

# Euclid STAR Prize 2020 Timeline





# Photometric Redshifts for Euclid

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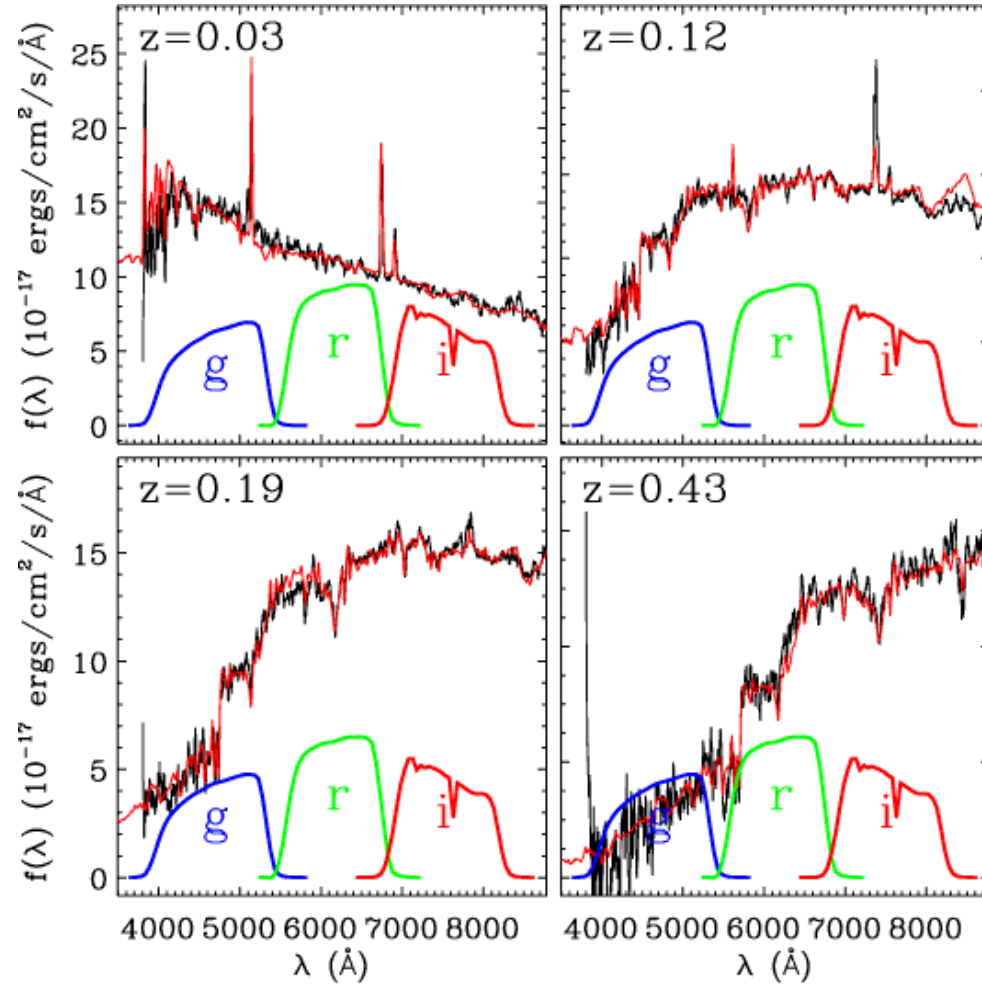


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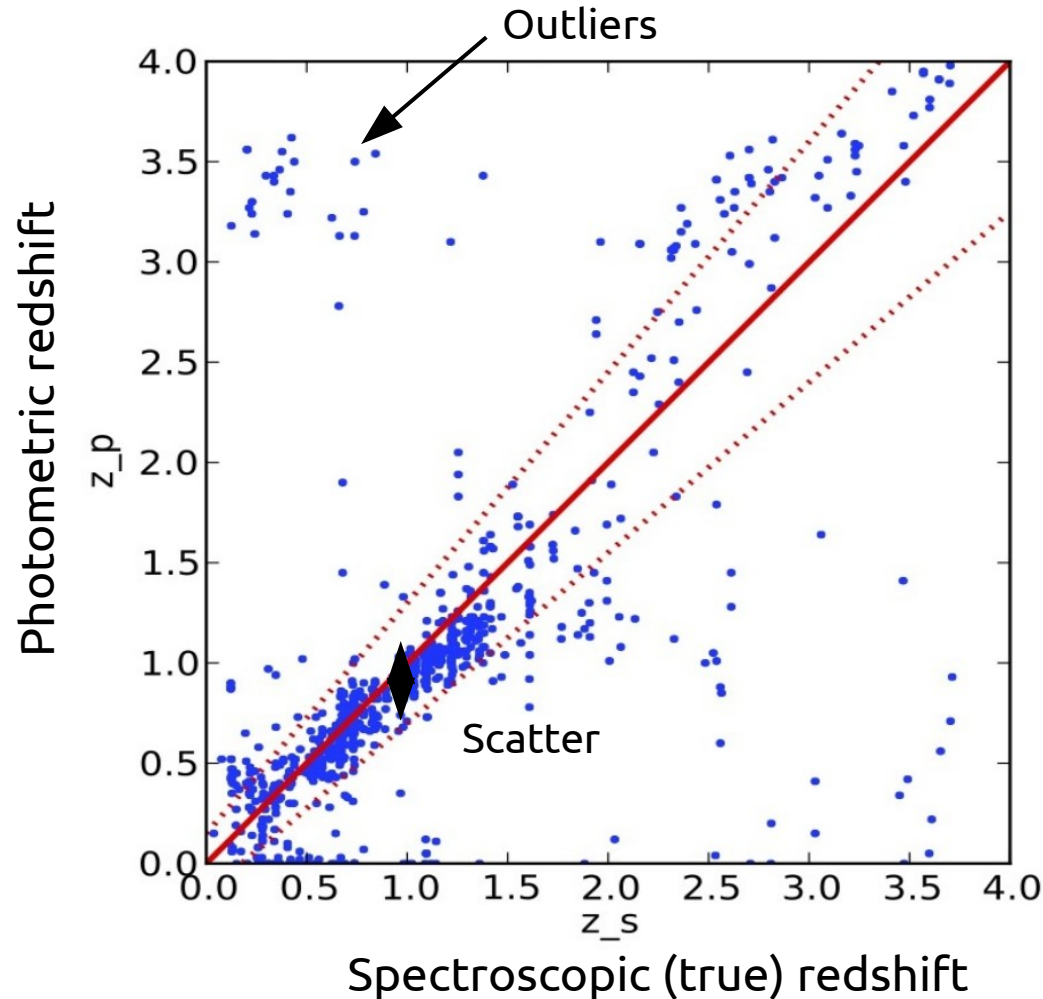
**FACULTÉ DES SCIENCES**  
Département d'astronomie

- OU-PHZ is a very important Swiss responsibility in Euclid linked to Euclid's **main science**
  - About 50 contributors
- Main task is to **determine redshifts for weak-lensing tomography**
  - Redshifts need to be determined from photometry alone (no spectroscopy)
- OU-PHZ needs also to
  - Identify stars (and AGN) for PSF determination (and contamination)
  - Provide unbiased colors from 500 to 1000 nm for shape reconstruction
  - Compute galaxy physical parameters (Mass, SFR, SFH, reddening,...)

