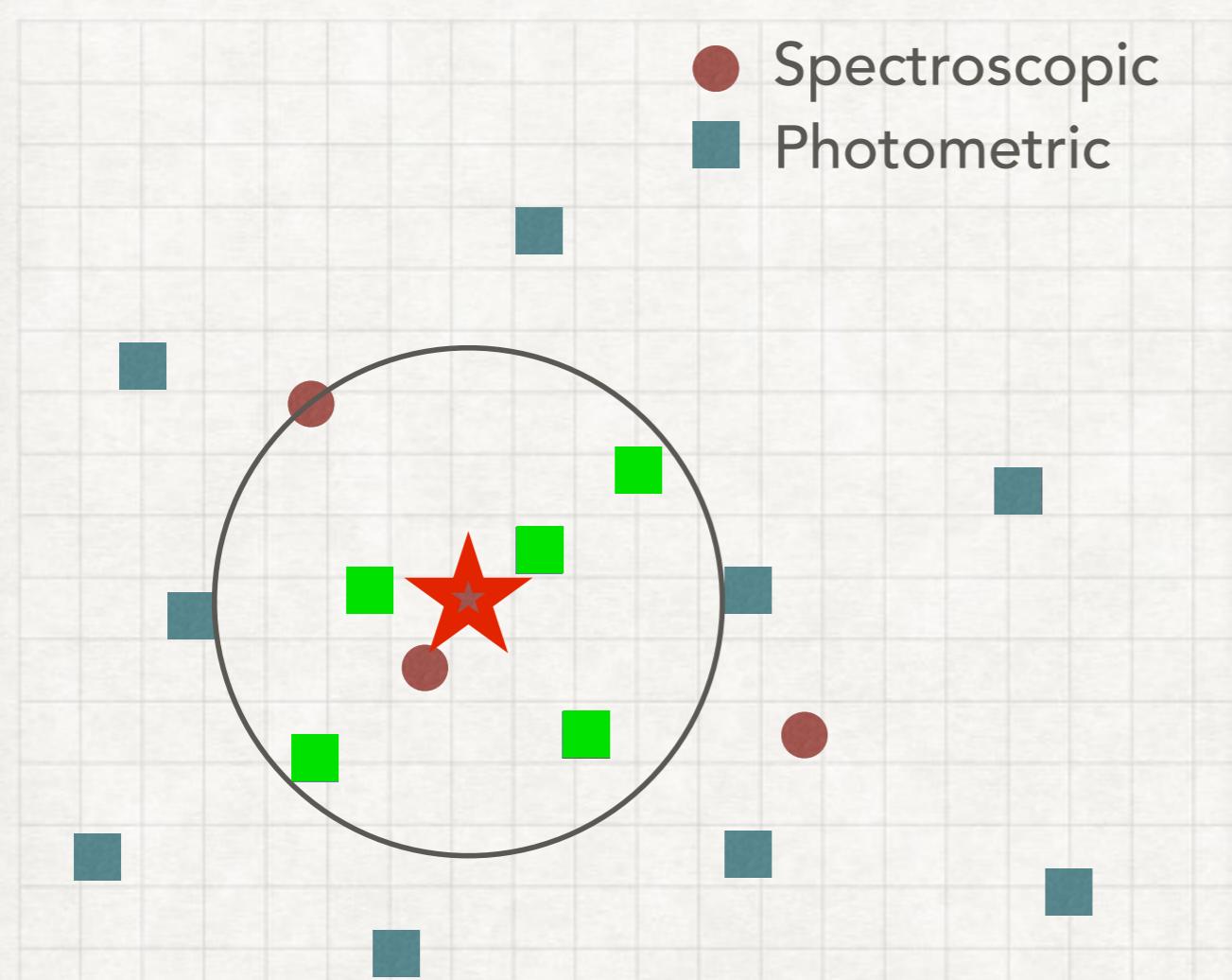
# CHOICE OF NEIGHBOUR DISTANCE/SOM PIXEL CLUSTERING



The choice of pixel scale imposes limitation on the mapping of the hyper-surface.

