



# Astrometric Redshifts of Supernovae

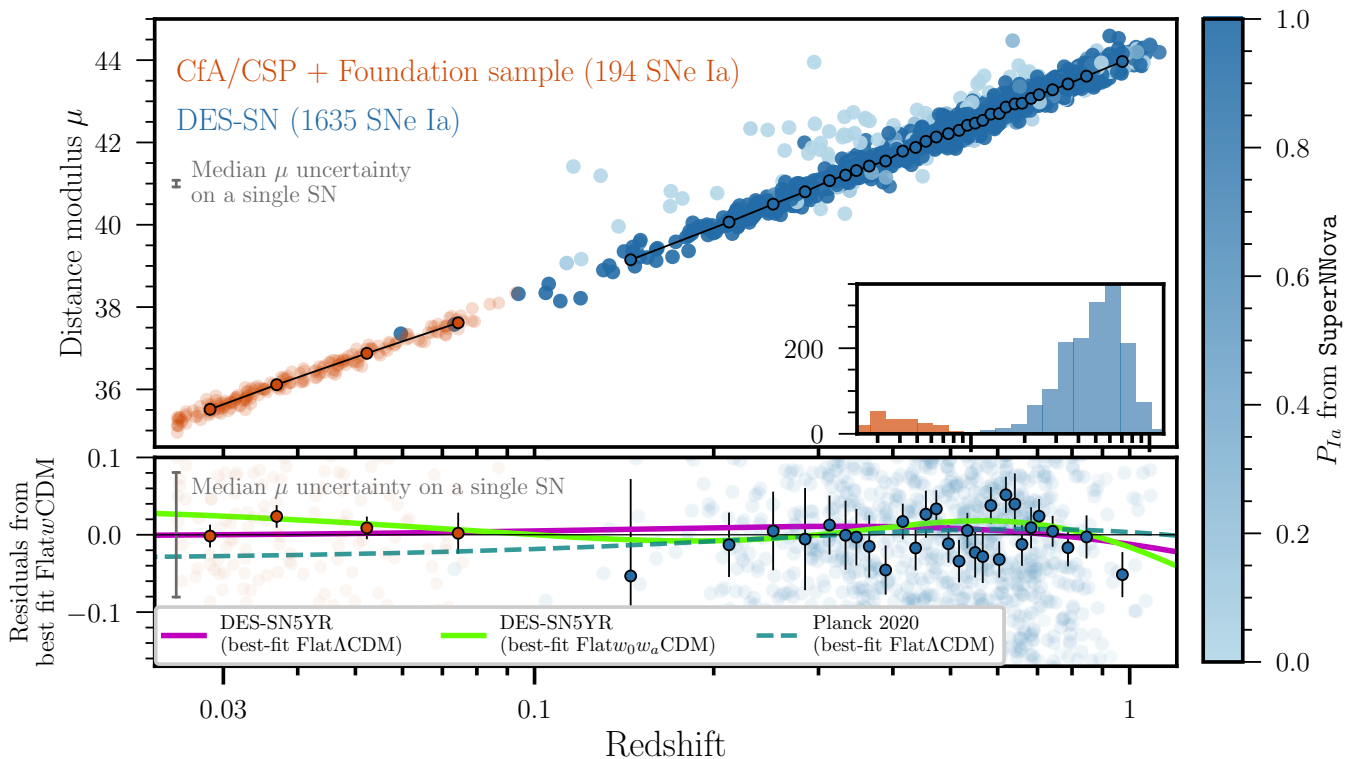
Jaemyoung (Jason) Lee, Masao Sako, Rick Kessler, and the LSST  
Collaboration

May 9th, 2024  
Fink-Brazil Workshop

[DESC Project 319 \(PUBID:149\)](#)

[arXiv:2405.04522](#)

# Redshifts in SN Ia Cosmology





# Motivation

- Too many Rubin (LSST) SNe Ia for spectroscopic follow-up => non-spectroscopic redshift measurements (photo- $z$ 's) crucial



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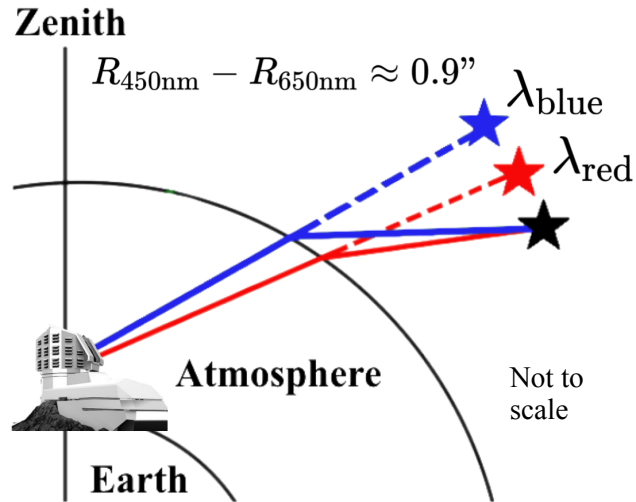


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- **(Nearly) independent** redshift measurement from conventional photo- $z$ 's
- Astrometric redshifts shown to be useful for sources with distinct emission lines (quasars)
- No additional measurements needed

# Differential Chromatic Refraction (DCR)

Atmospheric refractive index:  $\lambda$ -dependent



More DCR shift at higher air mass (AM),  
or amount of air along the line of sight (AM: 1.0 at zenith)

# DCR shift by redshift

$$\bar{R}_{\text{DCR}} = \frac{\int_0^\infty R_0(\lambda) \tan z_a F(\lambda) S(\lambda) d\lambda}{\int_0^\infty F(\lambda) S(\lambda) d\lambda}$$