# Photometric Redshifts



# Photo-z vs True Redshifts



Euclid requirements :

Precision (scatter):

$$\sigma_z < 0.05(1+z)$$

Outlier fraction : < 10 %  
beyond  $0.15(1+z)$

Accuracy (bias):

knowledge of  $\bar{z}$  in any  
tomographic bin better than  
 $0.002(1+z)$

Produce PDF of photometric  
redshifts

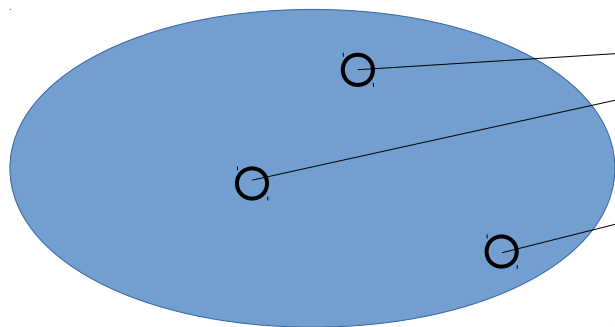
# Challenges of Euclid Photo-z's



- **Inhomogeneous photometry** (different surveys, large sky area)
- Some regions of the sky with **high Galactic reddening**
- Very demanding requirement on **photo-z accuracy** (bias)
  - But moderate requirements on photo-z precision
- Huge number of objects makes the problem computationally hard
- Need for a **fully statistical framework** to propagate all uncertainties correctly into the cosmological parameters

# Photometric-redshift Algorithms

Colour/Flux(+...) space



Redshift space



$$z=f(\text{colours}, \text{fluxes})$$

Mapping  $f$  can be constructed based on **prior astrophysical knowledge** :

- Template-fitting: Hyper-Z, Le Phare, BpZ, **Phosphoros**,...

Or it can be discovered **using known (spectroscopic) redshifts**:

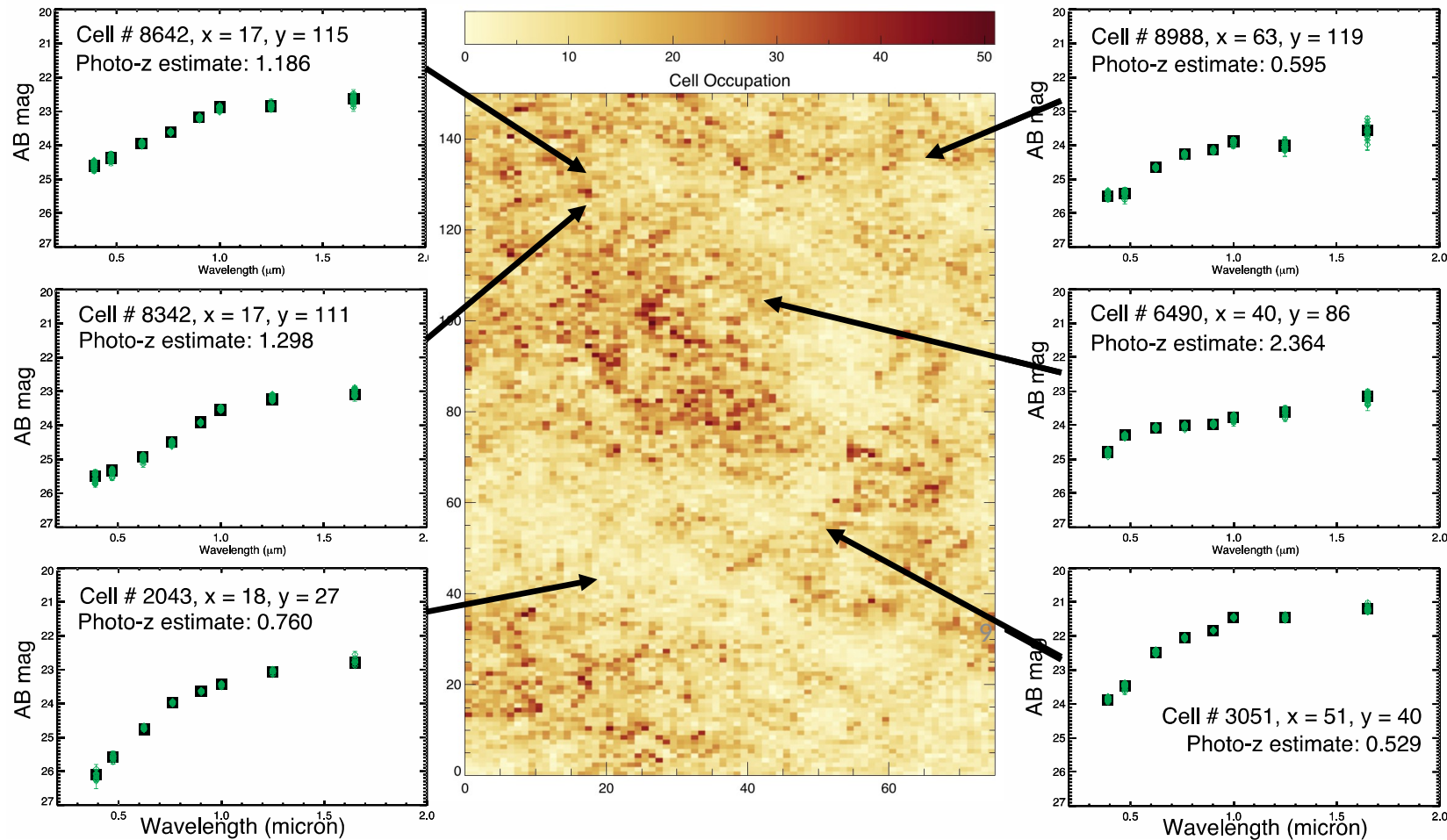
- Machine-learning: Nearest neighbours, Perceptron, Support vector regression, Random Forest, Adaboost, Gaussian Processes, ...
- **See Guillaume's presentation**