↳ Clustering the pixel data avoids this while not biasing the calibration

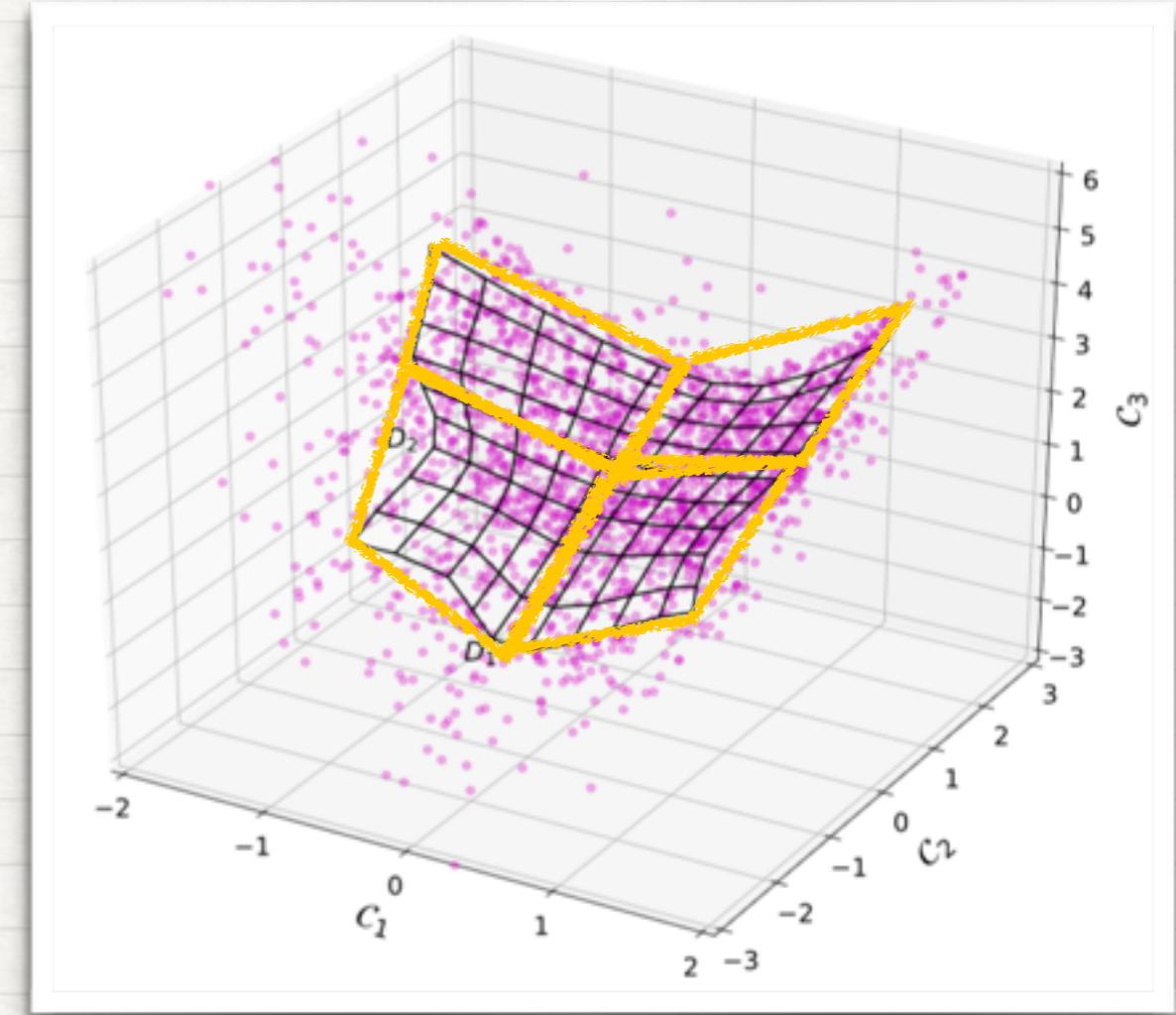
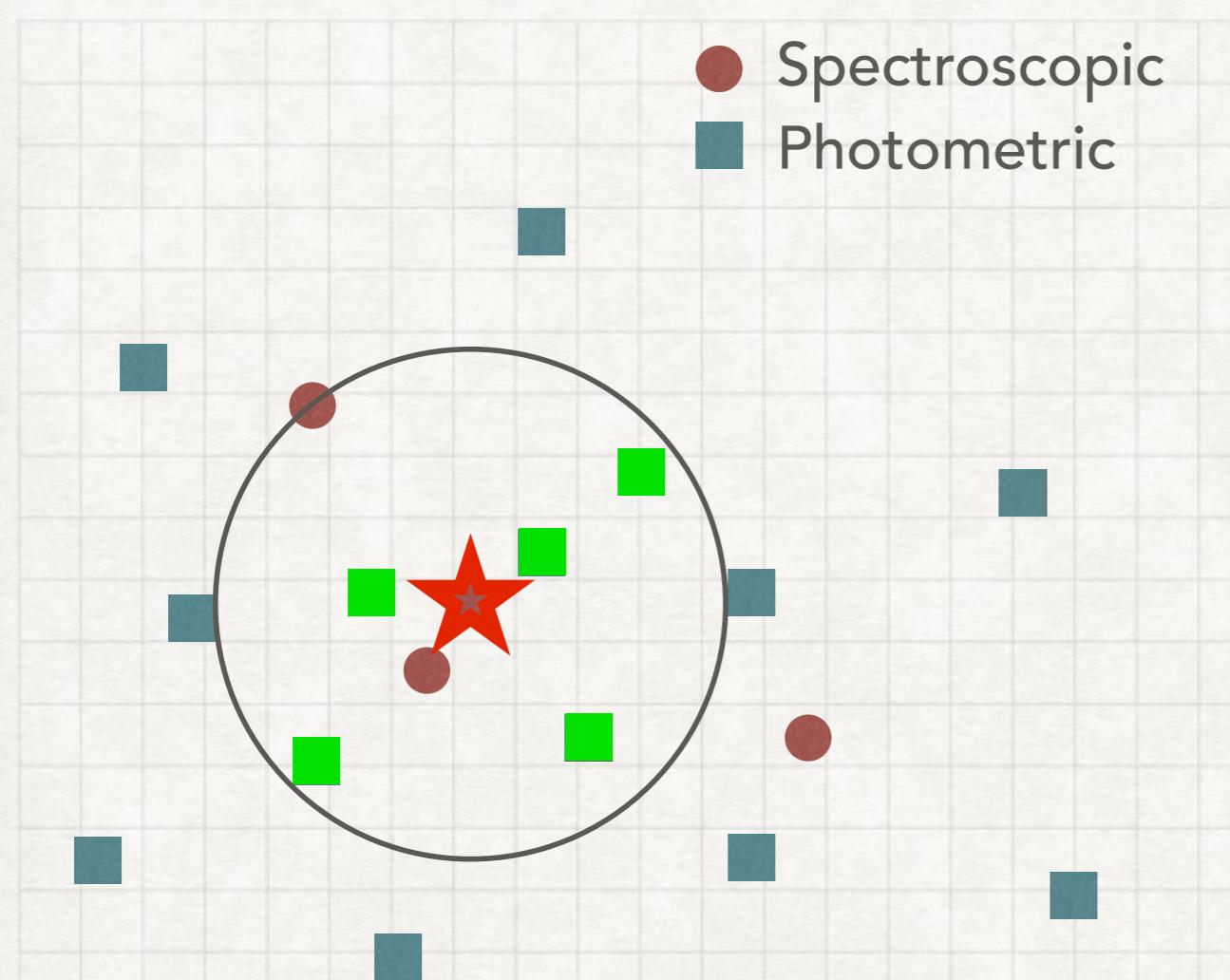
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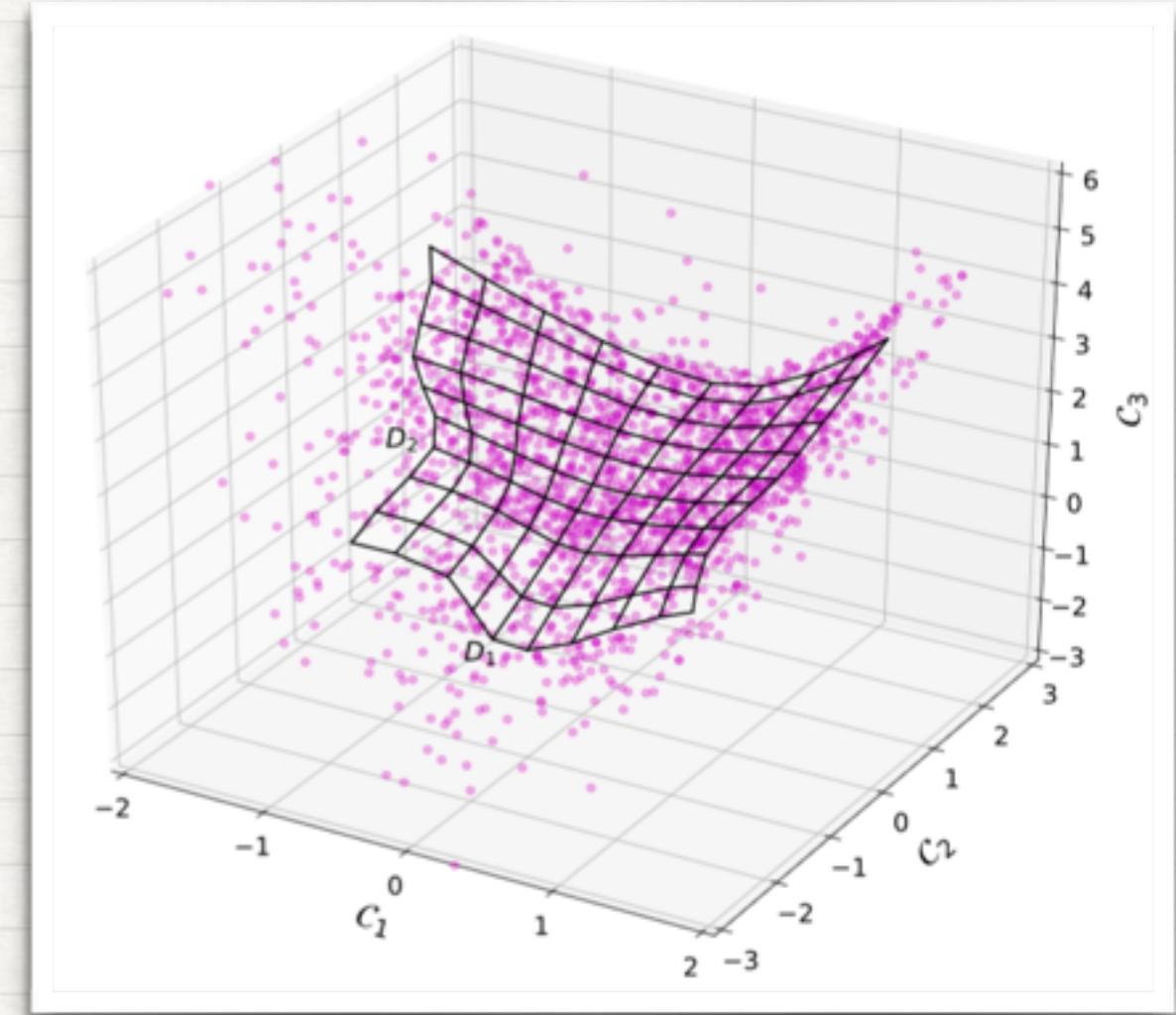
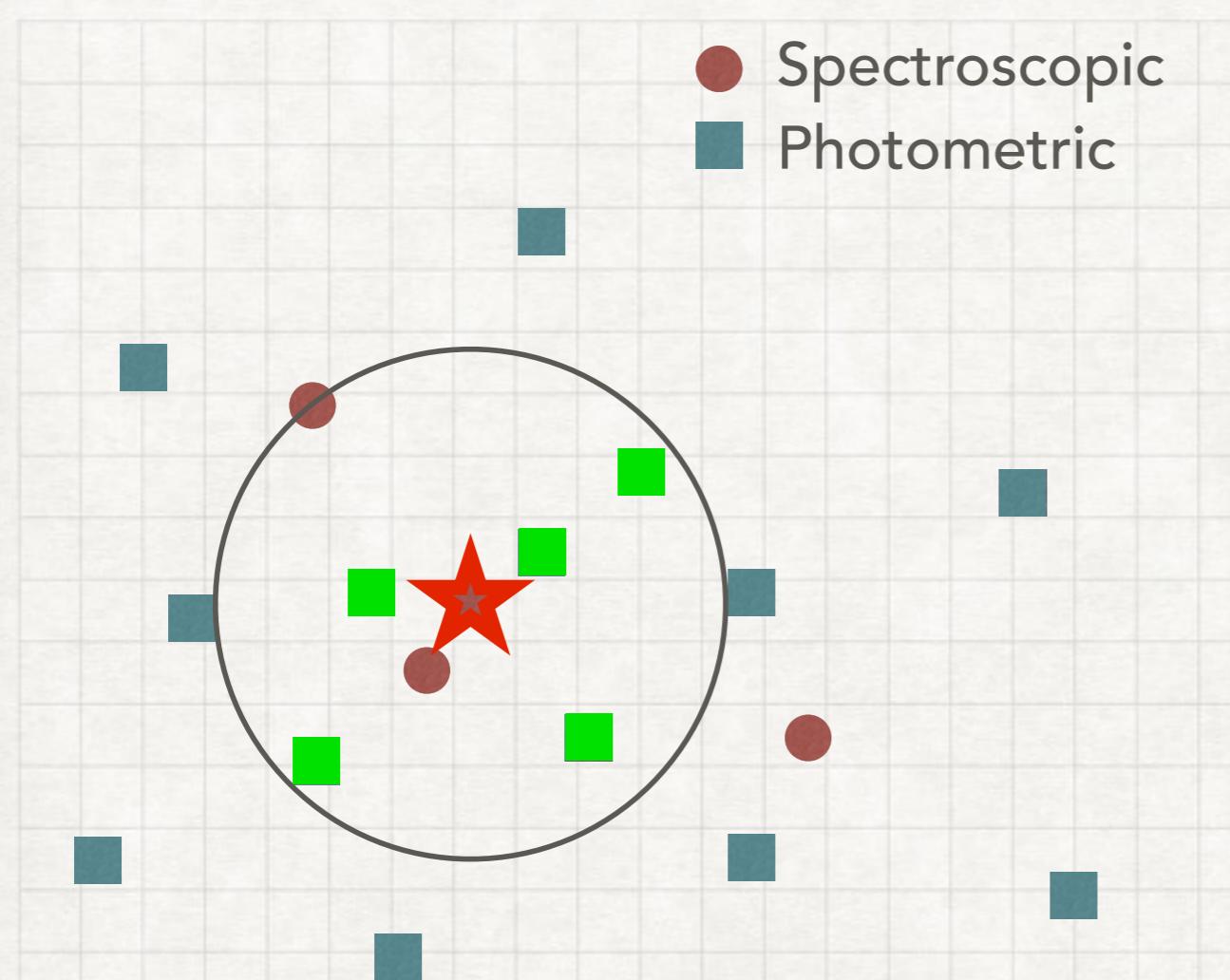
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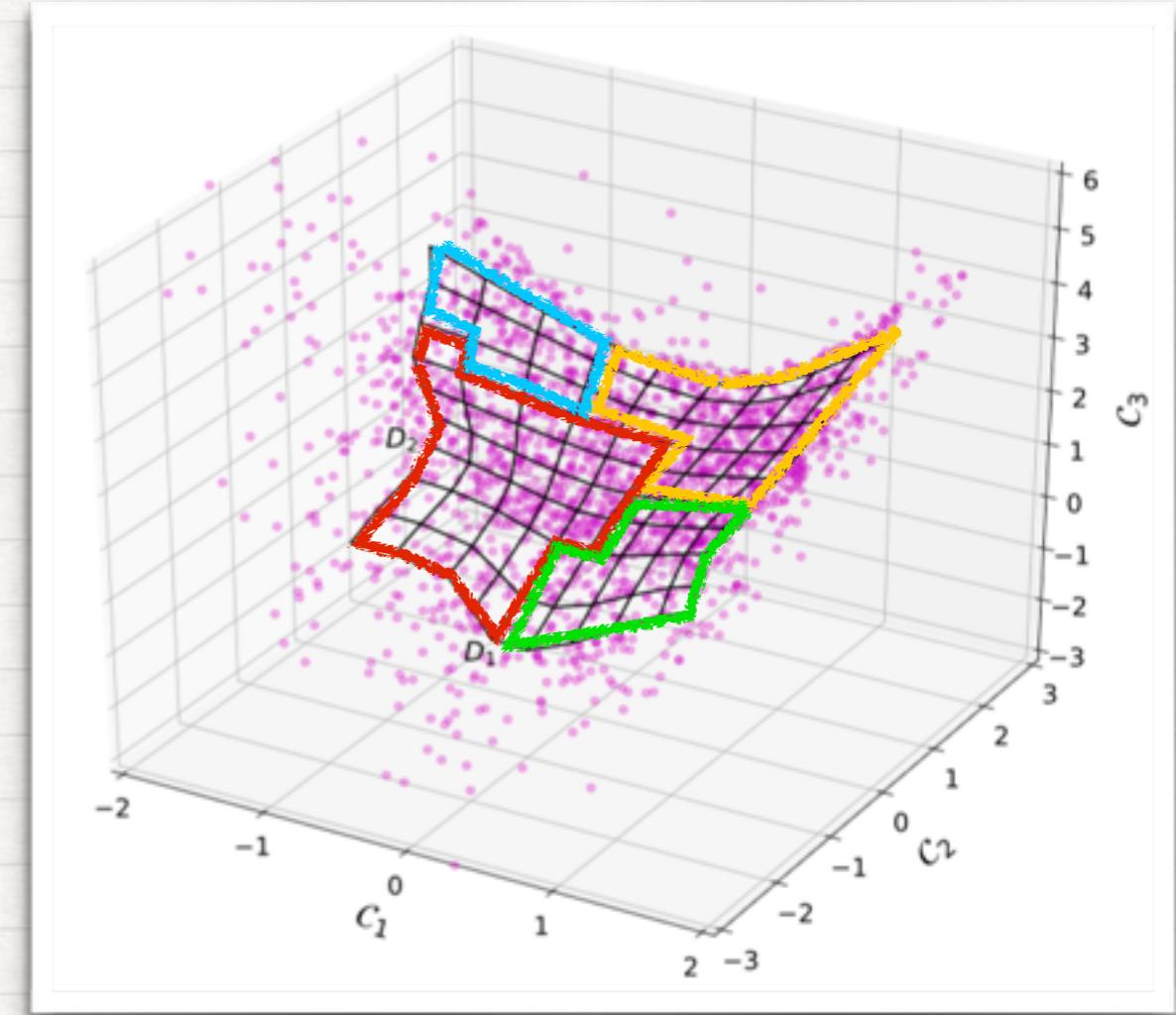