$S(\lambda)$  : Source Flux,  $F(\lambda)$  : Filter Function

$R_0(\lambda)$  :  $f(n(\lambda))$ ,  $z_a$  : zenith angle

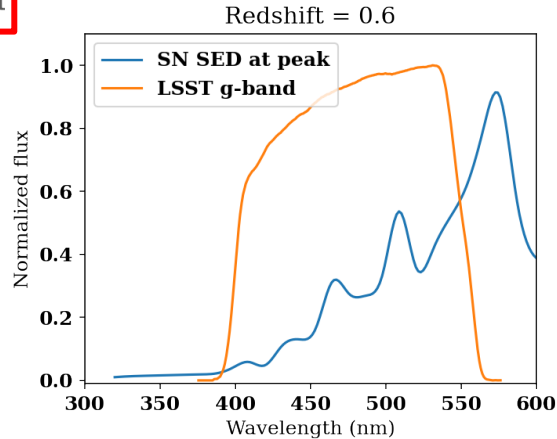
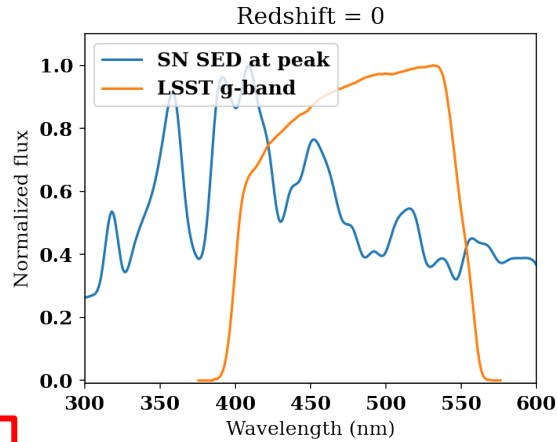


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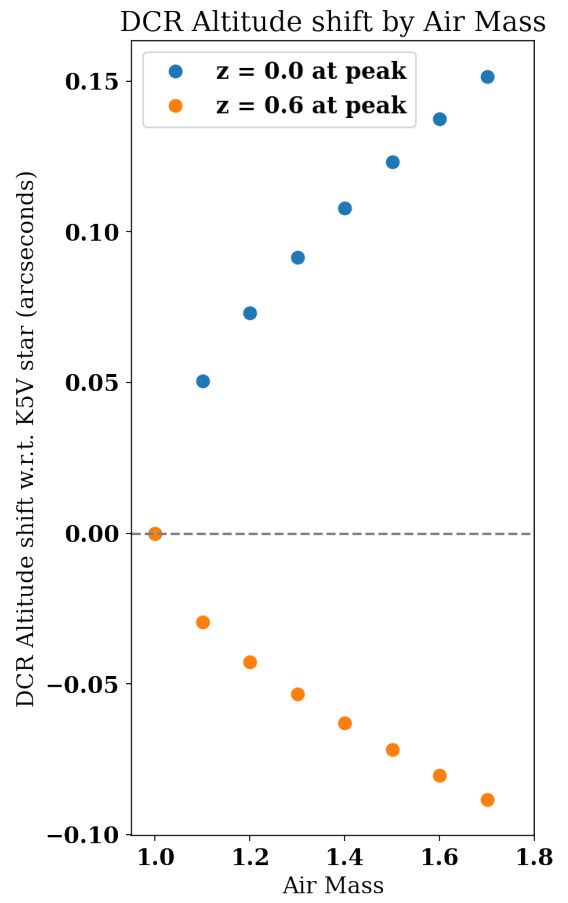
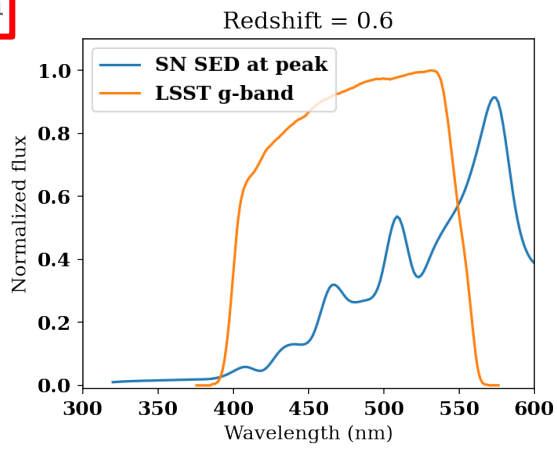
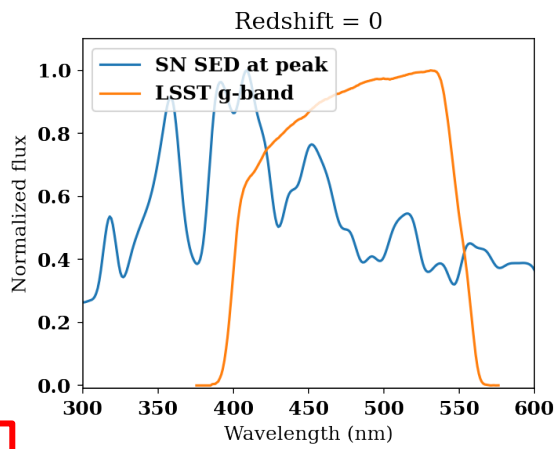


# DCR shift by redshift

$$\bar{R}_{DCR} = \frac{\int_0^\infty R_0(\lambda) \tan z_a F(\lambda) S(\lambda) d\lambda}{\int_0^\infty F(\lambda) S(\lambda) d\lambda}$$

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# Methodology

- SNANA simulations (ELAsTiCC) for realistic astrometric simulations (OBS)
- SED (spectral energy distribution) from underlying model to calculate DCR shifts according to  $[z, AM, \text{color } (c), \text{stretch } (x_1)]$  analytically for *ugriz* bands (MODEL)