# Template-fitting vs Machine-learning

	Template-fitting	Machine-learning
Probabilistic	✓	✗
Does not require any understanding of physics	✗(✓)	✓(✗)
Does not require complete training sample	✓	✗
Implement priors naturally	✓	✗
Can be free of hidden priors	✓	✗
Can use heteroscedatic data (but useful?)	✗	✓
Little sensitive to extrapolations	✓	✗
Computationally efficient	✗	✓
Easy to prevent over/underfitting	✗	✓
Optimal for the test sample	✗	😊

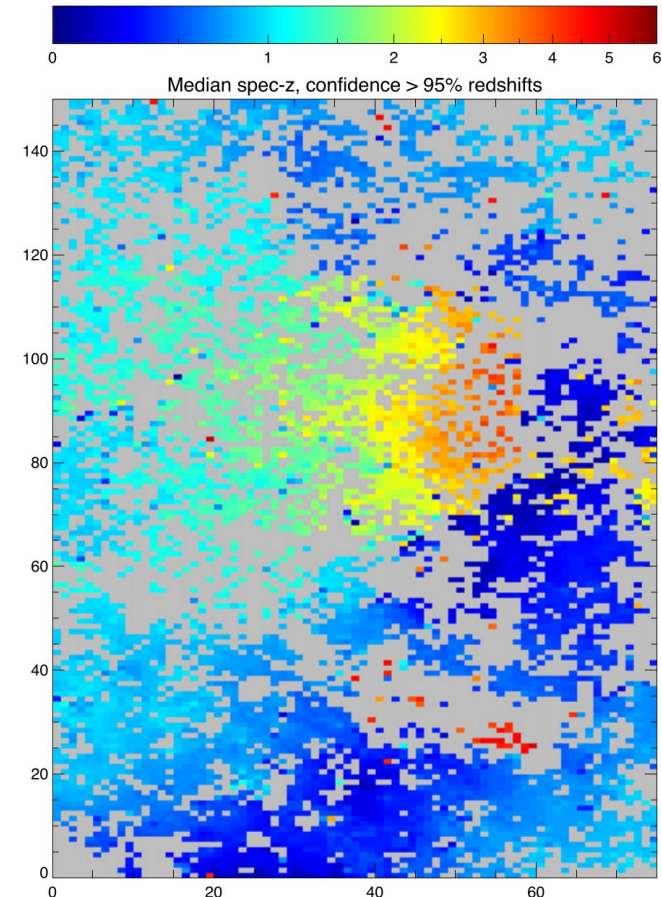
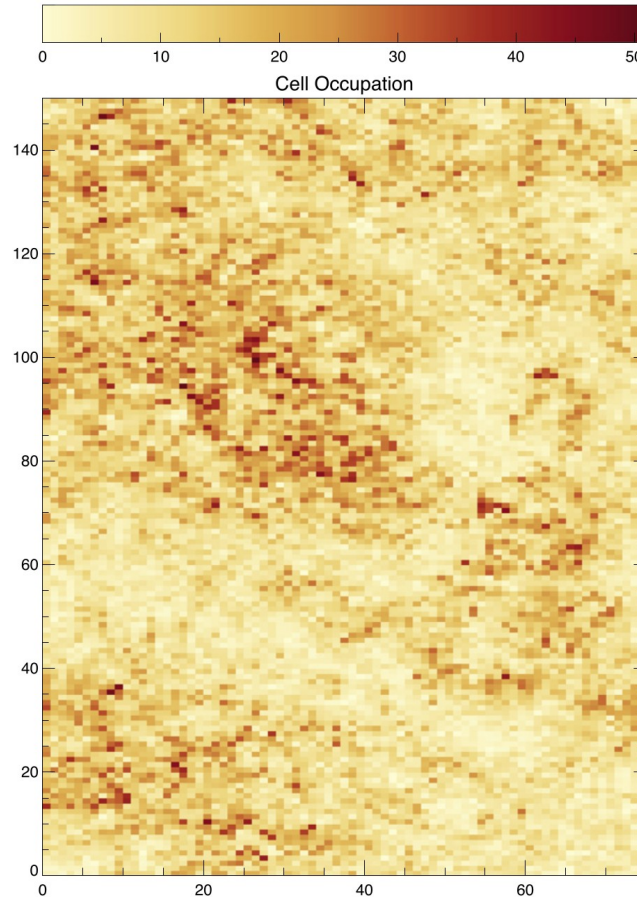