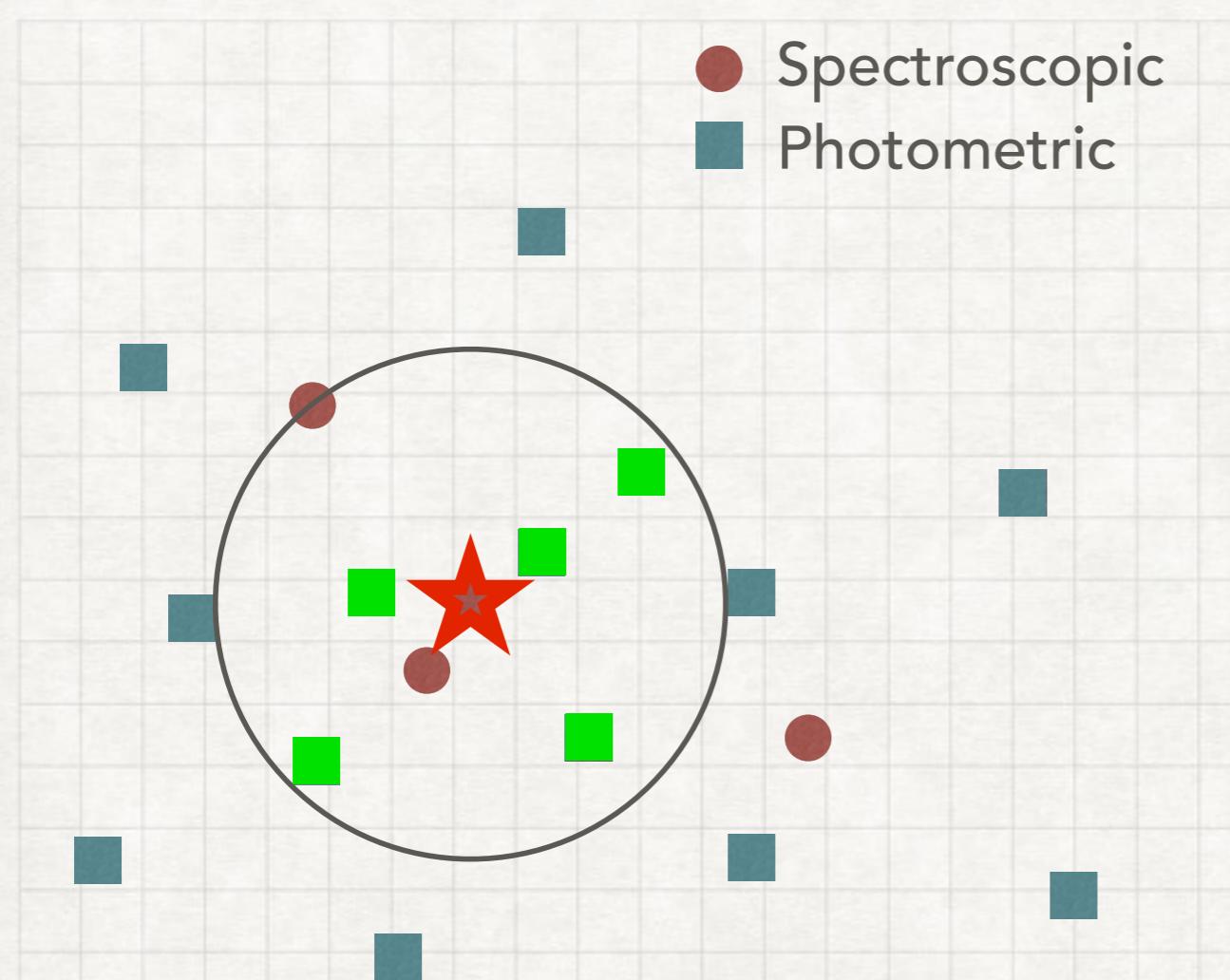
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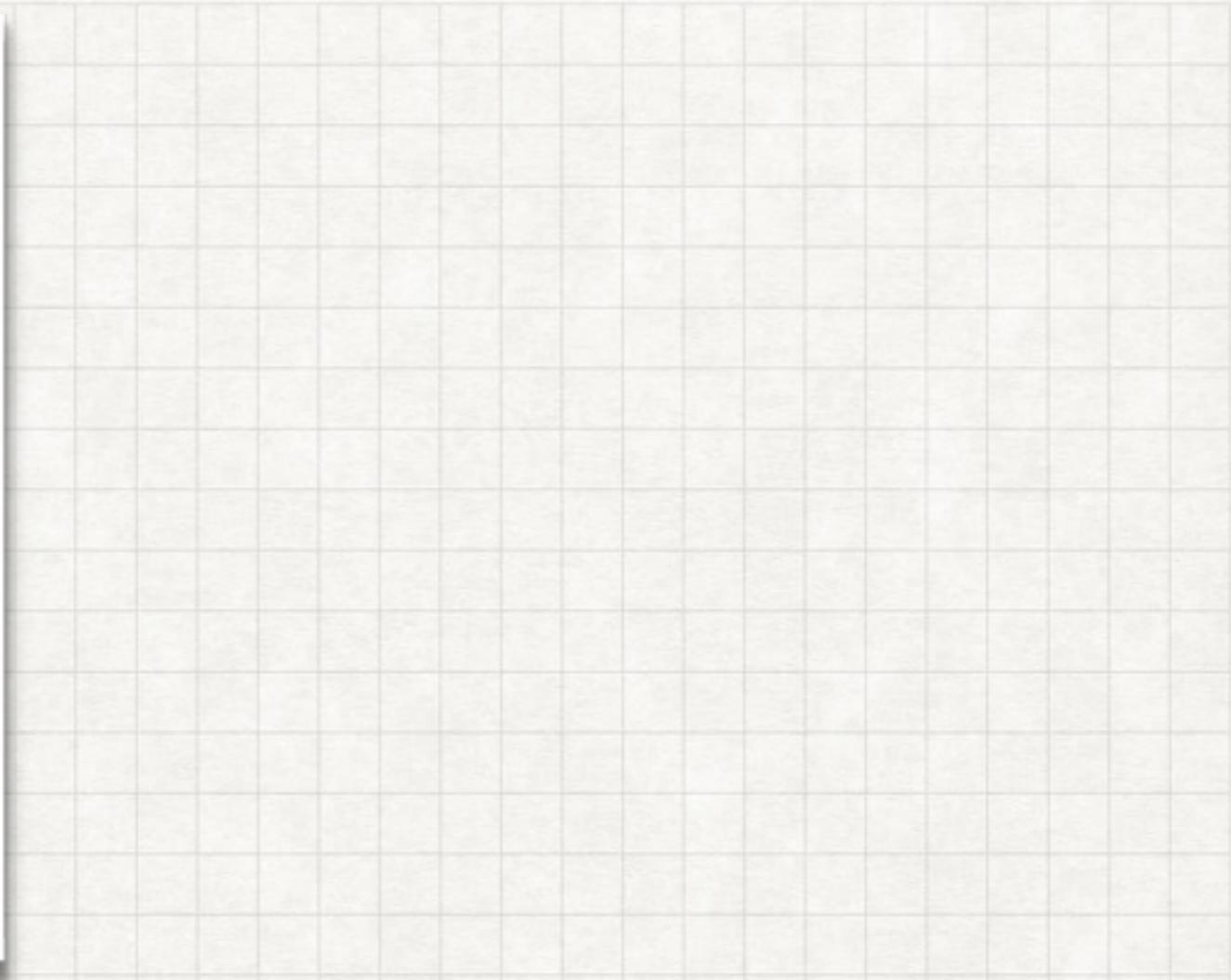
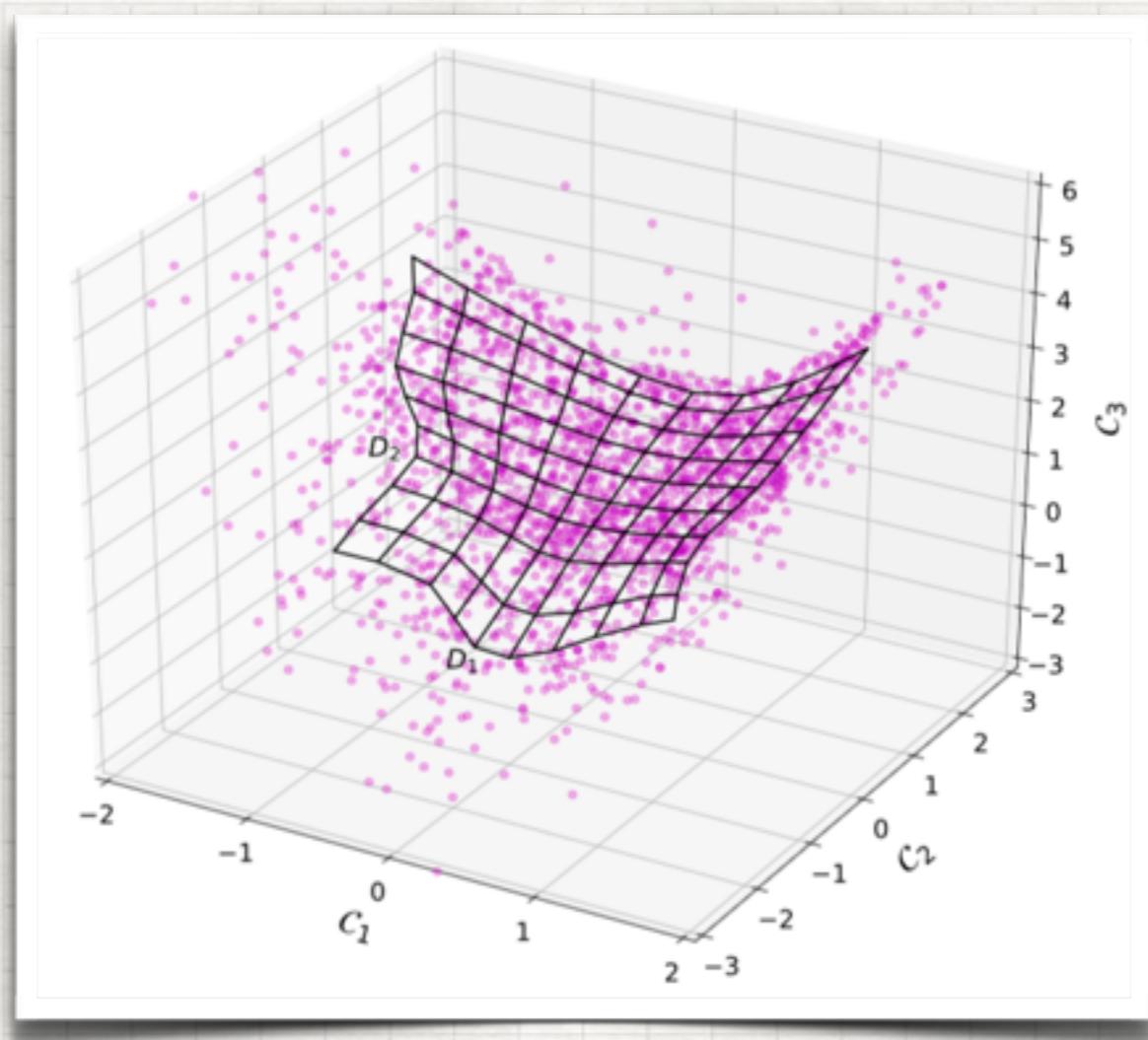
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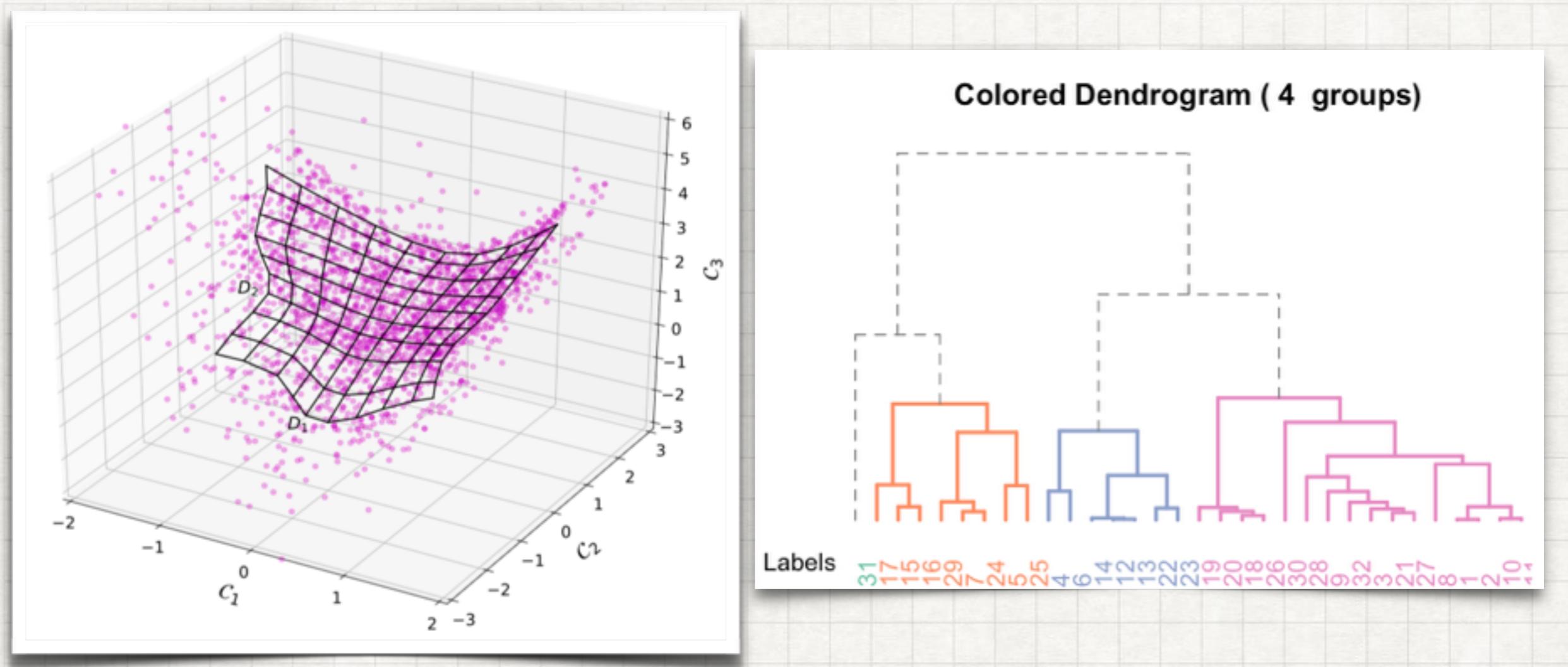
# CHOICE OF NEIGHBOUR DISTANCE/SOM PIXEL CLUSTERING