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$$\text{➤ } \chi^2 = \sum \frac{(\Delta_{\text{OBS}} - \Delta_{\text{MODEL}})^2}{\sigma_{\text{stat}}^2 + \sigma_{\text{syst}}^2}, P(z, x_1, c) = e^{-\frac{\chi^2}{2}}$$

- Construct PDF => marginalize over  $c, x_1$

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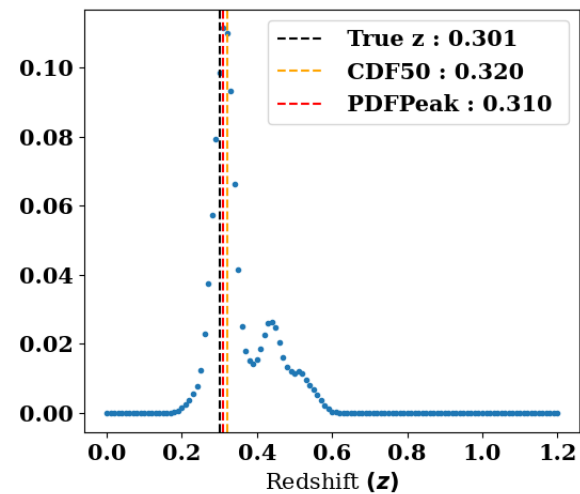
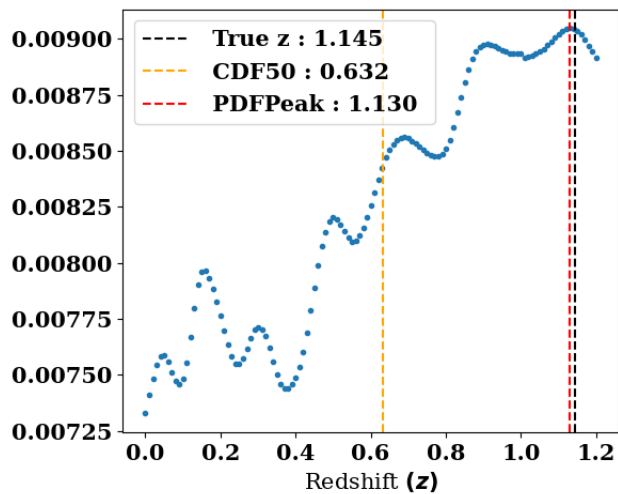
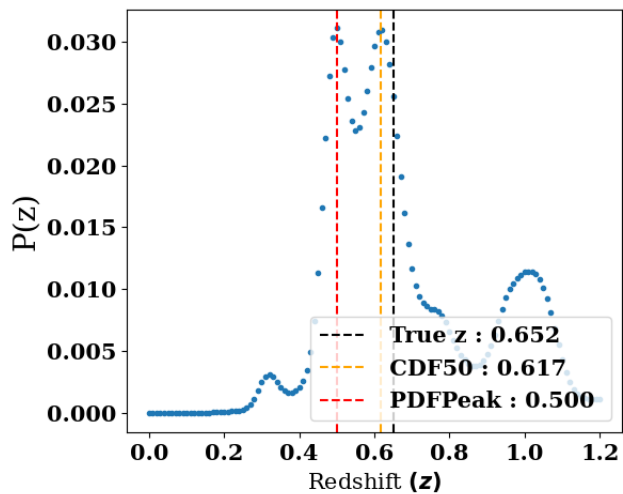
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- Construct PDF => marginalize over  $c, x_1$

- **Two methods to get astro- $z$  ( $z_{\text{DCR}}$ ):**

- **CDF50: 50th percentile of CDF  $\pm 1\sigma$ , PDFPeak: Peak of PDF  $\pm 1\sigma$**

# Two methods of $z_{\text{DCR}}$ estimation: CDF50 vs. PDFPeak





# Selection Cuts (for cosmology quality SNe)

- Light-curve requirements:  $T_{\text{rest}} < -3$  &  $T_{\text{rest}} > 10$  (69% of candidates remaining)
- 3 bands with  $S/N > 8$  for Realistic SIMs (43% of candidates remaining)
- LCFIT+z convergence + FITPROB  $> 0.01$  when combining with SN photo-z's (33 ~ 37% of candidates remaining)

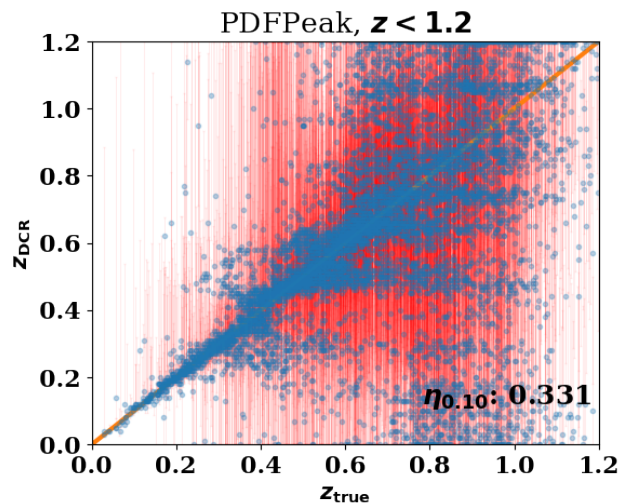
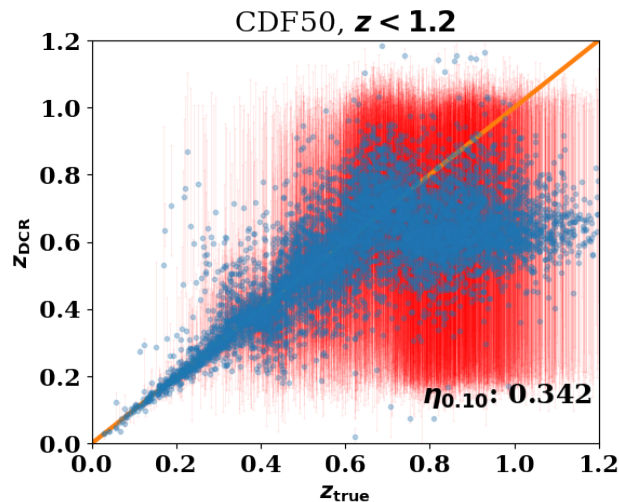
# Performance metrics