# Self-Organizing Map of Galaxy Colour Space



# Completeness of the Training Sample

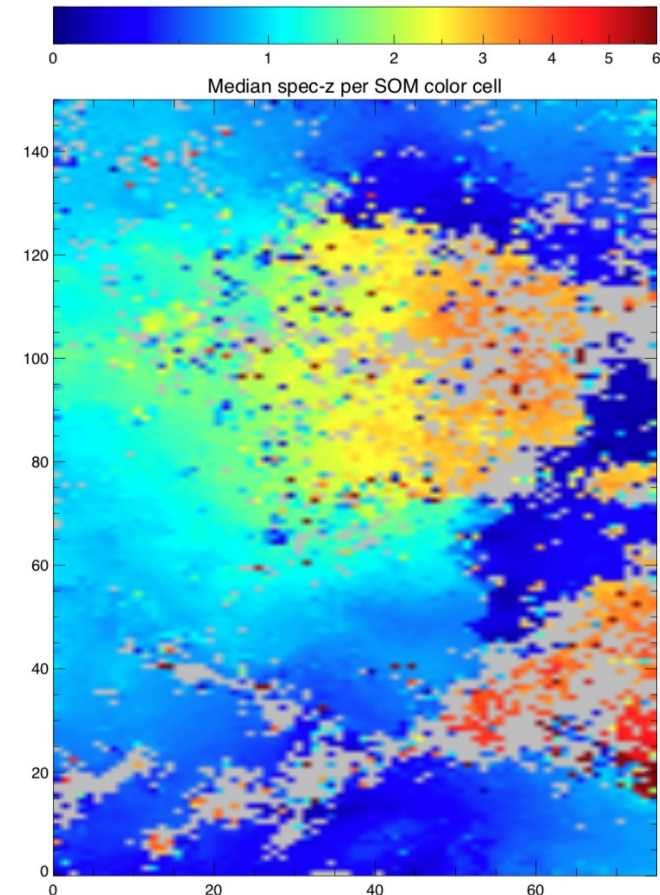
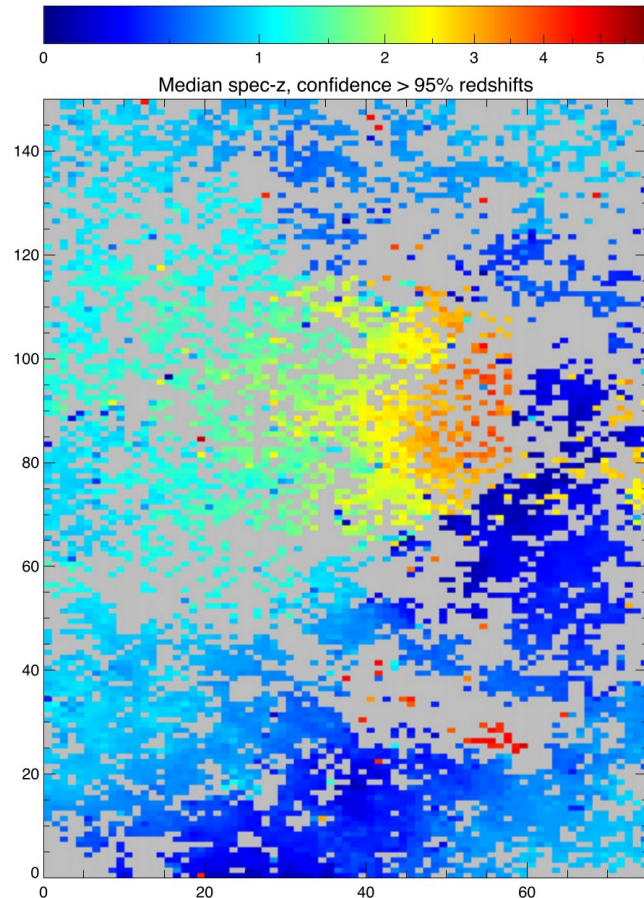
- A self-organizing map can be used to find out where are the missing objects in color space
- Large areas of the color space lack any spec-z coverage, some parts in high-density areas



# C3R2 Program



- Complete Calibration of the Color-Redshift Relation
- 50 nights on Keck, 200h on VLT
- Goal is to determine  $\langle z \rangle$  in each cell