The choice of association number is then relevant.

↳ An arbitrary number of associations can be constructed using hierarchical clustering of the SOM cells

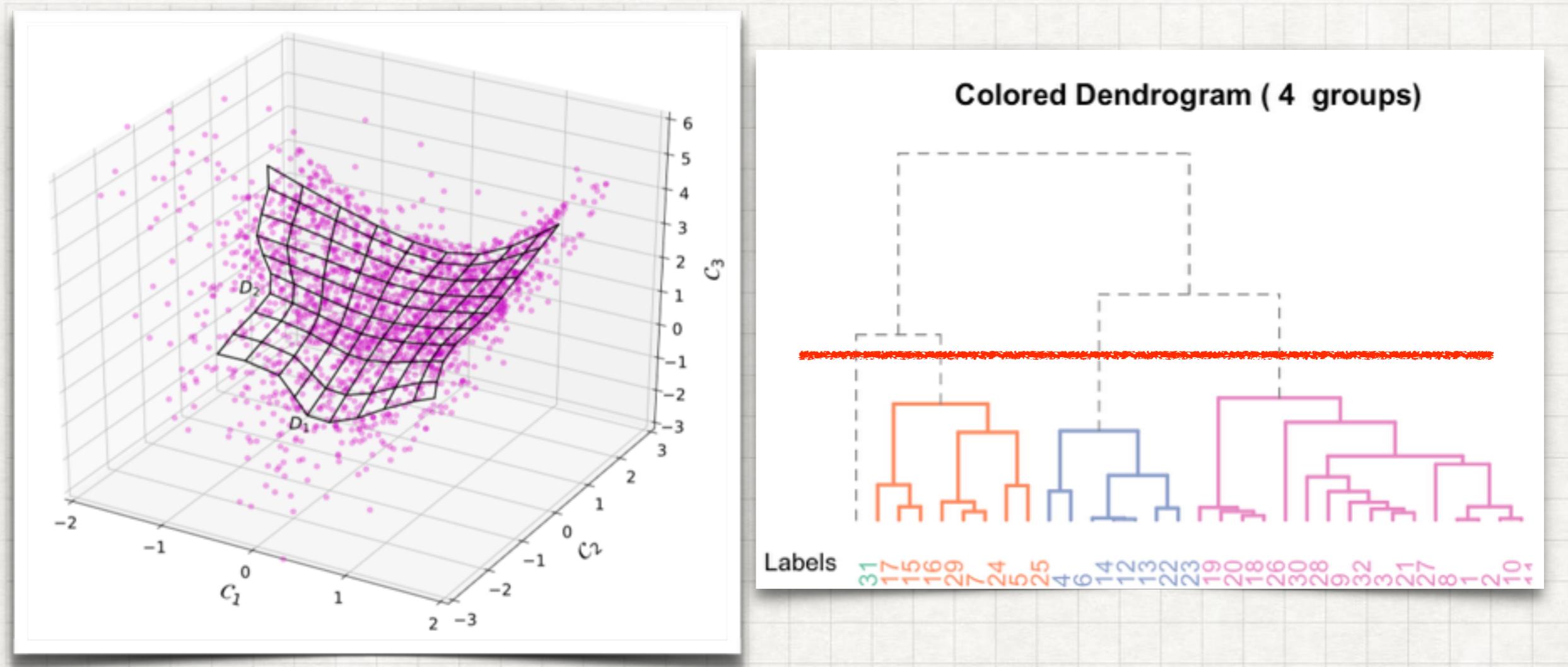
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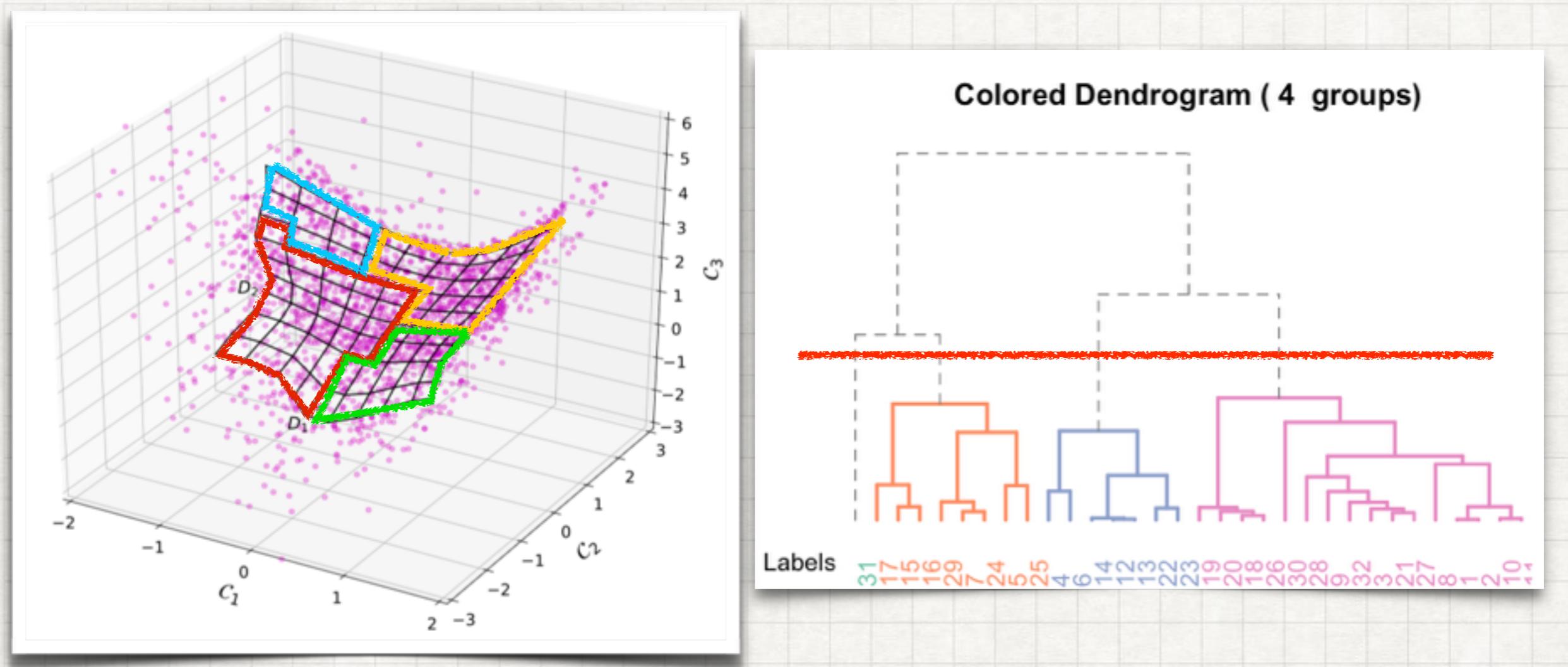
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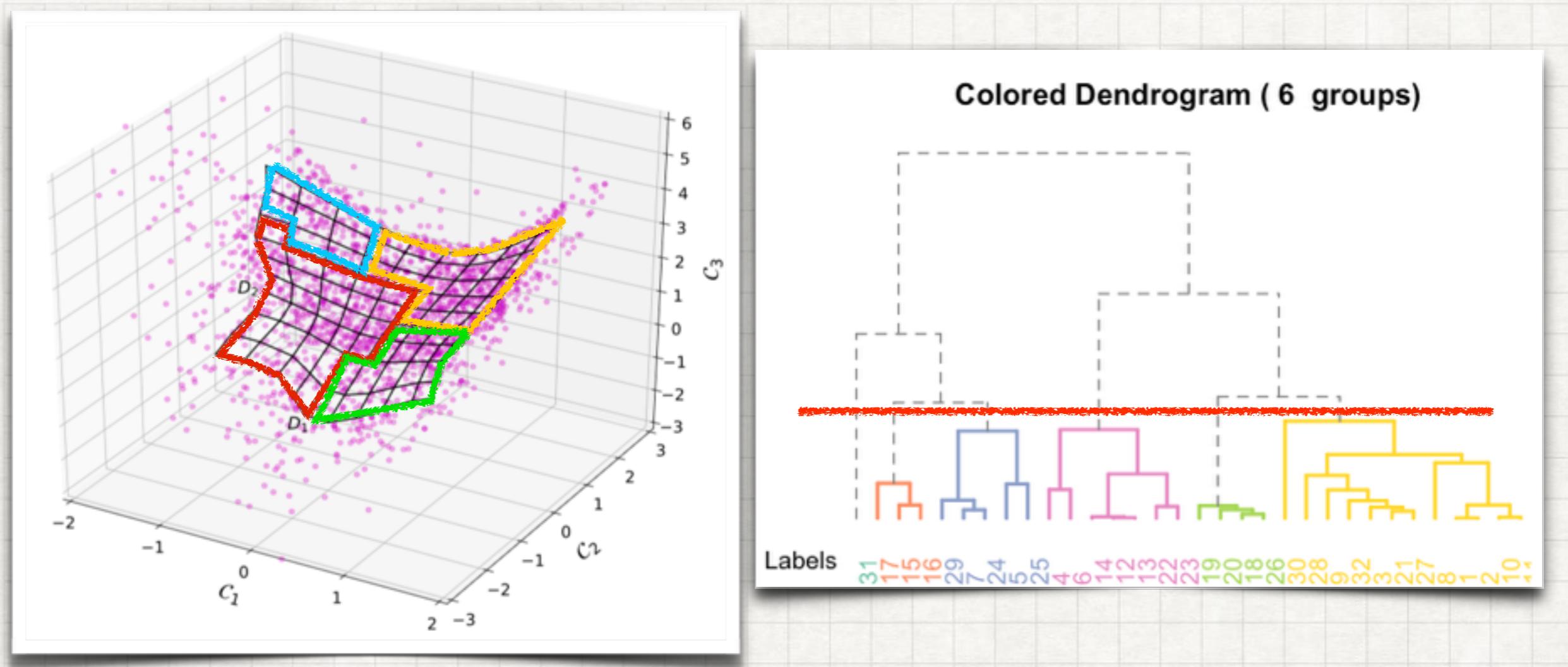
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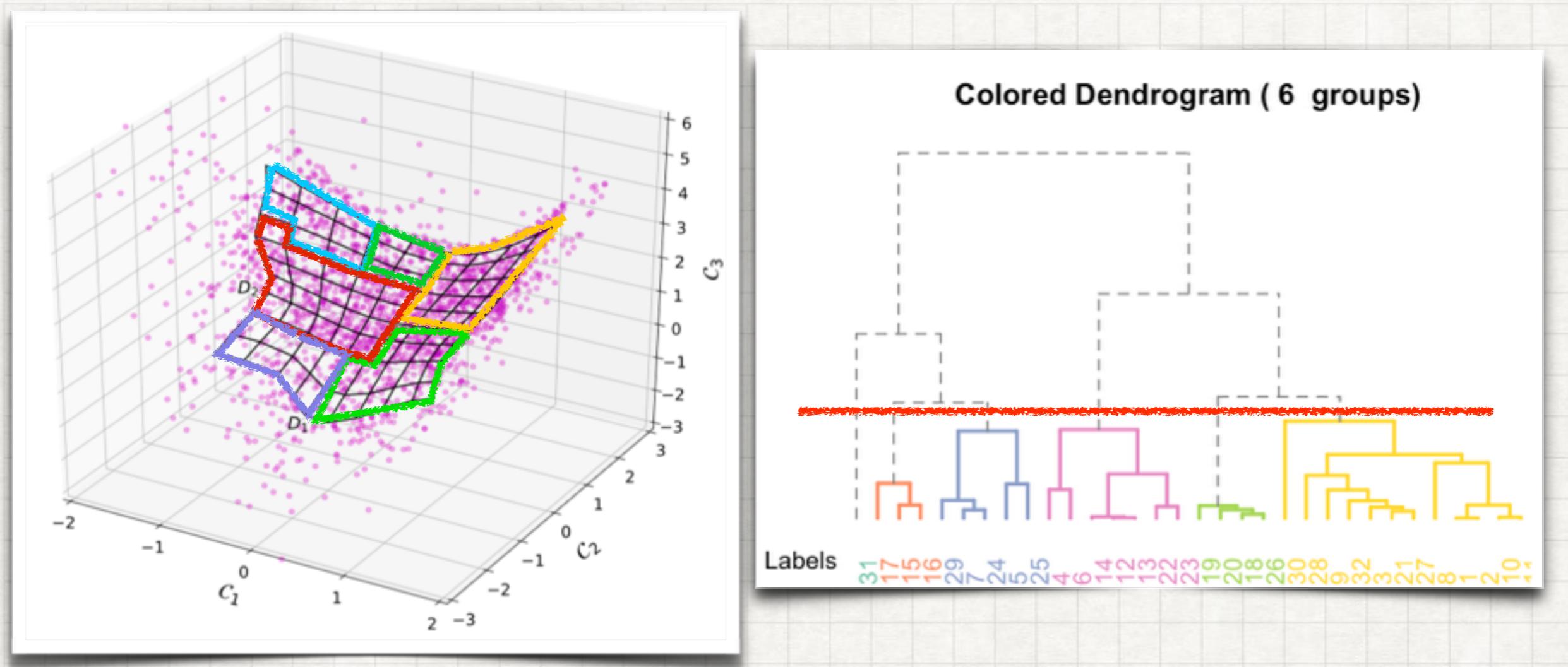
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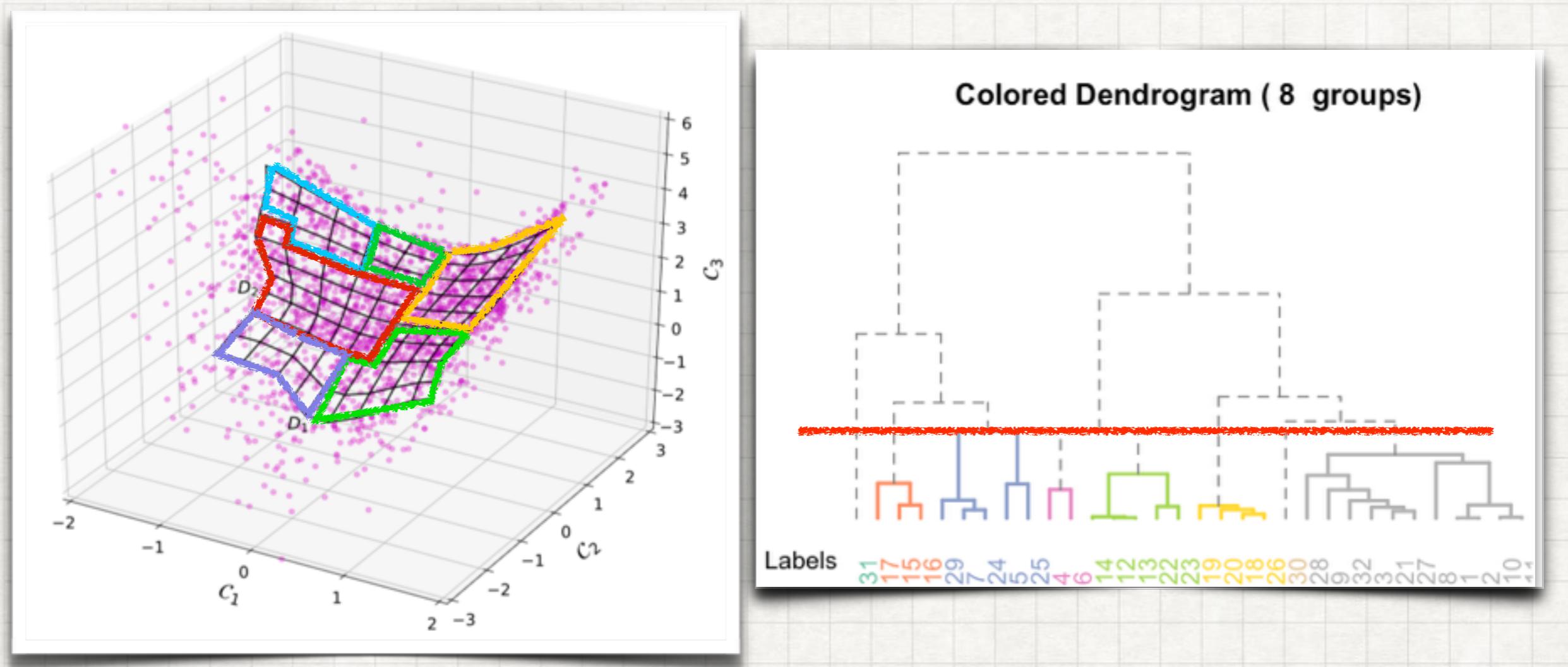
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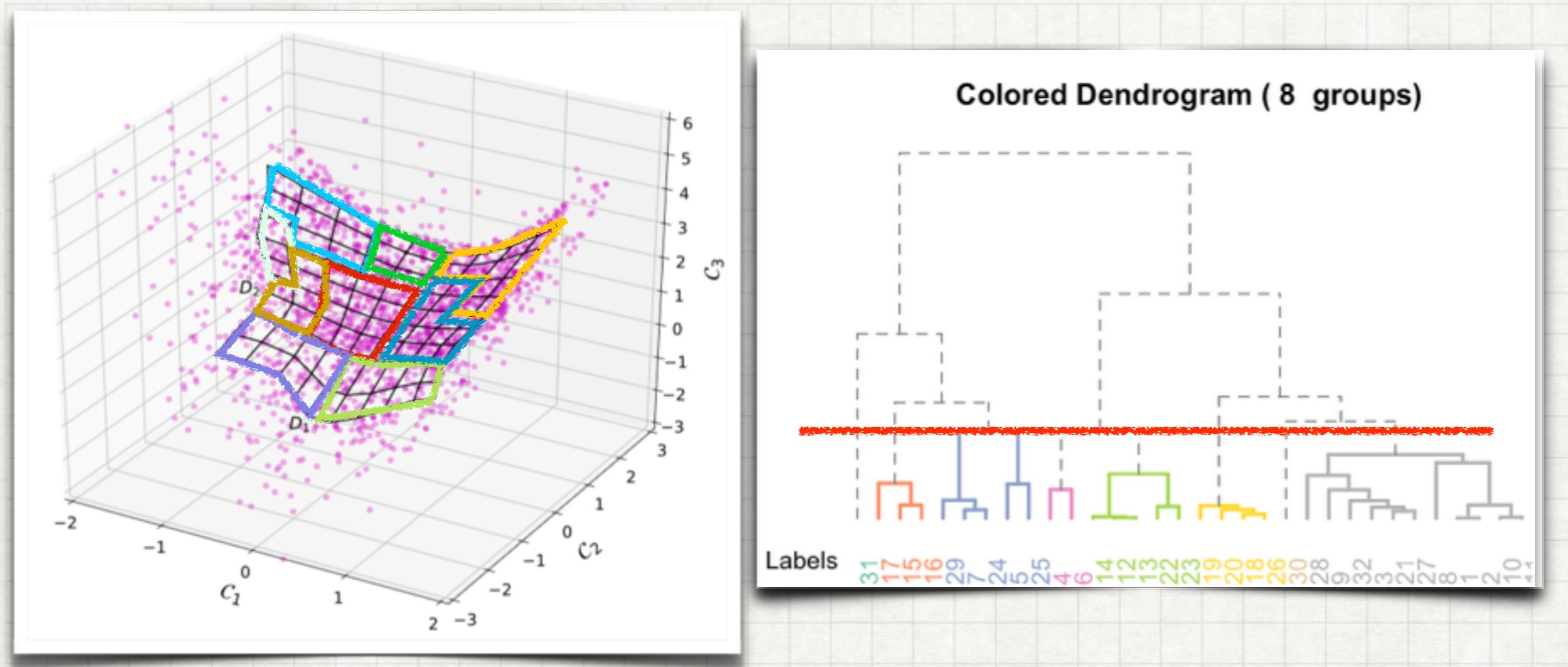
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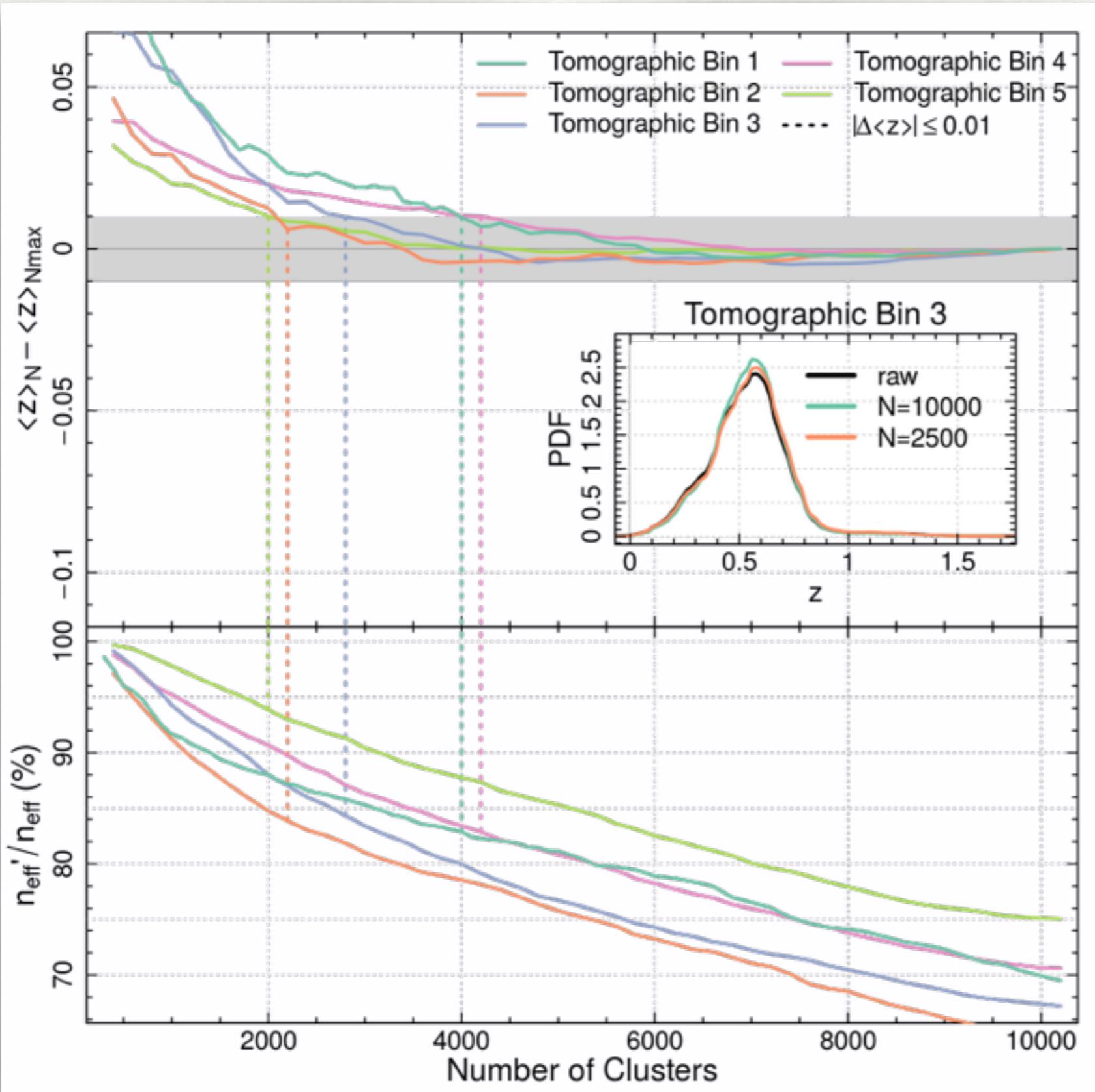
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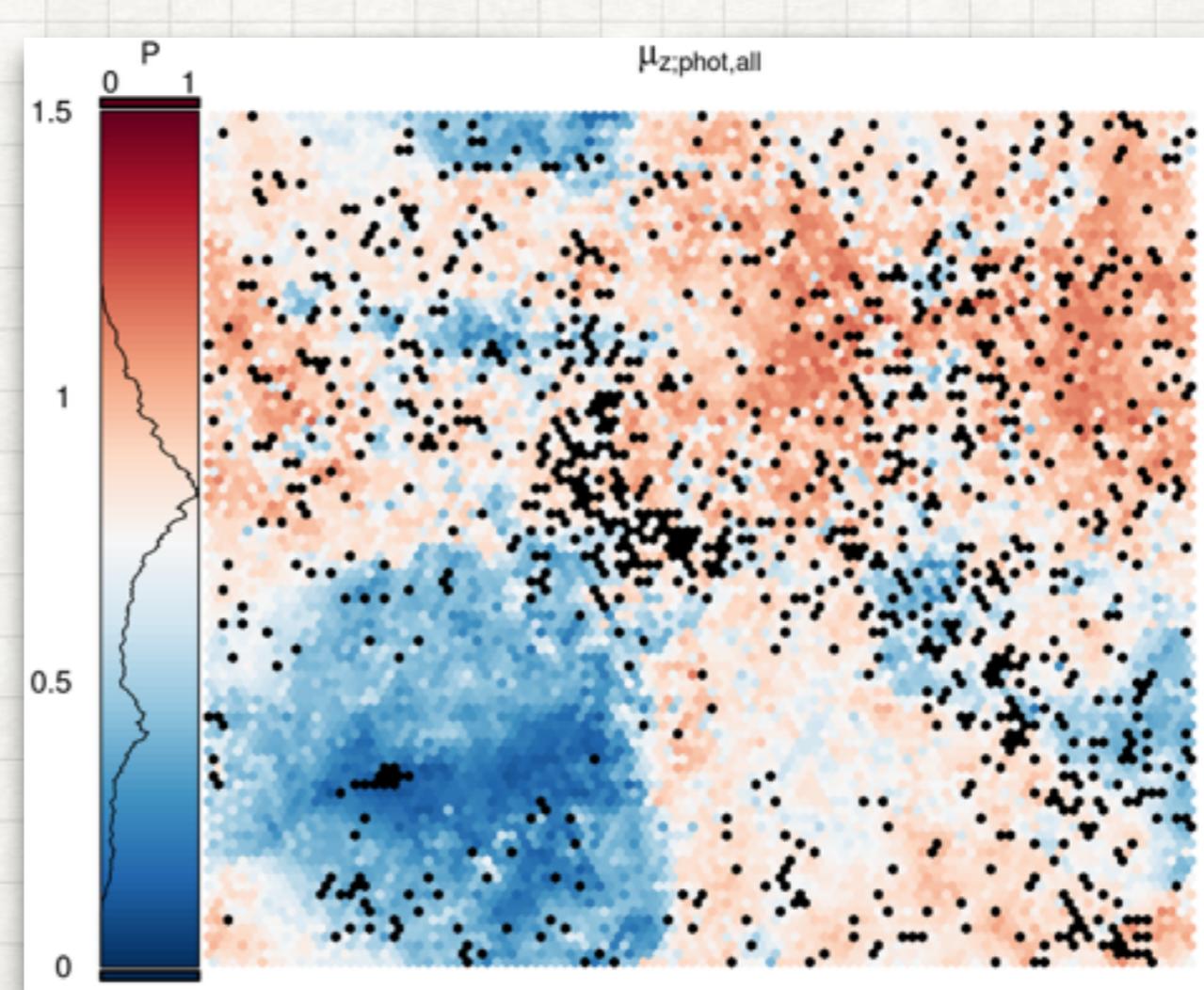
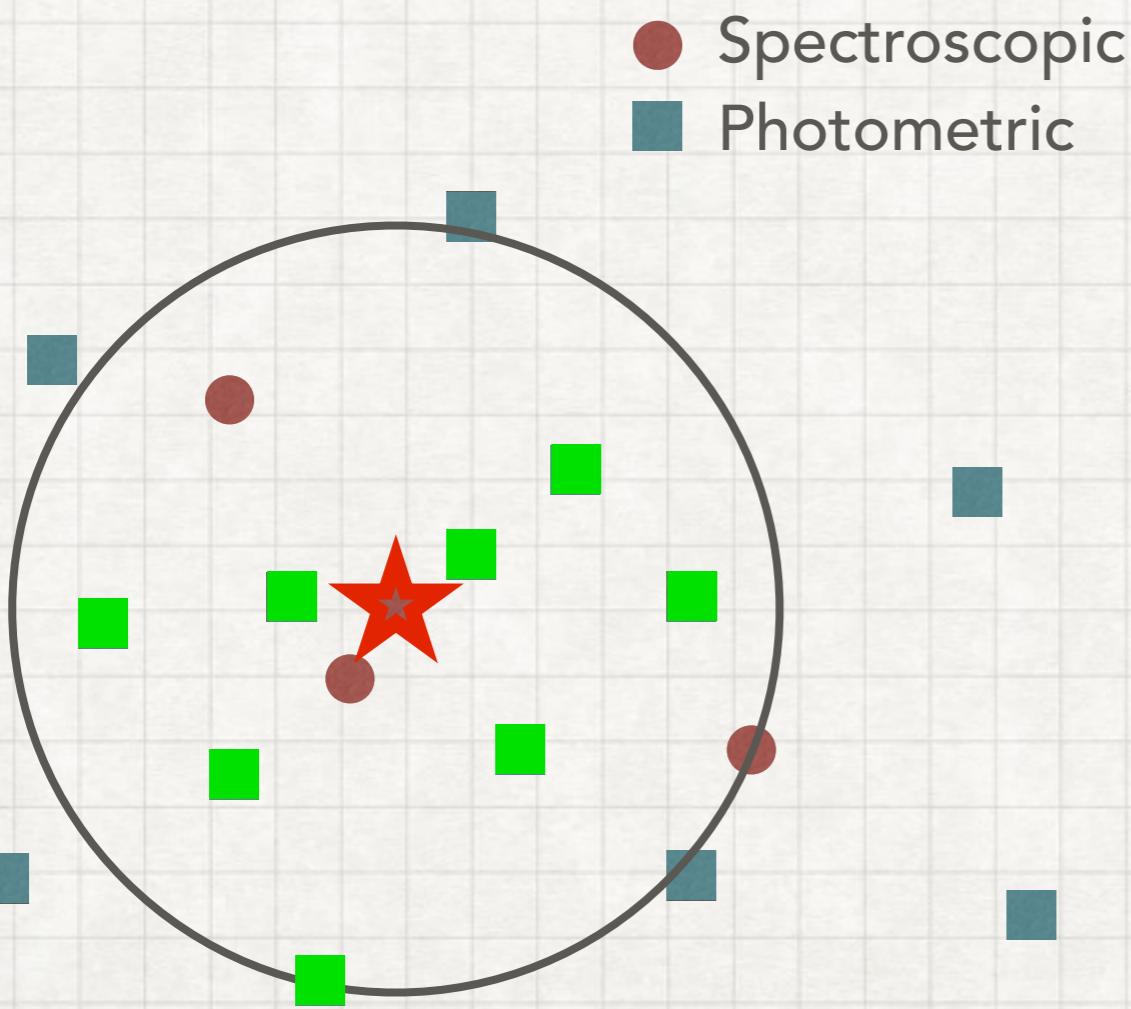
# SOM PIXEL CLUSTERING