- Residuals  $\Delta z \equiv \frac{z_{\text{estimate}} - z_{\text{true}}}{1 + z_{\text{true}}}$ , Bias  $\equiv \langle \Delta z \rangle$
- (Accuracy) Outlier fraction  $\eta_x$ , fraction of candidates where  $|\Delta z| > x$  (Default:  $x = 0.10$ )
- (Precision) MAD (Median absolute deviation) deviation  
 $\sigma_{\text{MAD}} = 1.4826 \times \text{Median} |\Delta z - \text{Median}(\Delta z)|$



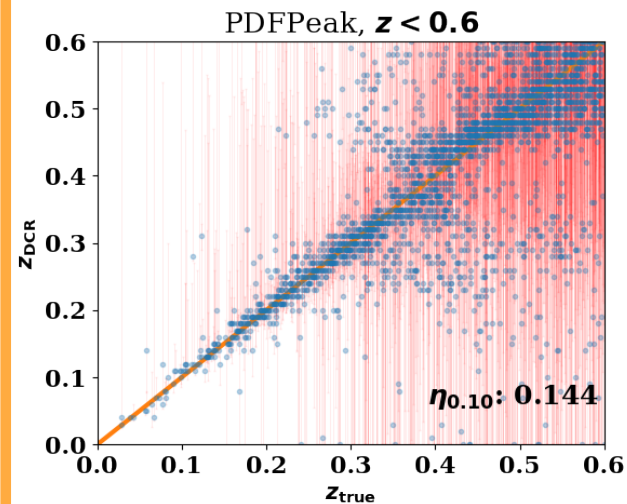
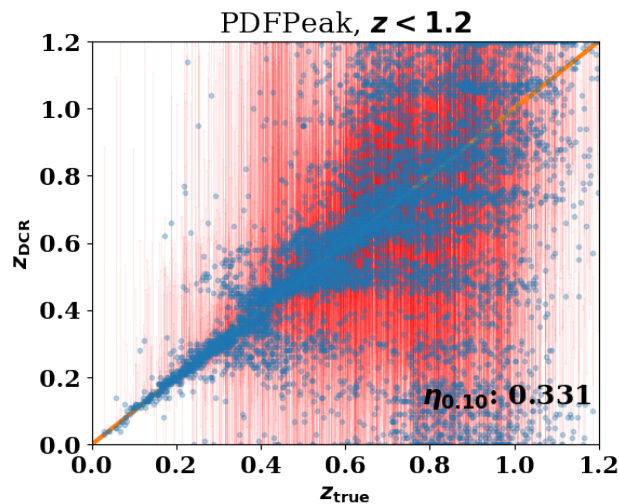
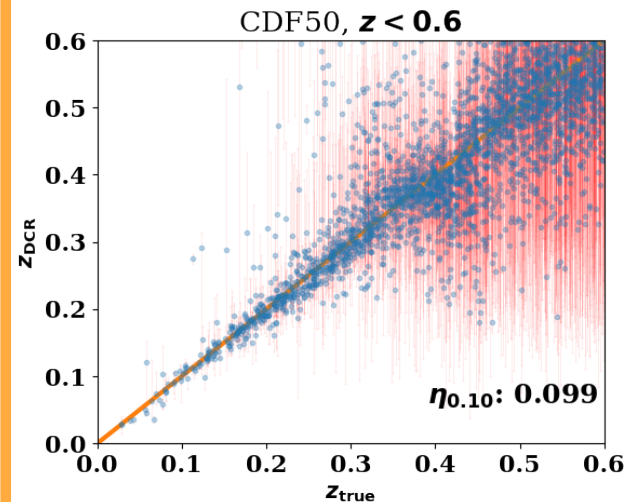
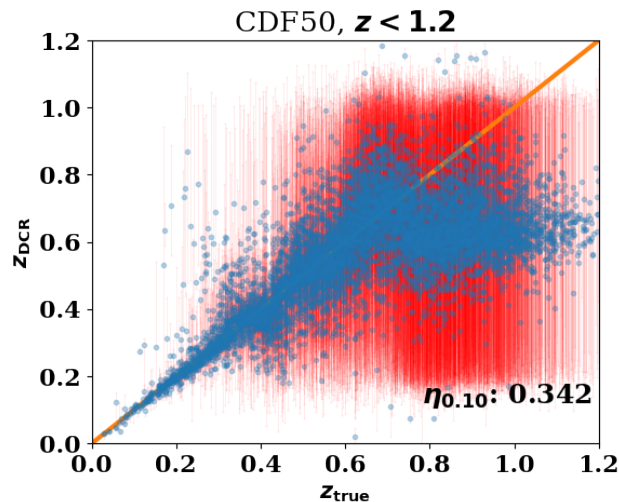
# Realistic SIMs