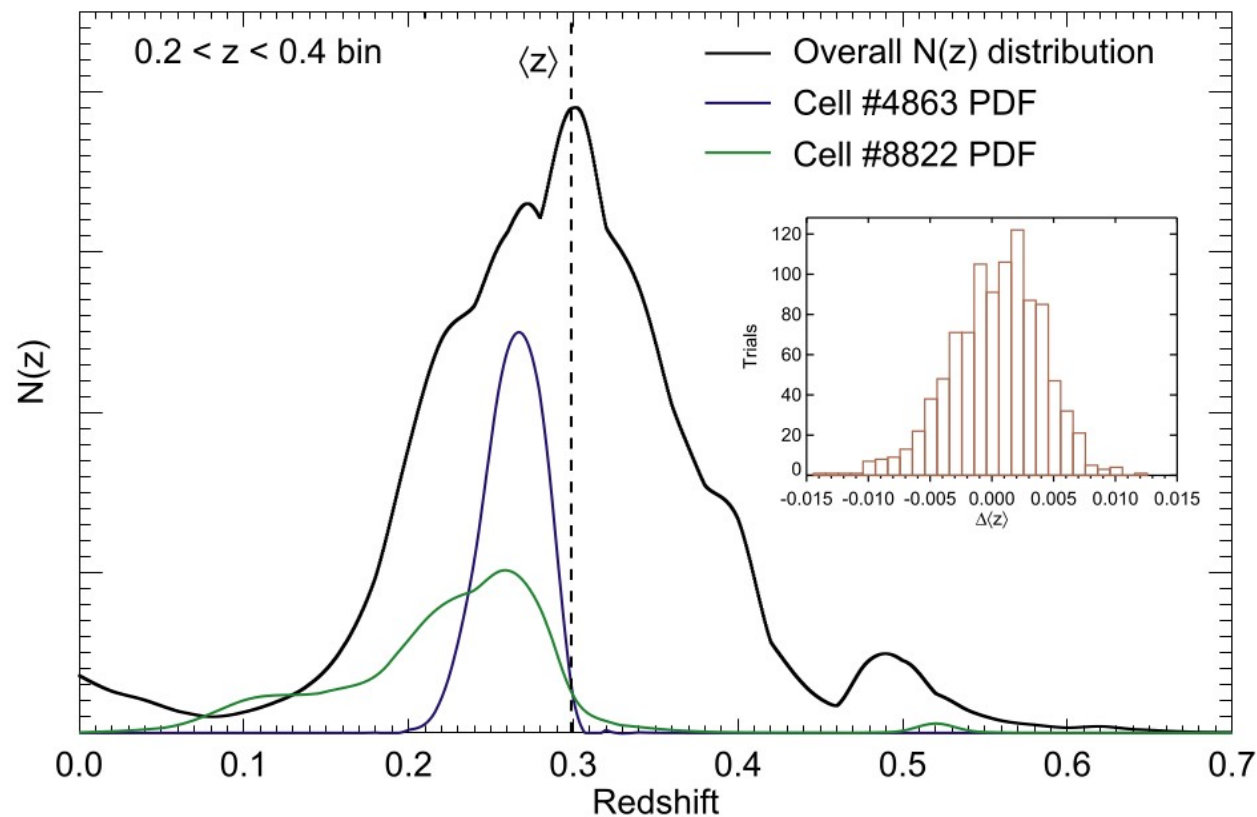




# Photo-z Calibration using the SOM

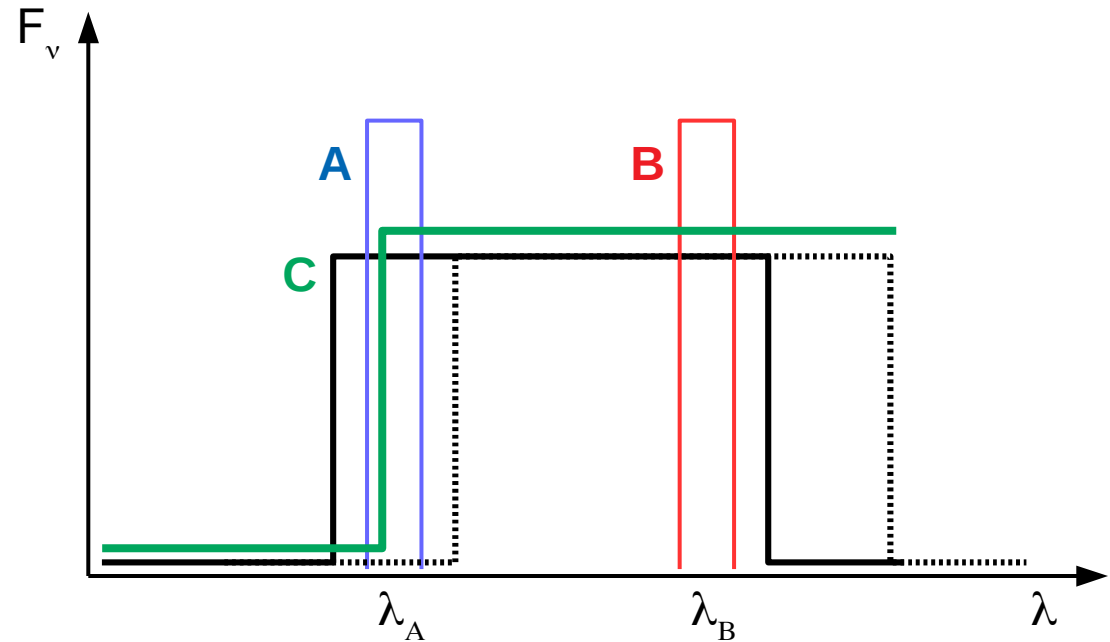
$$\langle z \rangle = \frac{1}{\sum_i n_i} \sum_i n_i \langle z_i \rangle$$

$N(z)$



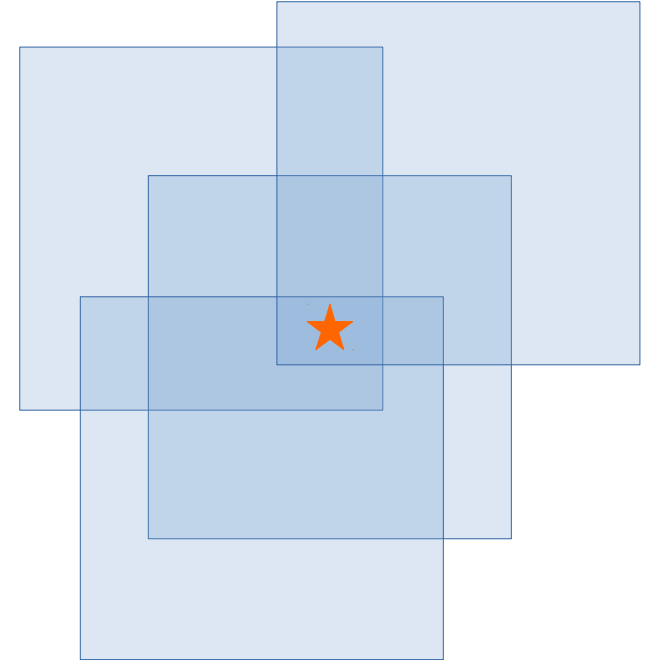
# Inhomogeneous Photometric Survey

- Difference in transmissions lead to inhomogeneous colors
- This can immediately introduce a bias in the photo-z if the effect is not taken into account
- Usually addressed using a colour-term correction
  - But depends on SED

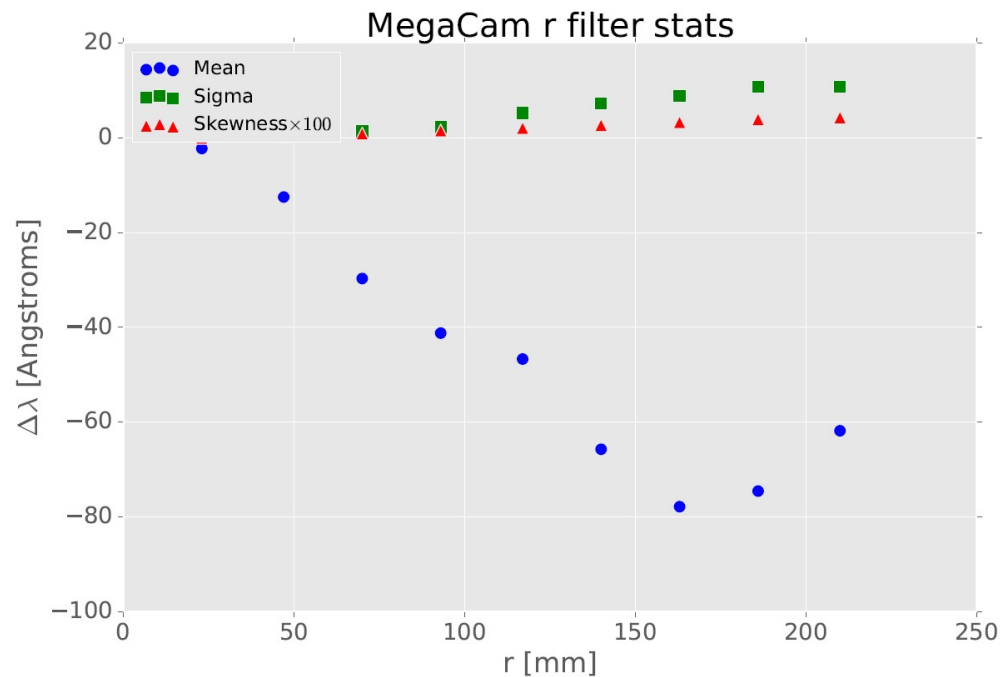
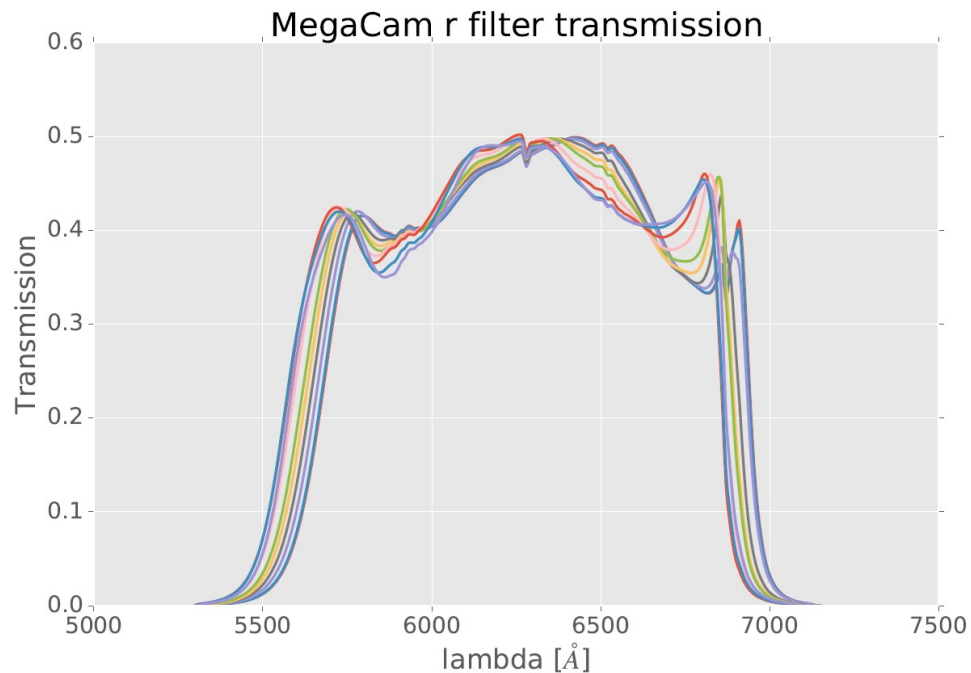