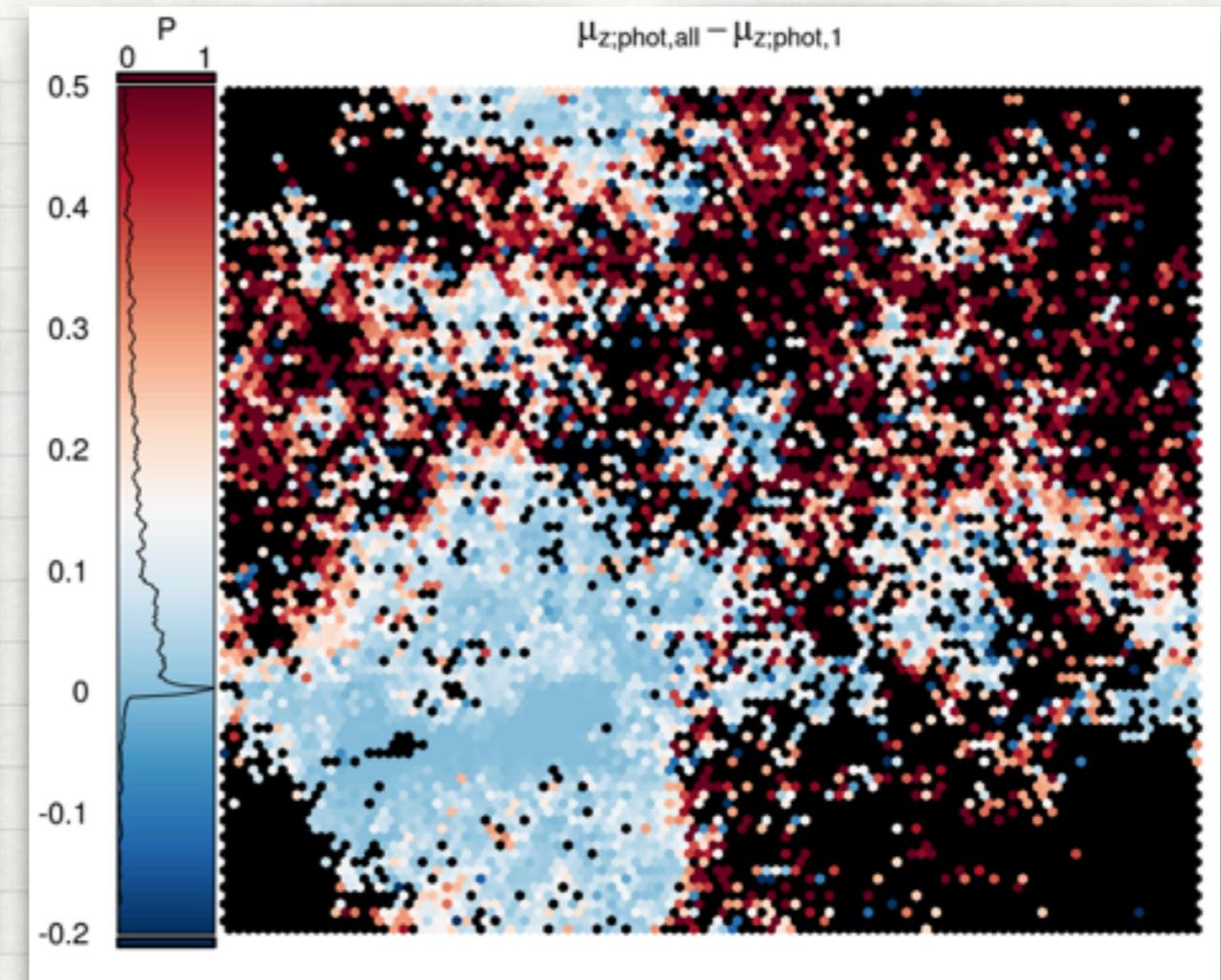
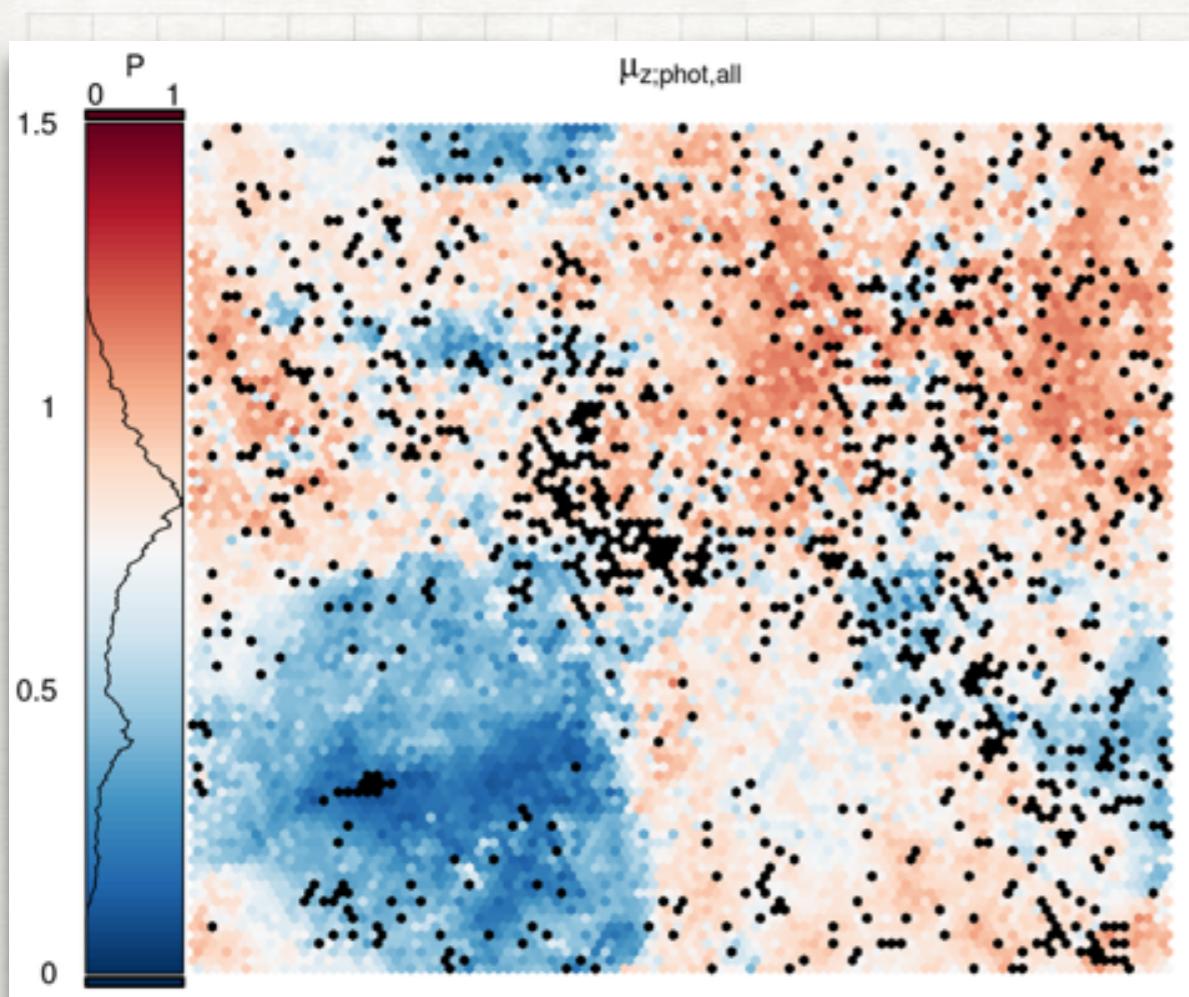
The goal is redshift calibration. So we can select the cluster number that produces unchanging mean-z results.  
↳ mean-z is quite stable, neff is not

# REDSHIFT CALIBRATION REQUIREMENTS

The reweighting method requires that the associations have  $\sim$ unique redshifts



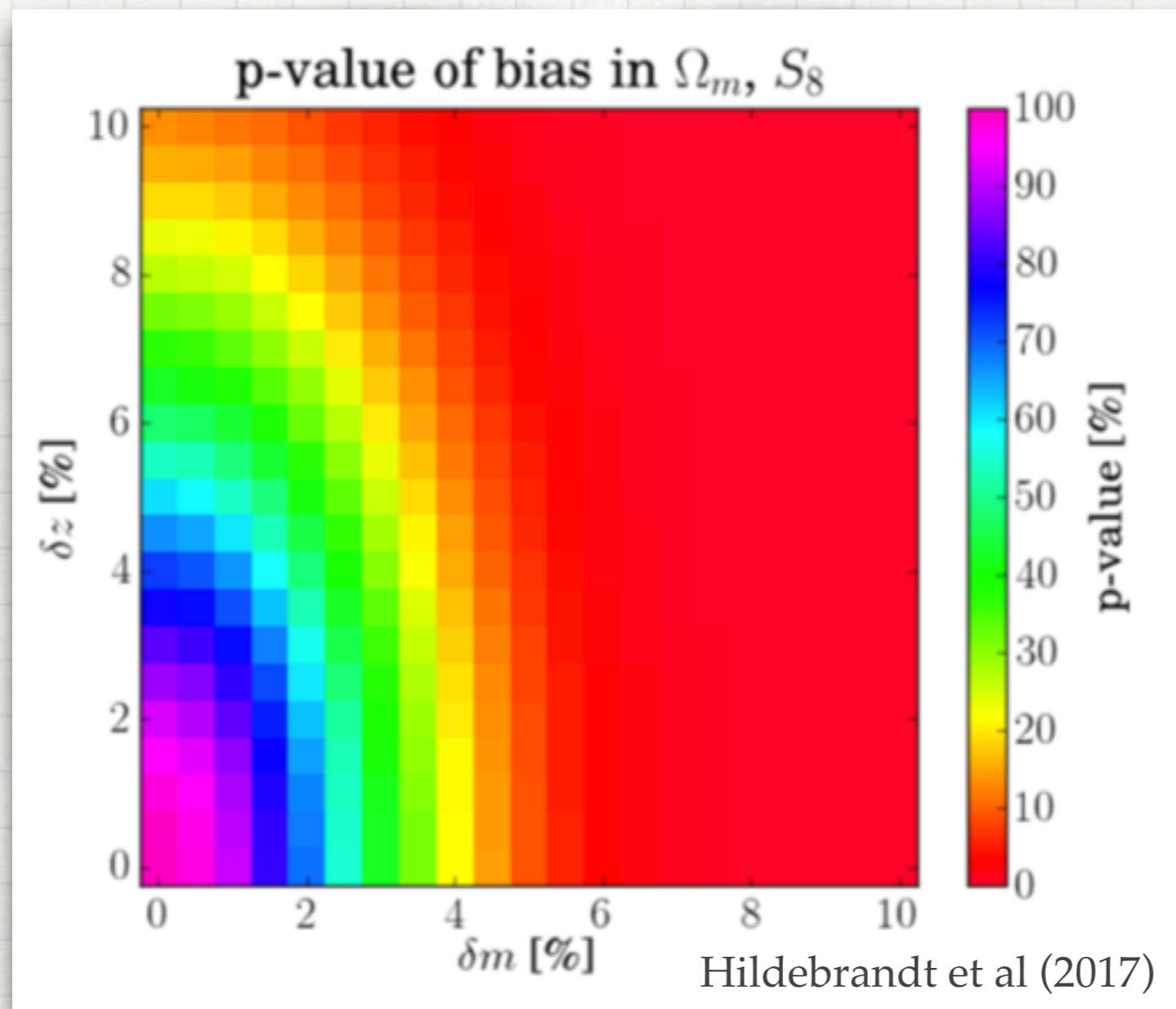
# SPECTROSCOPIC VALIDATION FOR COSMIC SHEAR TOMOGRAPHIC BINNING



With tomographic binning, we apply a strong redshift-dependant selection effect. This modifies the  $N(z)$  of the SOM pixels, and necessitates like-for-like binning of the spectra.

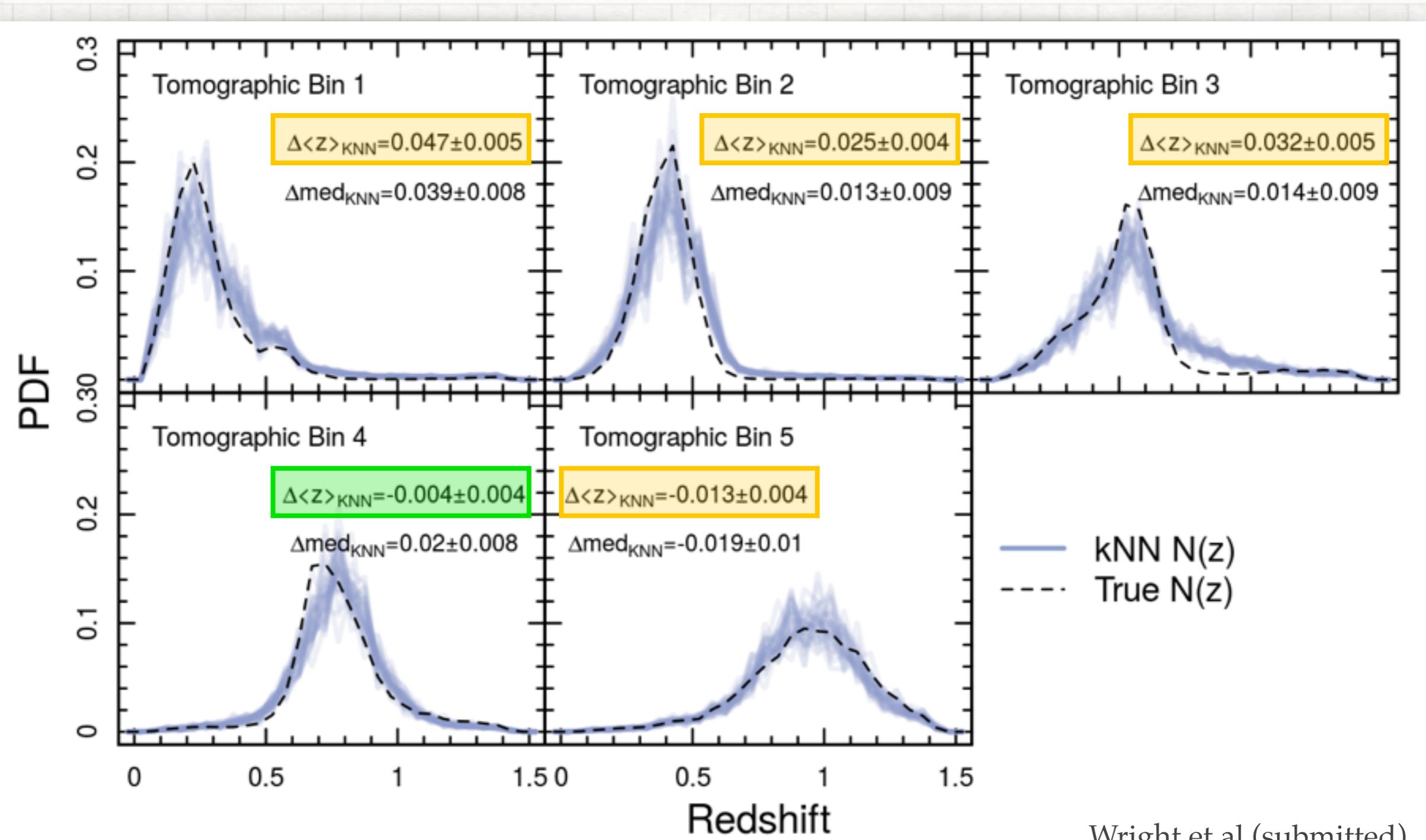
Not previously implemented in kNN methods  
Wright et al (submitted)

# TESTING THE DIRECT CALIBRATION



# TESTING THE DIRECT CALIBRATION

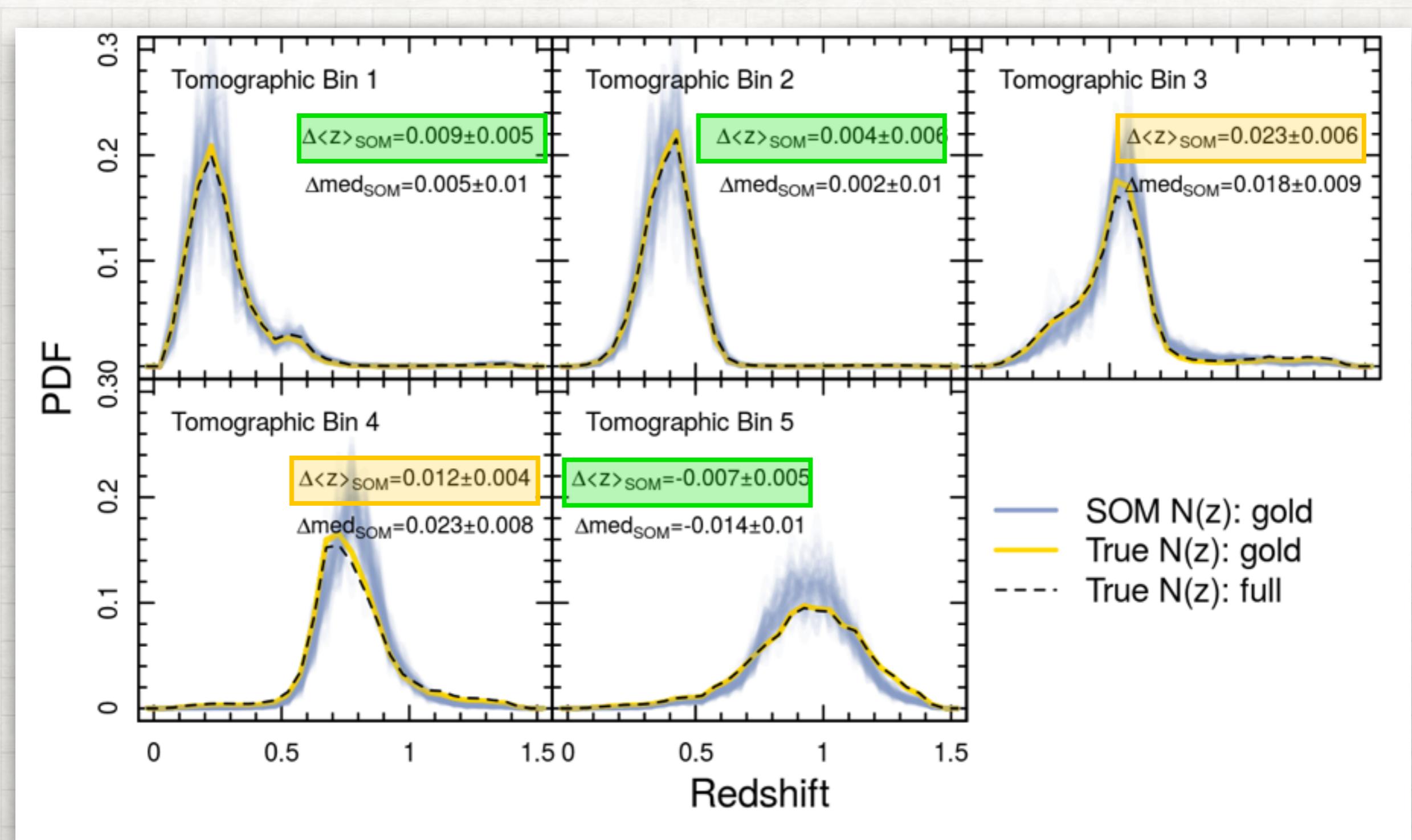
100 spectroscopic lines-of-sight & KiDS-like photometry using MICE2