- $\sigma_{\text{sys}} = 5 \text{ mas}$
- Realistic  $T_{\text{obs}}$  (epoch),  
marginalized over  $x_1, c$
- Degradation at high-redshifts
- $\Delta z \equiv \frac{z_{\text{estimate}} - z_{\text{true}}}{1 + z_{\text{true}}}$
- $\eta_x$ : fraction of candidates  
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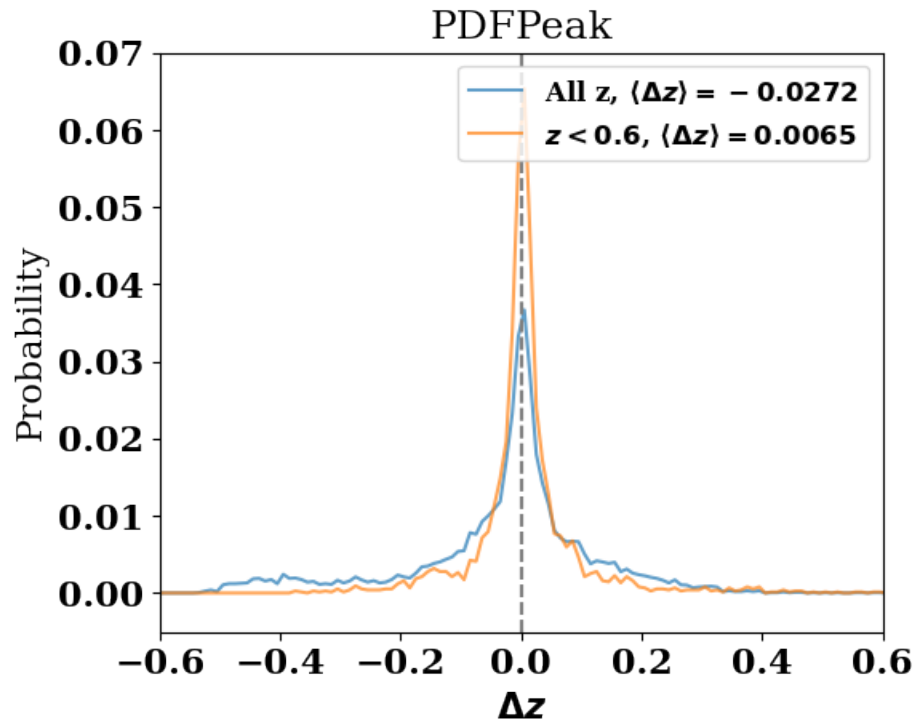
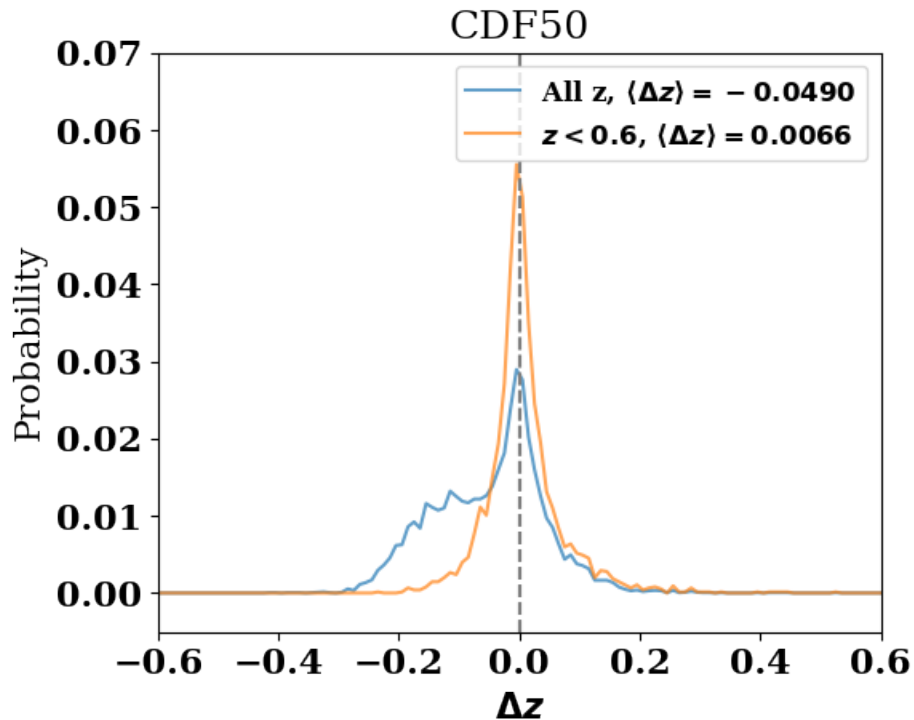


# Realistic SIMs

- $\sigma_{\text{syst}} = 5 \text{ mas}$
- Realistic  $T_{\text{obs}}$  (epoch),  
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- Degradation at high-redshifts
- $\Delta z \equiv \frac{z_{\text{estimate}} - z_{\text{true}}}{1 + z_{\text{true}}}$
- $\eta_x$ : fraction of candidates  
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# Realistic SIMs (Histograms)