# Variable Transmissions

- Variations in transmissions **in a single survey** can occur for many reasons:
  - **Inhomogeneous filters** and CCD QEs, dependent on the location in the FOV
  - Filter or CCD aging, dependent on time
  - **Galactic reddening**, dependent on sky position
- Some effects are mitigated by having several pointings, but not all, and not clear to which amount



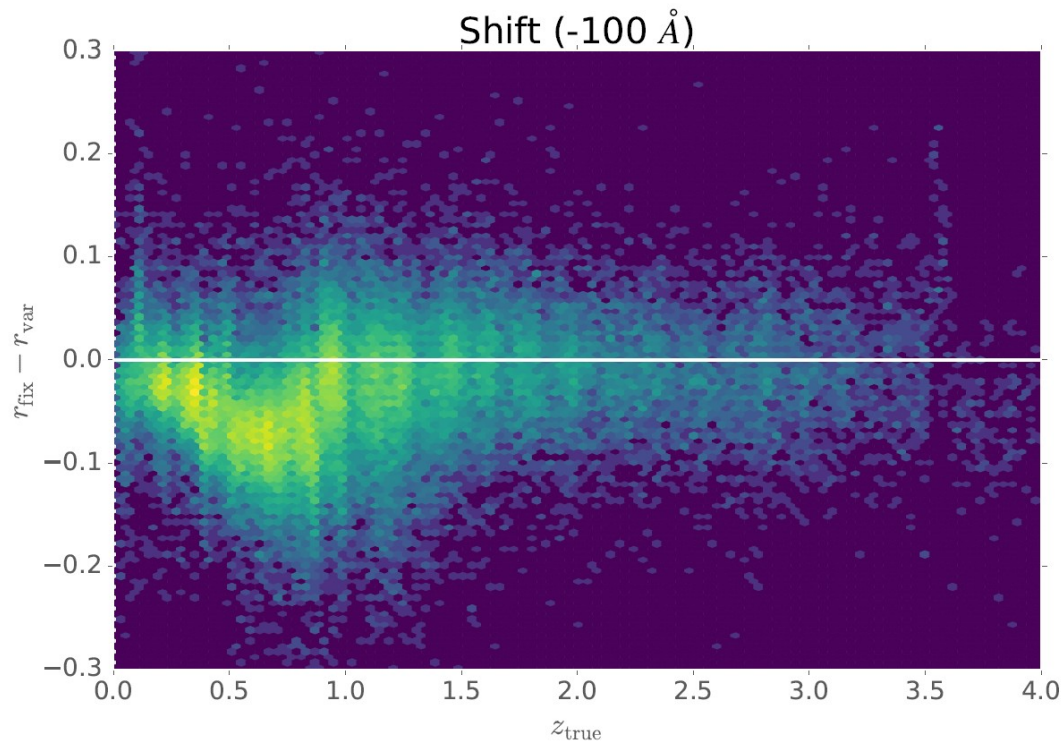
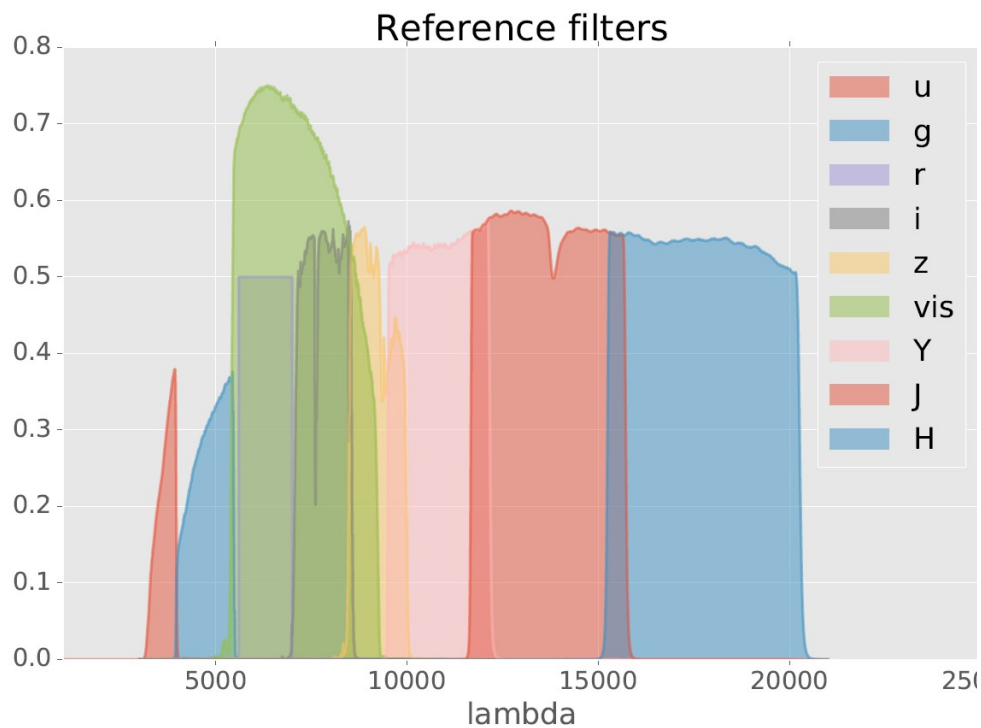
# MegaCam $r$ Filter Transmission(s)