Wright et al (submitted)

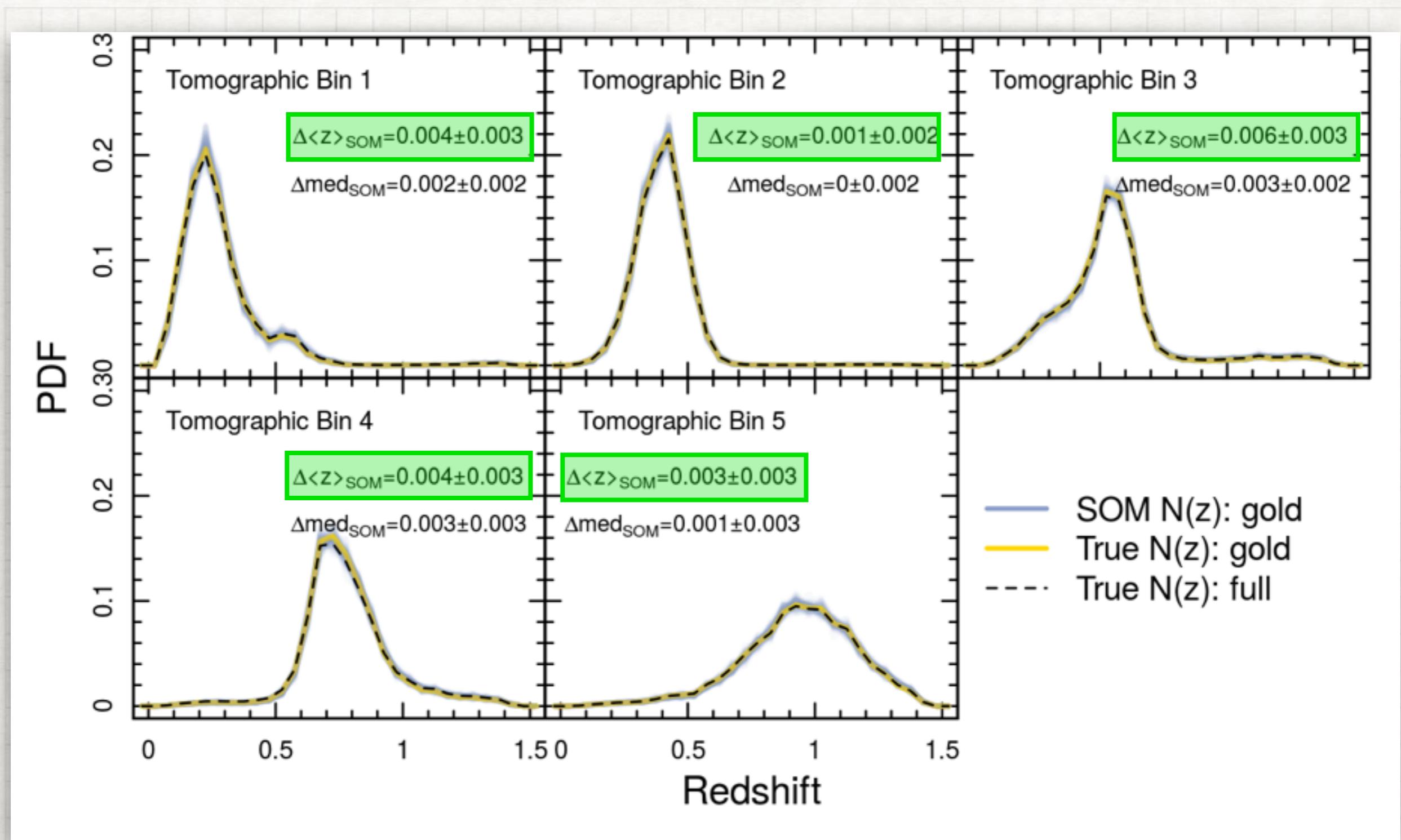
# SOM PHOTOMETRIC REDSHIFT CALIBRATION

## KIDS-LIKE DATA



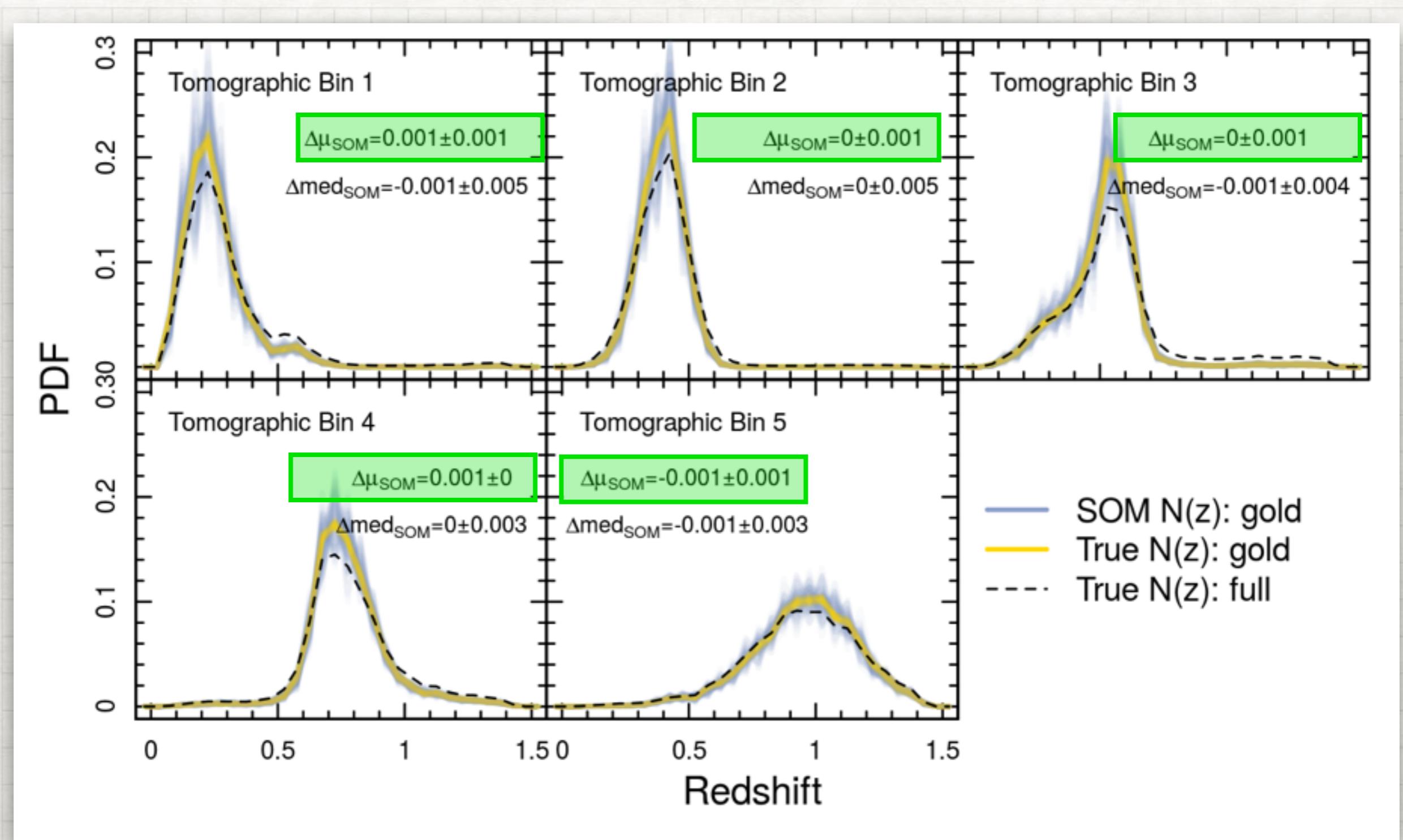
# SOM PHOTOMETRIC REDSHIFT CALIBRATION

## PERFECT SPECTROSCOPY, NOISY PHOTOMETRY



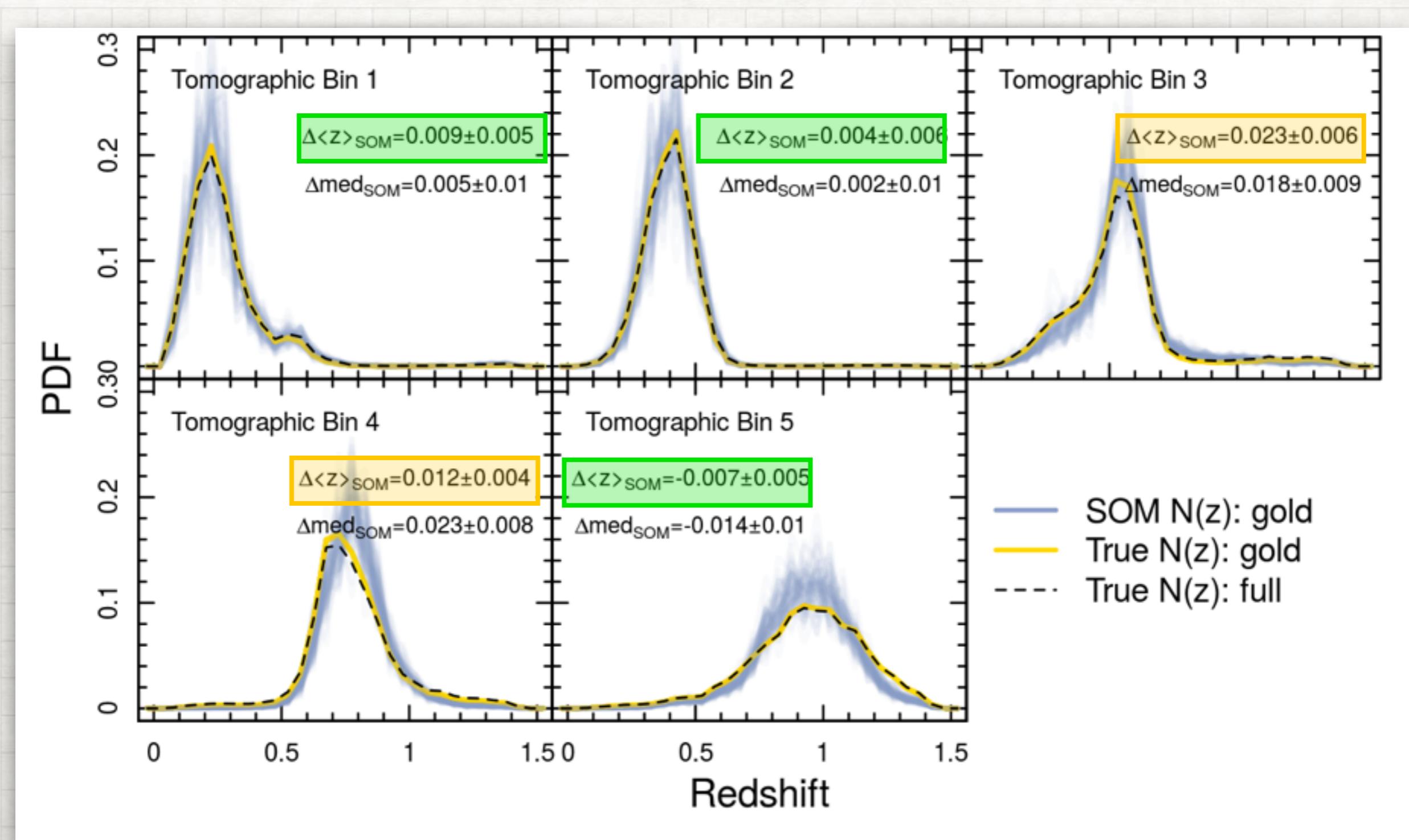
# SOM PHOTOMETRIC REDSHIFT CALIBRATION

## BIASED SPECTROSCOPY, NOISELESS PHOTOMETRY



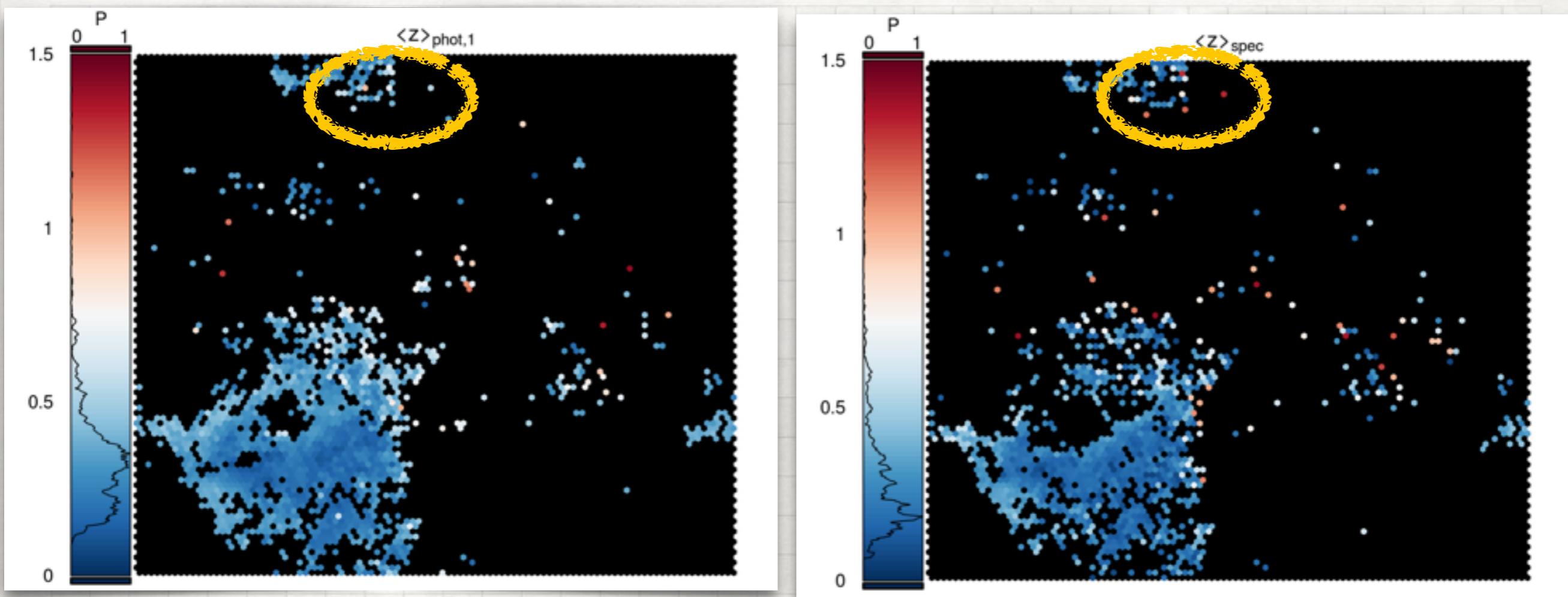
# SOM PHOTOMETRIC REDSHIFT CALIBRATION

## KIDS-LIKE DATA



# IMPROVING THE CALIBRATION FURTHER

## QUALITY CONTROL



The bias in the noisy+spectroscopically selected calibration comes from a small number of biased associations. These can be flagged and removed.