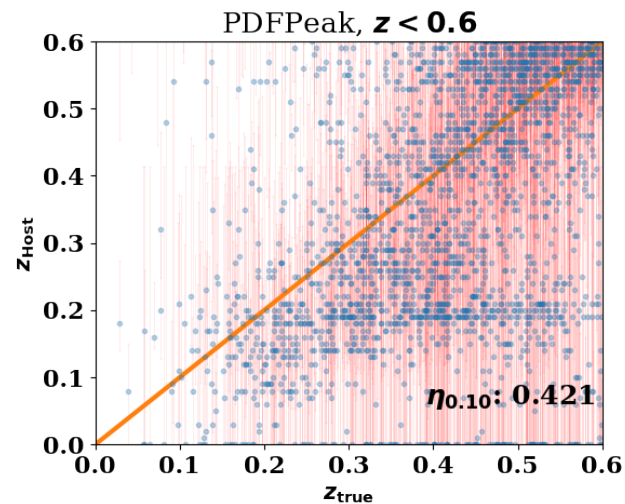
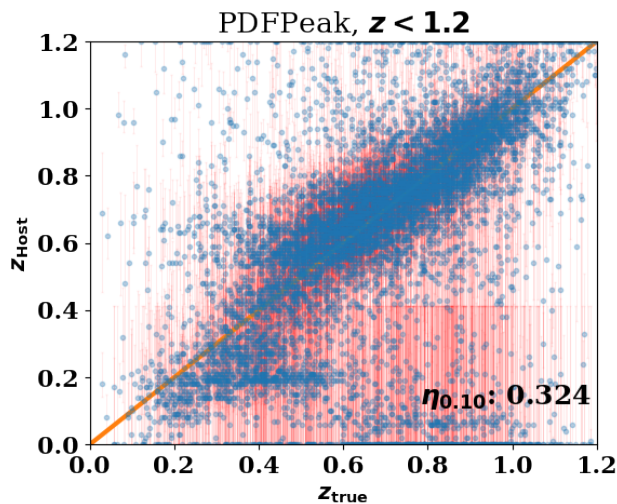
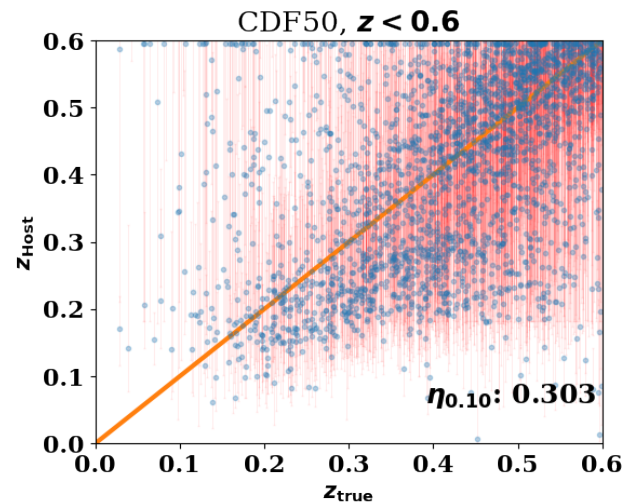
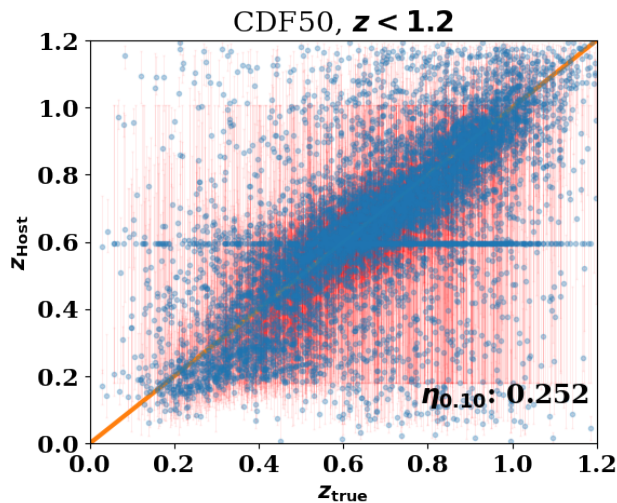




# Combining with photo- $z$ 's

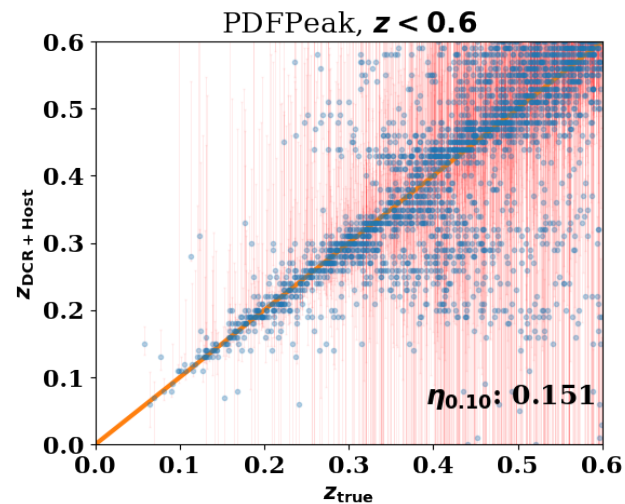
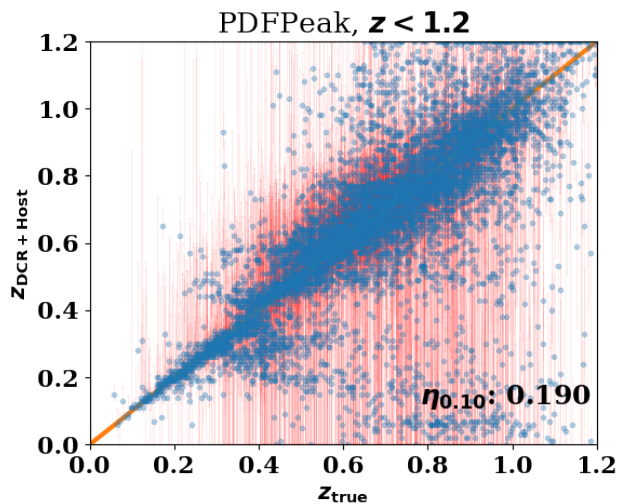
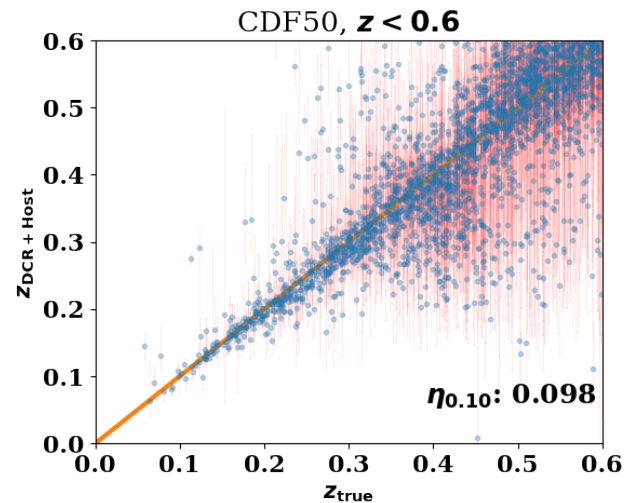
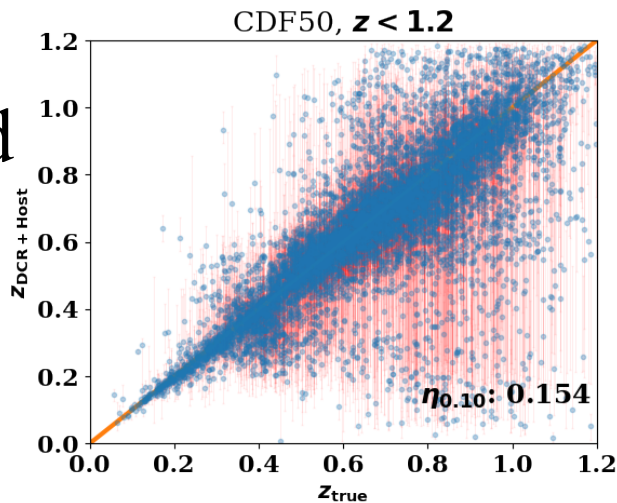
# Host photo- $z$ ( $z_{\text{Host}}$ )