Clear effect of the off-axis angle

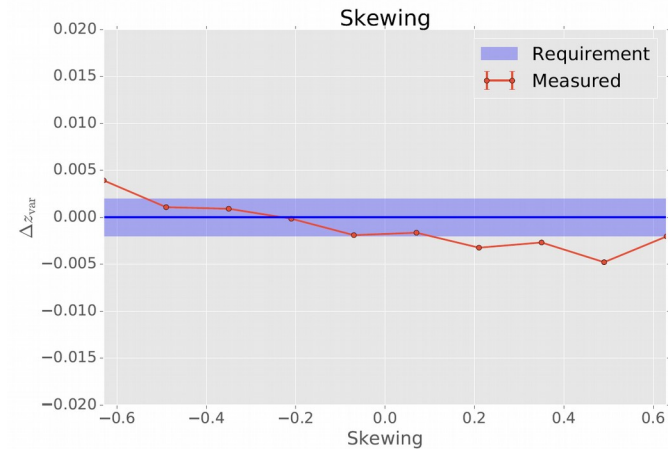
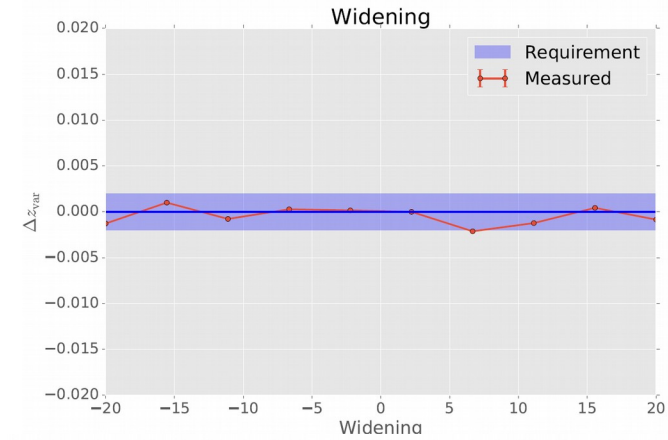
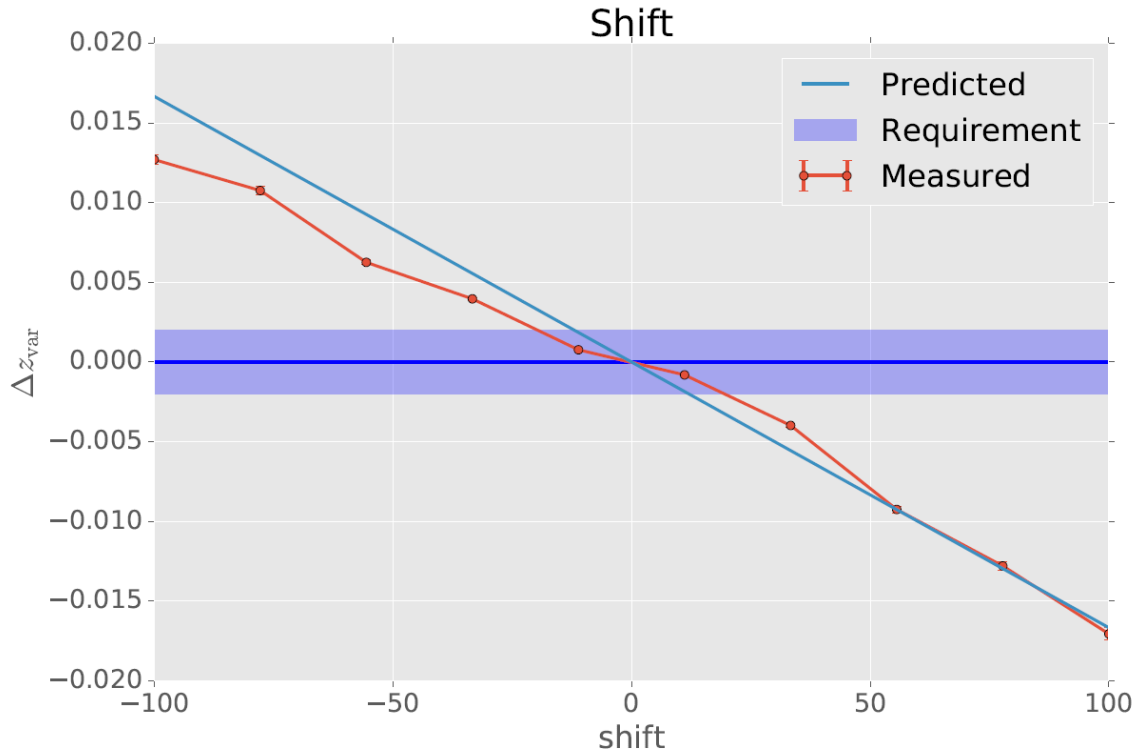
# Effect of Variable Transmissions