# FINAL CALIBRATION STATISTICS

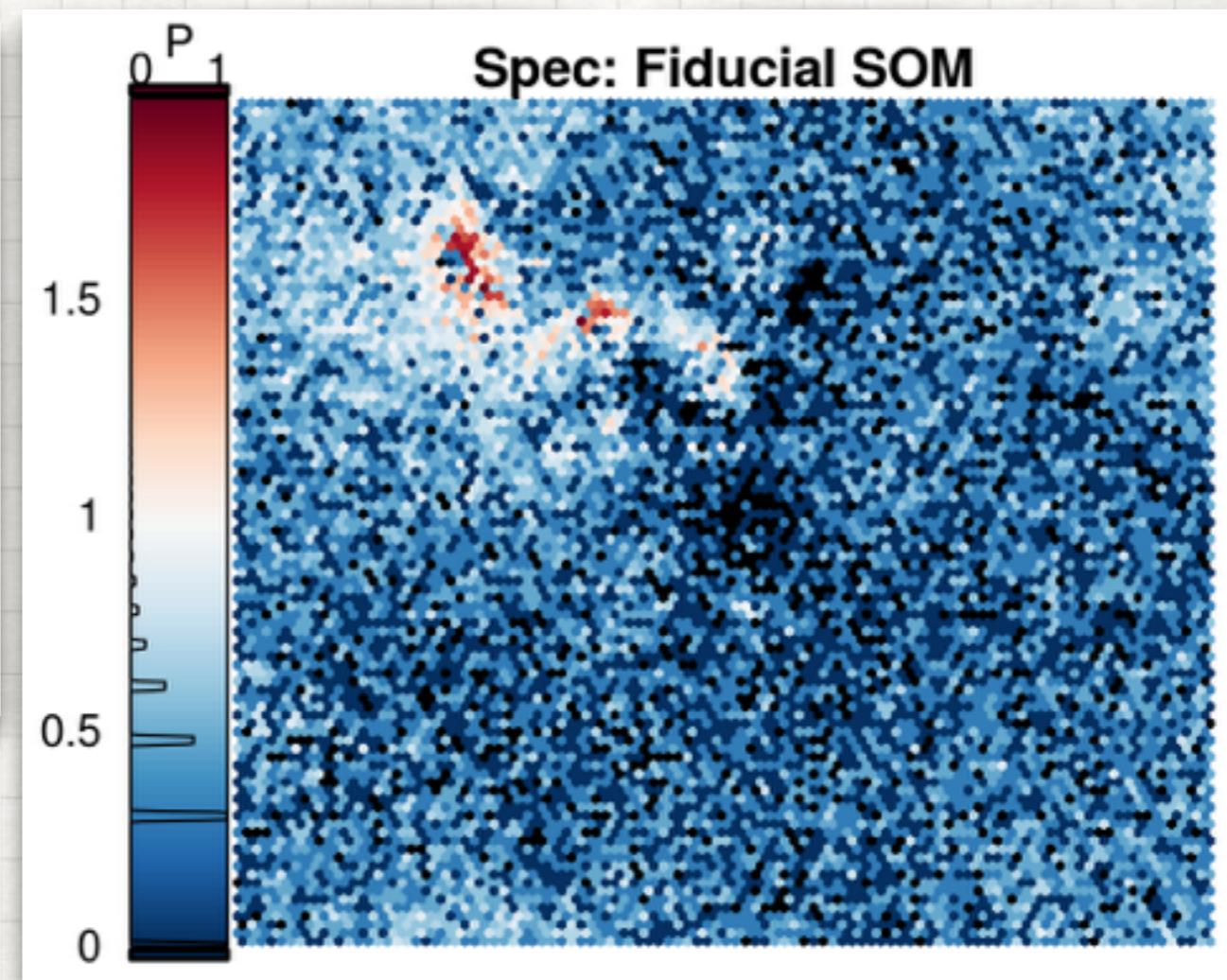
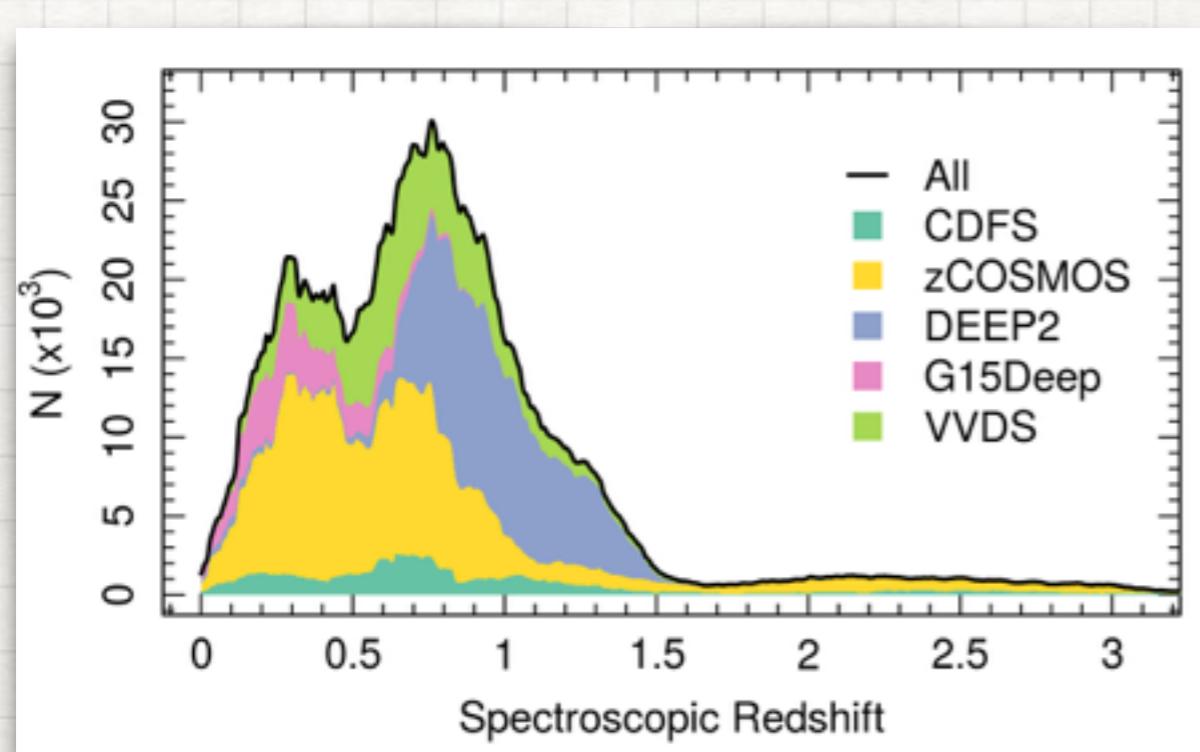
Dataset		Reconstruction Bias ( $\Delta\langle z \rangle$ )				
Type	Phot	bin1	bin2	bin3	bin4	bin5
perfect	exact	$0.001 \pm < 10^{-3}$	$\emptyset$	$\emptyset$	$\emptyset$	$\emptyset$
KV450	exact	$0.002 \pm 0.001$	$0.002 \pm 0.002$	$0.003 \pm 0.001$	$0.003 \pm 0.001$	$-0.001 \pm 0.001$
Perfect	noisy	$0.004 \pm 0.003$	$< 10^{-3} \pm 0.002$	$0.006 \pm 0.003$	$0.004 \pm 0.003$	$0.003 \pm 0.003$
KV450	noisy	$0.009 \pm 0.005$	$0.004 \pm 0.006$	<b><math>0.023 \pm 0.006</math></b>	<b><math>0.012 \pm 0.004</math></b>	$-0.007 \pm 0.005$
KV450	noisy+QC1	$< 10^{-3} \pm 0.005$	$0.002 \pm 0.006$	<b><math>0.013 \pm 0.006</math></b>	<b><math>0.011 \pm 0.004</math></b>	$-0.006 \pm 0.005$
KV450	noisy+QC2	$0.002 \pm 0.005$	$0.003 \pm 0.006$	$0.007 \pm 0.005$	$0.009 \pm 0.004$	$-0.006 \pm 0.004$
KV450	noisy	<i>using kNN association</i>				
		<b><math>0.047 \pm 0.005</math></b>	<b><math>0.025 \pm 0.004</math></b>	<b><math>0.032 \pm 0.005</math></b>	$-0.004 \pm 0.004$	<b><math>-0.013 \pm 0.004</math></b>

The updated calibration produces more accurate calibration, and is able to reach  $\sim 0.01 \pm 0.01$  precision in all bins when performing some quality control.

The interpretation is that this reduction in bias comes from the removal of non-represented photometric data

# SPECTROSCOPIC VALIDATION FOR COSMIC SHEAR

## TESTING SPECTRAL DATASETS

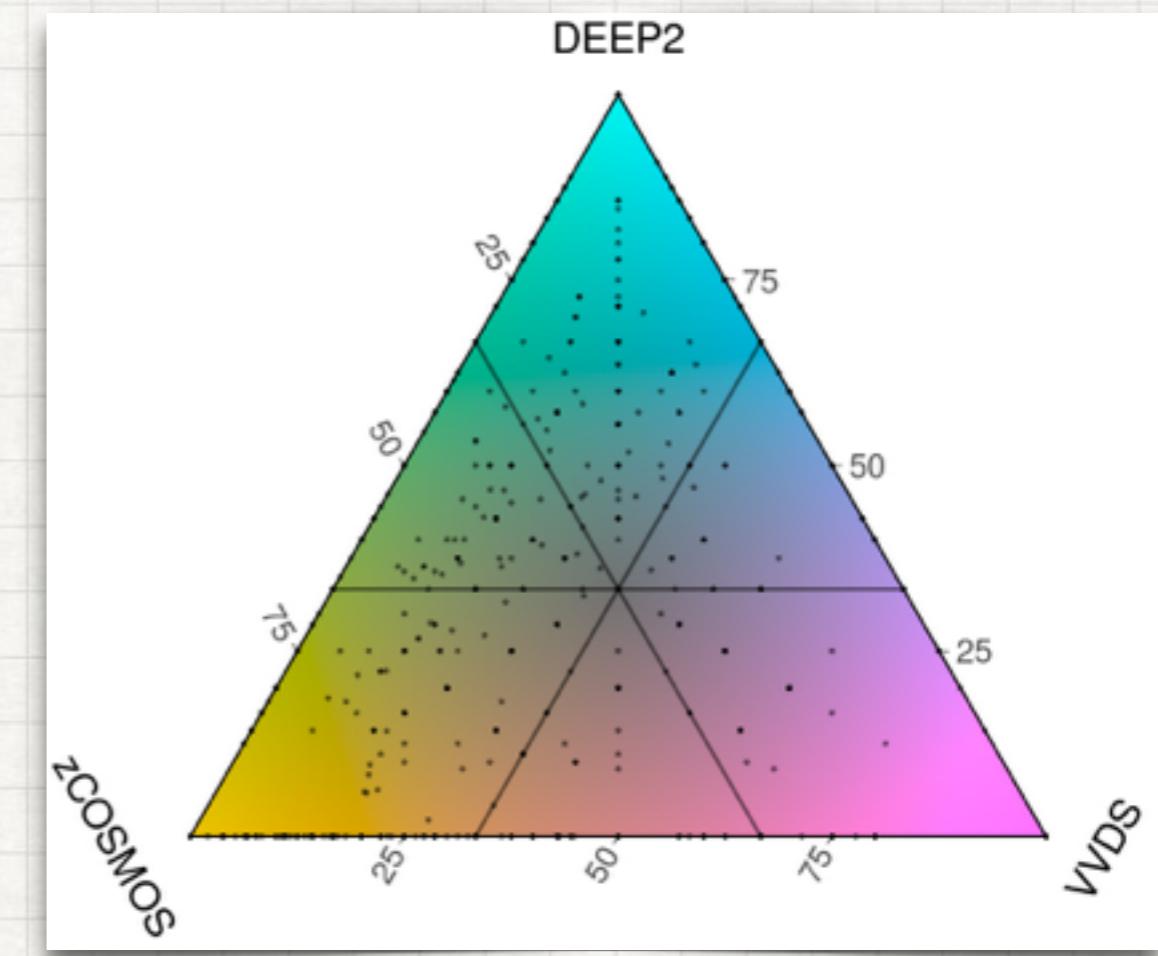
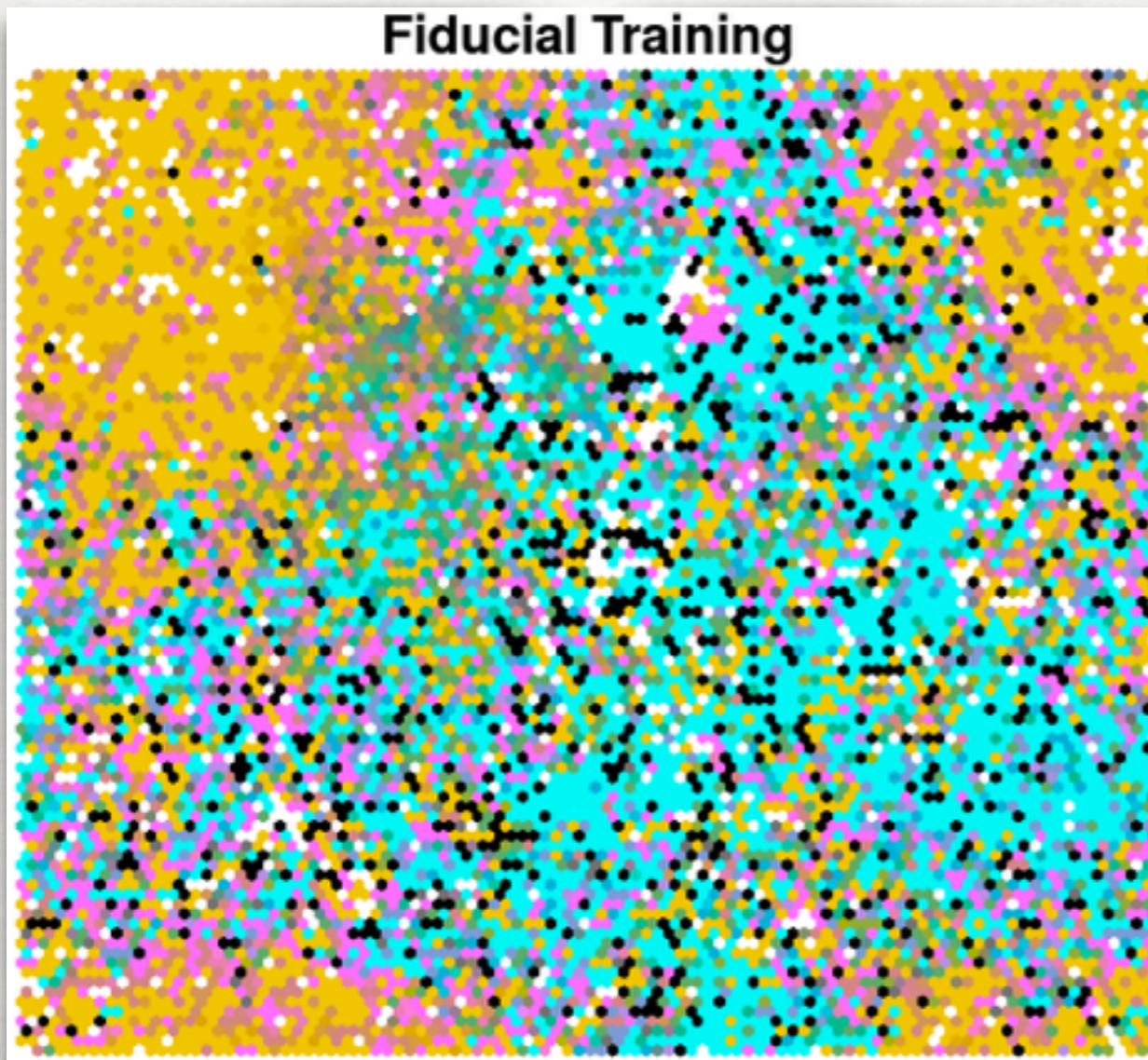


The cosmic shear spectroscopic compilation is an amalgam of many different spectroscopic datasets

This dataset has a complicated colour mapping, and various selection biases. How representative is our compilation? Does any one survey dominate our calibration?

# SPECTROSCOPIC VALIDATION FOR COSMIC SHEAR

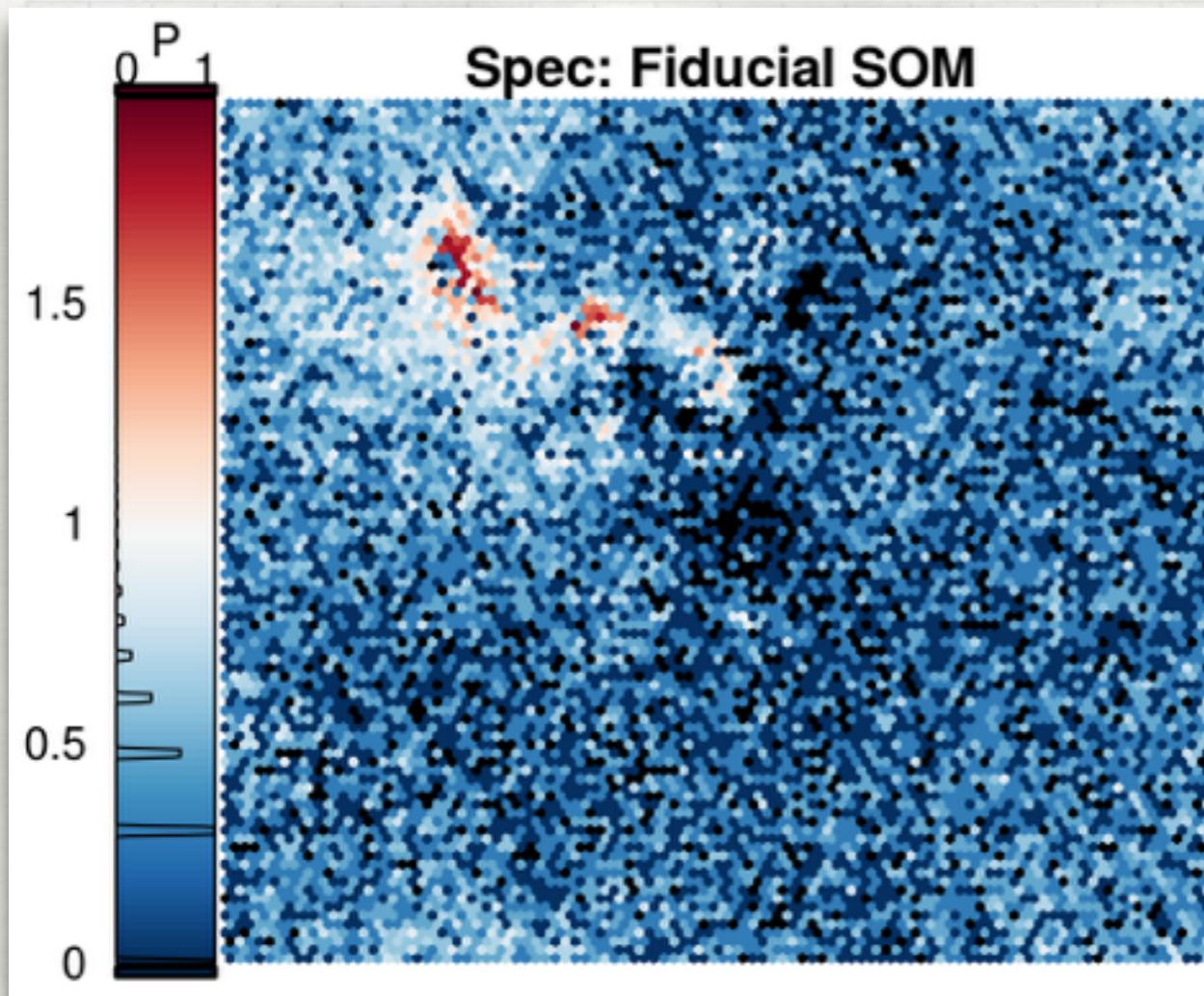
## TESTING SPECTRAL DATASETS



The spectroscopic data are highly complementary:  
rarely do two or more datasets occupy any one SOM pixel

# SPECTROSCOPIC VALIDATION FOR COSMIC SHEAR

## TESTING SPECTRAL DATASETS



9% of our spectroscopic maps pixels are devoid of spectra.