- From RAIL, better performance at high-redshifts

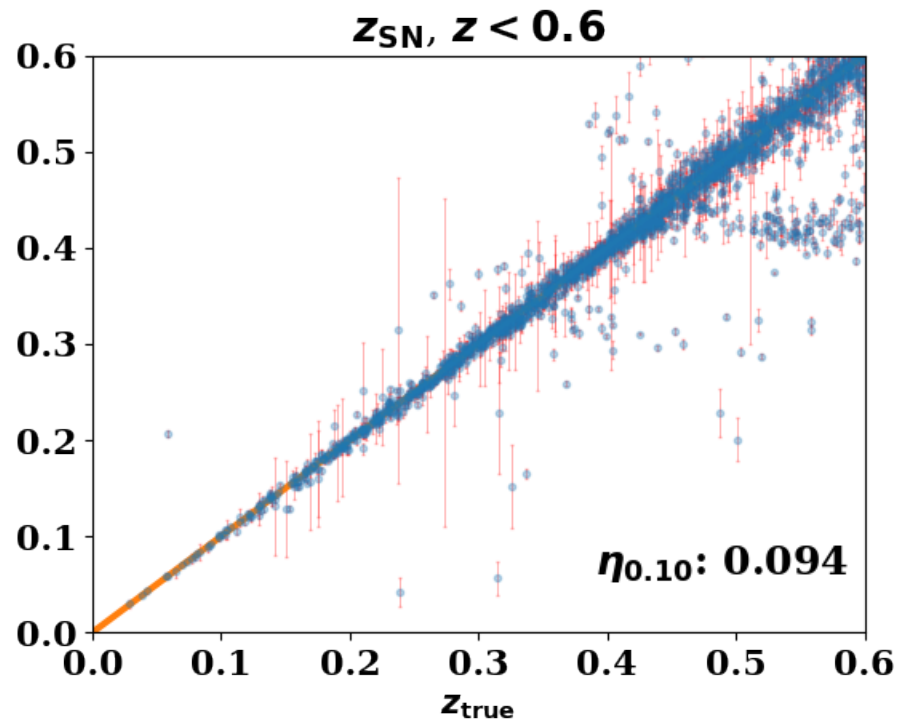
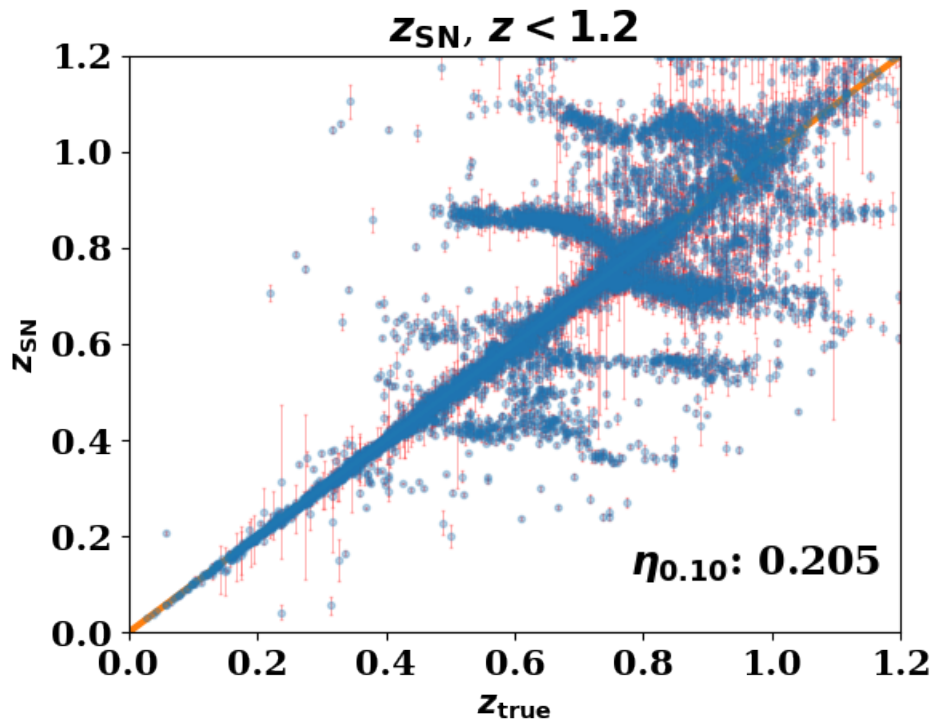