We simulate a catalogue based on COSMOS with varying  $r$  filter





# Bias due to Variable Transmissions



Main effect is a shift of the filter central wavelength

# How to deal with Variable Transmissions



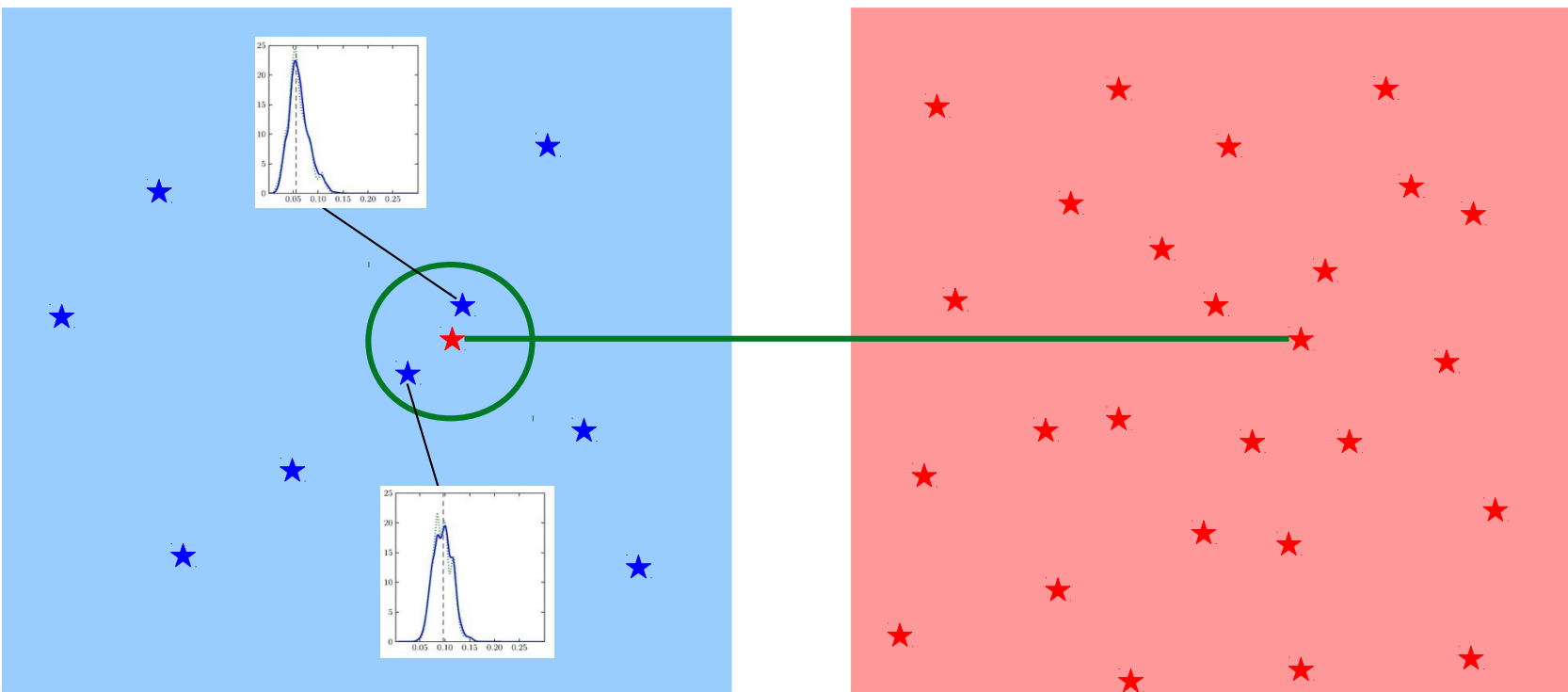
- Template fitting can deal easily with variable transmissions in principle, but **prohibitive computation time**, because the grid is different for each object
- ML has trouble, because **colour uniformity** is a pre-requisite
- But if we know the **SEDs of enough reference galaxies**, we can construct their colours in **any colour space**
  - However, we would need to train the ML algorithms for each object, making it computationally impossible
  - Unless the ML algorithm does not need training : **Nearest Neighbour** !!
- The **reference galaxies** are galaxies in well-known deep surveys, observed in many bands and a lot of redshifts: COSMOS, SXDF, VVDS-2H

# Nearest Neighbour

Reference Sample  
 $10^5$  objects  
Known PDZs

COLOUR SPACE

Euclid Sample  
Few  $10^9$  objects  
Unknown PDZs

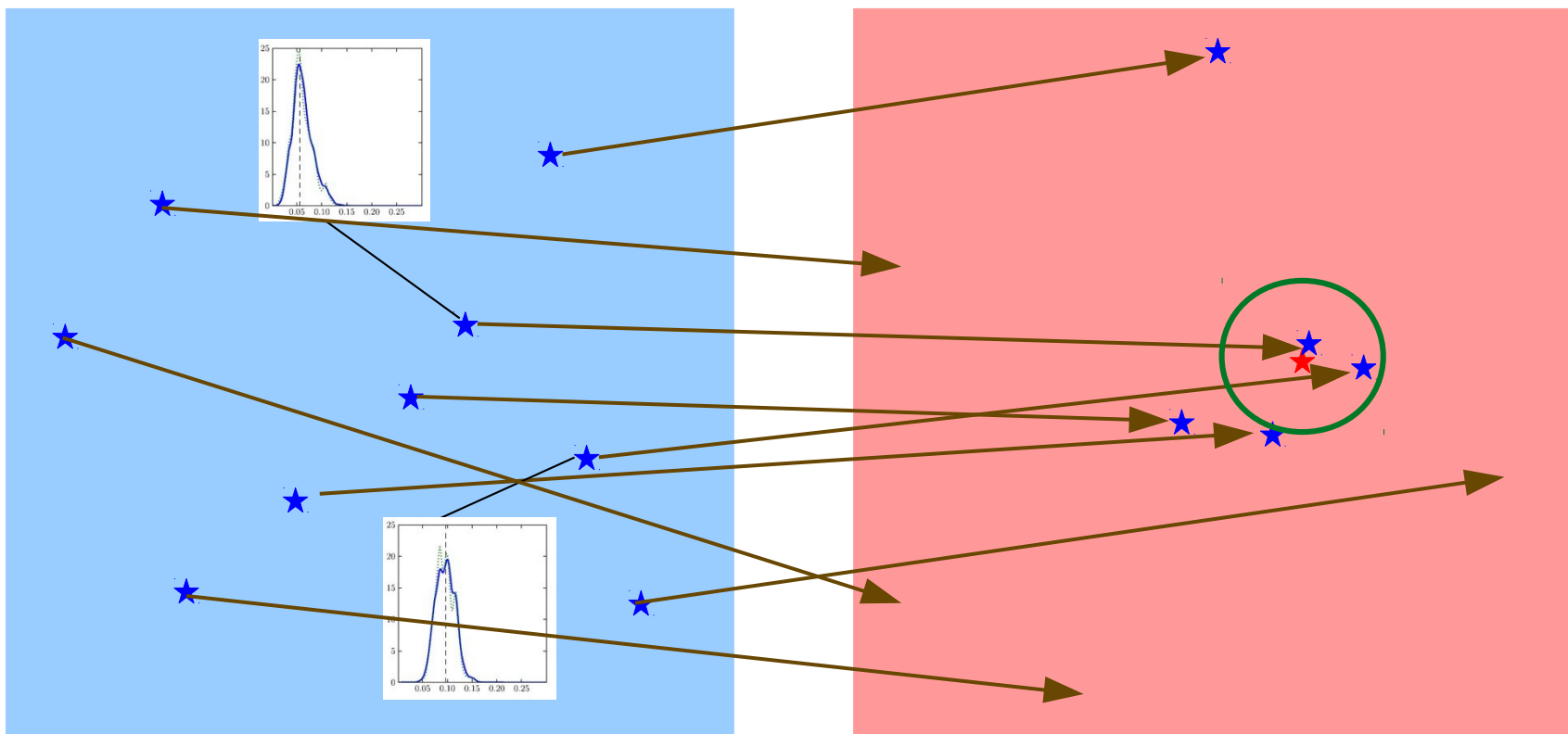


# NNPZ for Euclid

Reference Sample  
 $10^5$  objects  
Known PDZs  
Known SEDs

## COLOUR SPACE

Euclid Sample  
Few  $10^9$  objects  
Unknown PDZs



# Conclusions



- Some very challenging tasks
  - Calibration
  - Filter variations
  - Classification
  - Colour reconstruction
- Good baseline, but remains close to the feasibility limit