However these pixels host only 0.5% of the 2D cosmic shear photometric data.

- ↳ Pixels without spectroscopy are overwhelmingly in insignificant parts of the colour-space for KiDS.

# SPECTROSCOPIC VALIDATION FOR COSMIC SHEAR TOMOGRAPHIC BINNING

Spectroscopic Compilation	Training Size (all)	$f_{\text{pix}}$ (all, %)	$z_B \in [0.1, 1.2]$	$n'_{\text{eff}}/n_{\text{eff}}(\%)$				
				All	bin1	bin2	bin3	bin4
Full Sample	25373	91.9	99.5	82.7	83.9	84.6	82.7	94.0
CDFS only	2044	17.4	67.3	57.1	58.7	53.2	40.2	54.6
zCOSMOS only	9930	<b>48.5</b>	79.7	<b>74.9</b>	<b>75.3</b>	65.8	60.3	63.2
DEEP2 only	6919	43.7	73.8	17.8	5.5	35.2	<b>68.8</b>	<b>89.5</b>
G15DEEP only	1792	10.1	42.0	63.1	63.6	44.1	19.8	14.0
VVDS only	4688	34.7	<b>81.4</b>	54.9	72.8	<b>70.7</b>	57.3	70.2
without CDFS	23329	89.1	98.9	81.5	82.6	82.2	81.0	93.0
without zCOSMOS	15443	77.6	97.8	<b>76.4</b>	80.0	80.8	78.9	92.6
without DEEP2	18454	<b>73.4</b>	<b>93.4</b>	81.6	83.0	80.4	<b>72.2</b>	<b>80.8</b>
without G15DEEP	23581	90.6	99.5	80.1	83.6	84.3	82.7	94.0
without VVDS	20685	84.7	98.1	81.3	<b>79.5</b>	<b>77.5</b>	80.1	92.9

The 2D-cosmic shear sample is very well represented by the compilation. Tomographic binning reduces the representation as noise begins to play a more significant role.

# SPECTROSCOPIC VALIDATION FOR COSMIC SHEAR

## TOMOGRAPHIC BINNING

Samp/ Photom	All Bins	Bin 1	Bin 2	Bin 3	Bin 4	Bin 5
All Spectra	99.5 %	83 %	84 %	85 %	83 %	94 %
Without DEEP2	93 %	82 %	83 %	80 %	72 %	81 %
Without zCOSMOS	98 %	76 %	80 %	81 %	79 %	93 %

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All Spectra	99.5 %	83 %	84 %	85 %	83 %	94 %
Without DEEP2	93 %	82 %	83 %	80 %	72 %	81 %
Without zCOSMOS	98 %	76 %	80 %	81 %	79 %	93 %

Our highest-z tomographic bin is our *best* represented.

DEEP2 informs best the highest redshift bins. zCOSMOS the lowest.

# SPECTROSCOPIC VALIDATION FOR COSMIC SHEAR

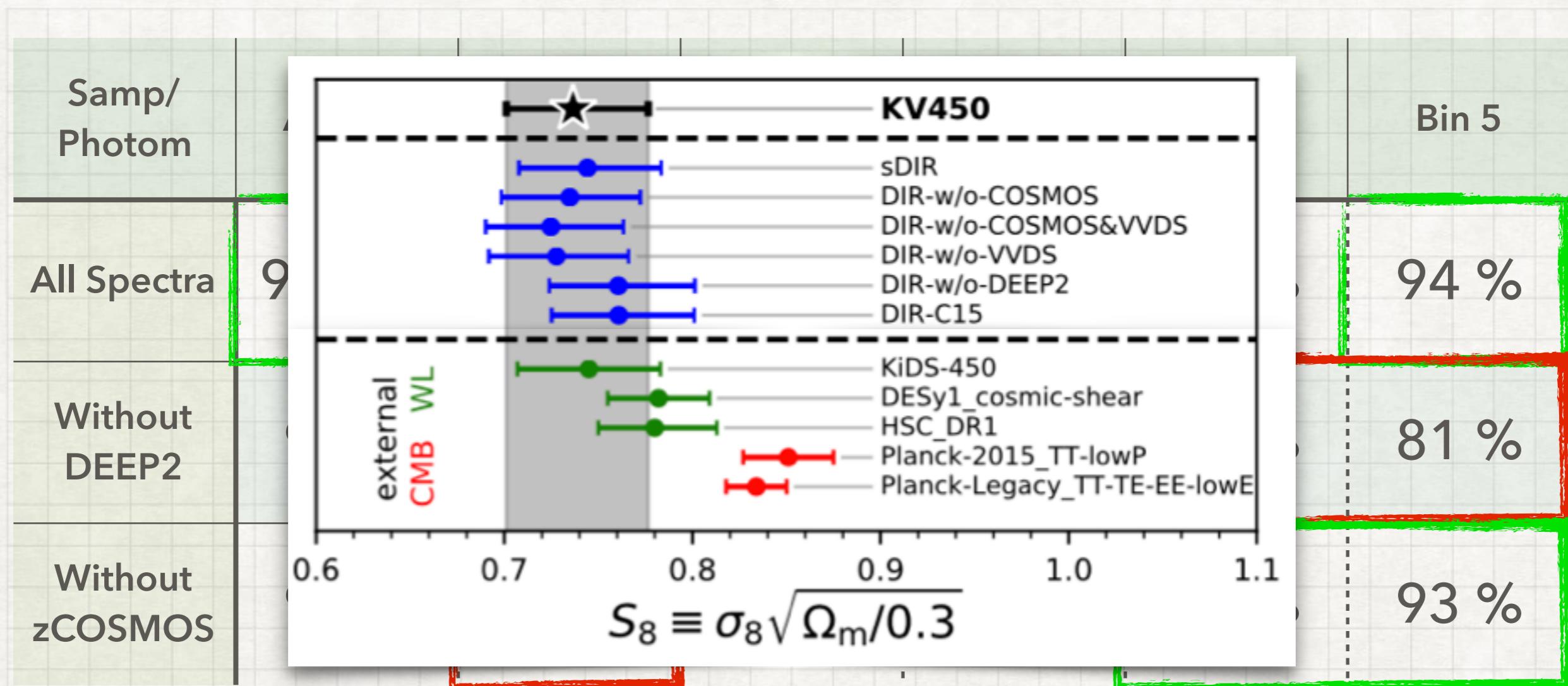
## TOMOGRAPHIC BINNING

Samp/ Photom	All Bins	Bin 1	Bin 2	Bin 3	Bin 4	Bin 5
All Spectra	99.5 %	83 %	84 %	85 %	83 %	94 %
Without DEEP2	93 %	82 %	83 %	80 %	72 %	81 %
Without zCOSMOS	98 %	76 %	80 %	81 %	79 %	93 %

Our highest-z tomographic bin is our *best* represented.

DEEP2 informs best the highest redshift bins. zCOSMOS the lowest.

# SPECTROSCOPIC VALIDATION FOR COSMIC SHEAR TOMOGRAPHIC BINNING

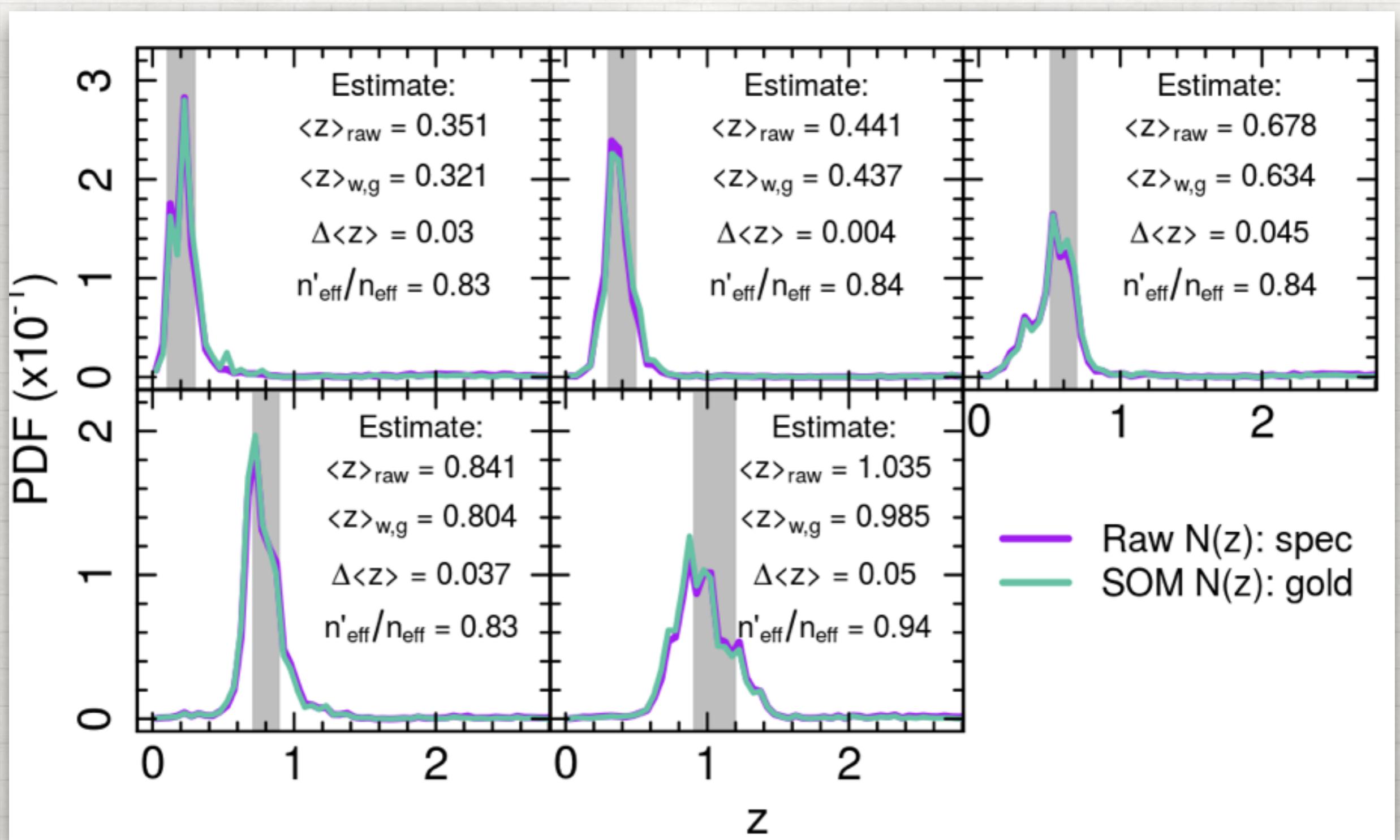


Our highest-z tomographic bin is our *best* represented.

DEEP2 informs best the highest redshift bins. zCOSMOS the lowest.

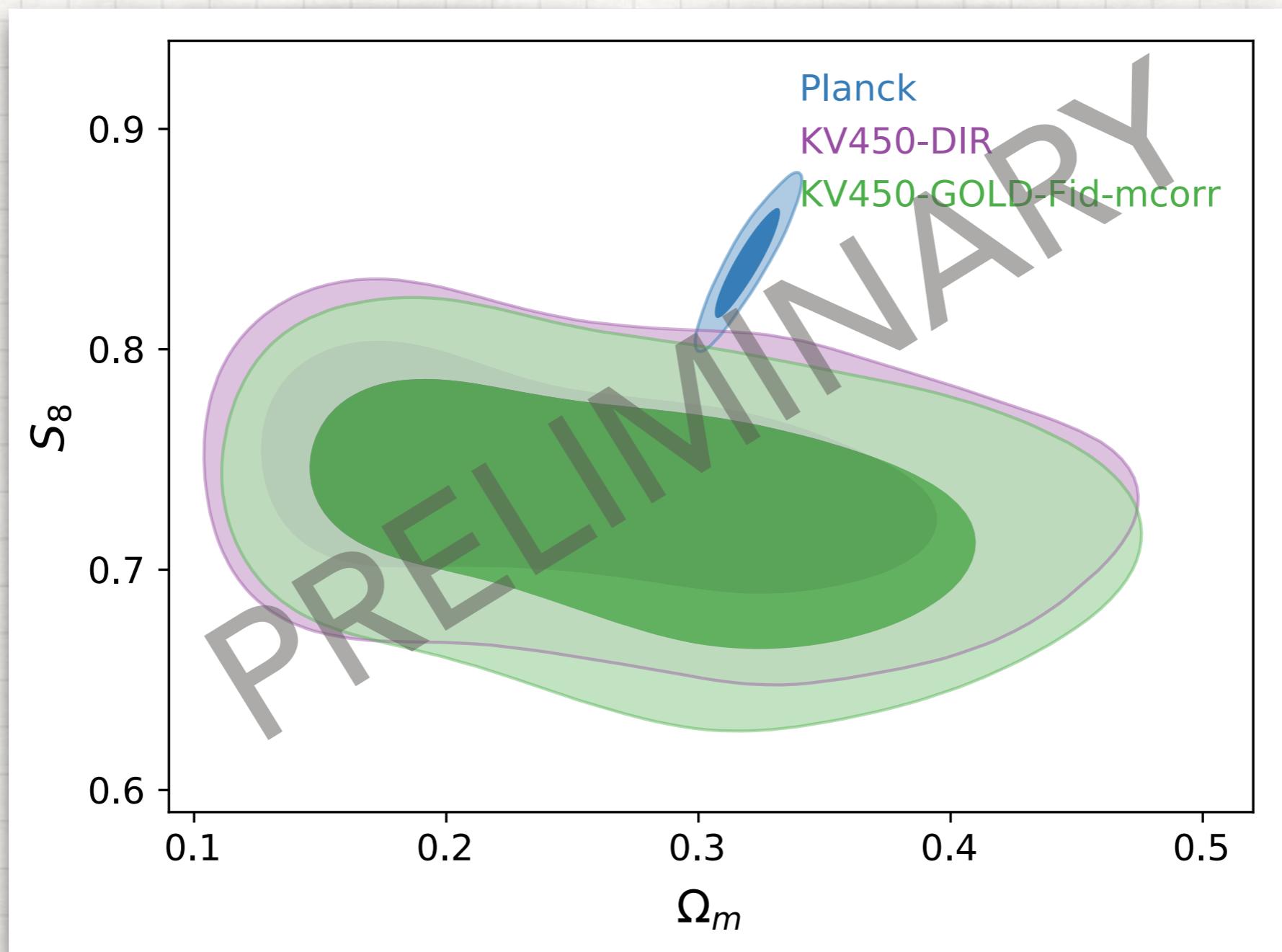
# NEW PHOTOMETRIC REDSHIFT CALIBRATION

## KV450 DATA



# NEW COSMOLOGICAL PARAMETER ESTIMATES

## KV450 GOLD SAMPLE



# SUMMARY

- At KiDS-level noise, the MICE2 redshift distributions are unbiased in all bins except for bin 4, but this is limited to MICE2.
- In the noiseless case, the SOM redshift calibration is unbiased.
- The bin 4 bias is caused by a colour-redshift degeneracy present in the DEEP2 colour-space.  
Future analyses need to be robust to this.
- The dominant factor in the accuracy of the redshift calibration is noise in the spectroscopic compilation.
- The SOM calibration method allows construction of a 100% represented 'gold' sample, which we will use for a reanalysis of KV450 cosmology.
- The KiDS spectroscopic compilation represents >99% of the 2D photometric lens sample.