# Astro- $z$ Combined ( $z_{\text{DCR+Host}}$ )

- Noticeably better performance at low-redshifts

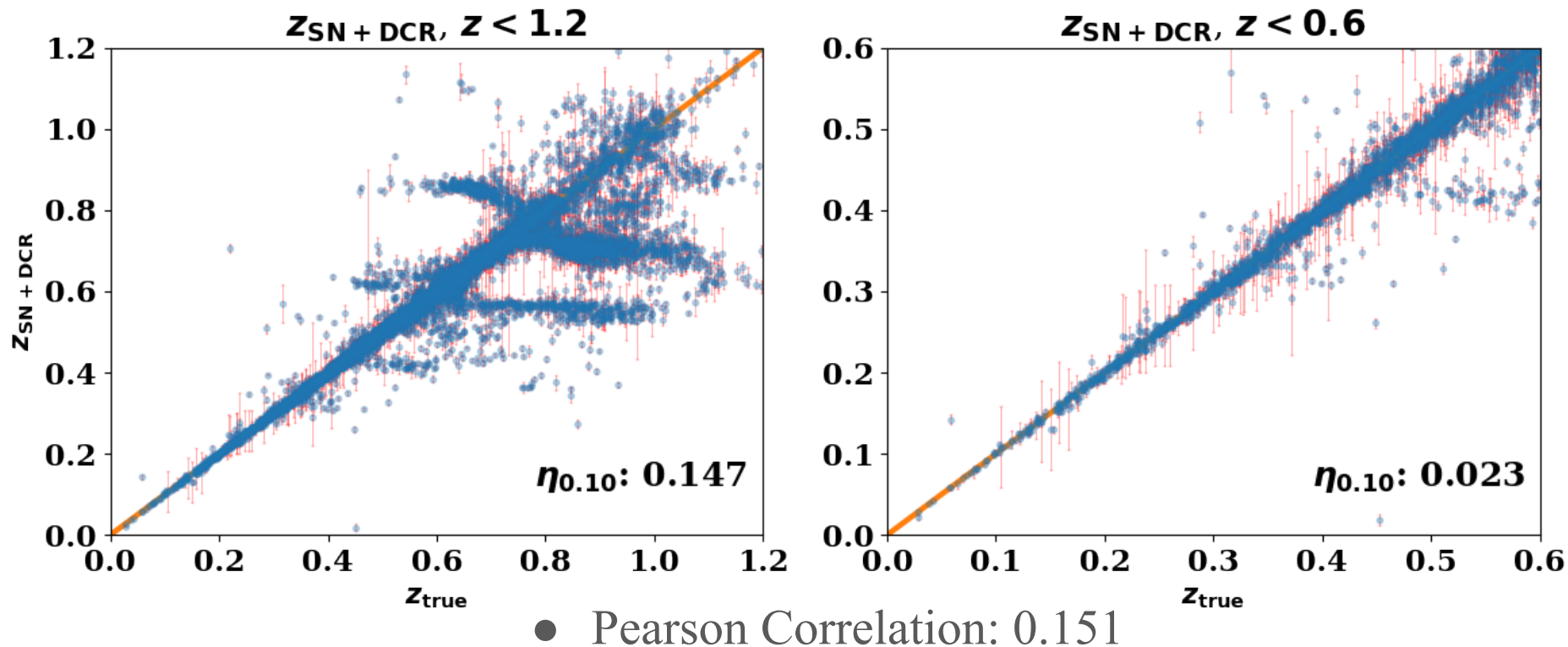


# SN photo- $z$ ( $z_{\text{SN}}$ )



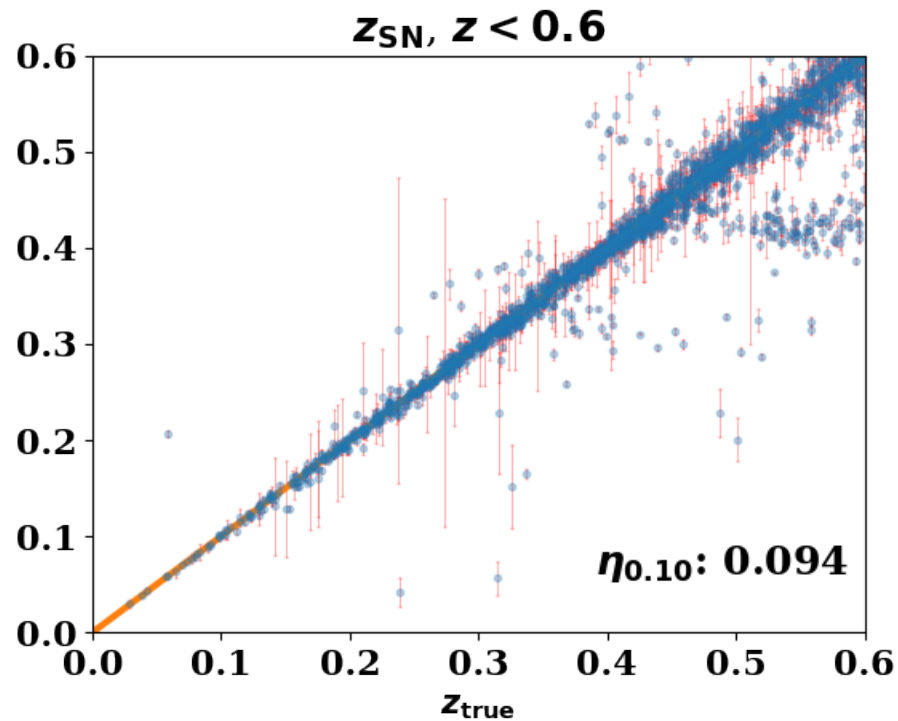
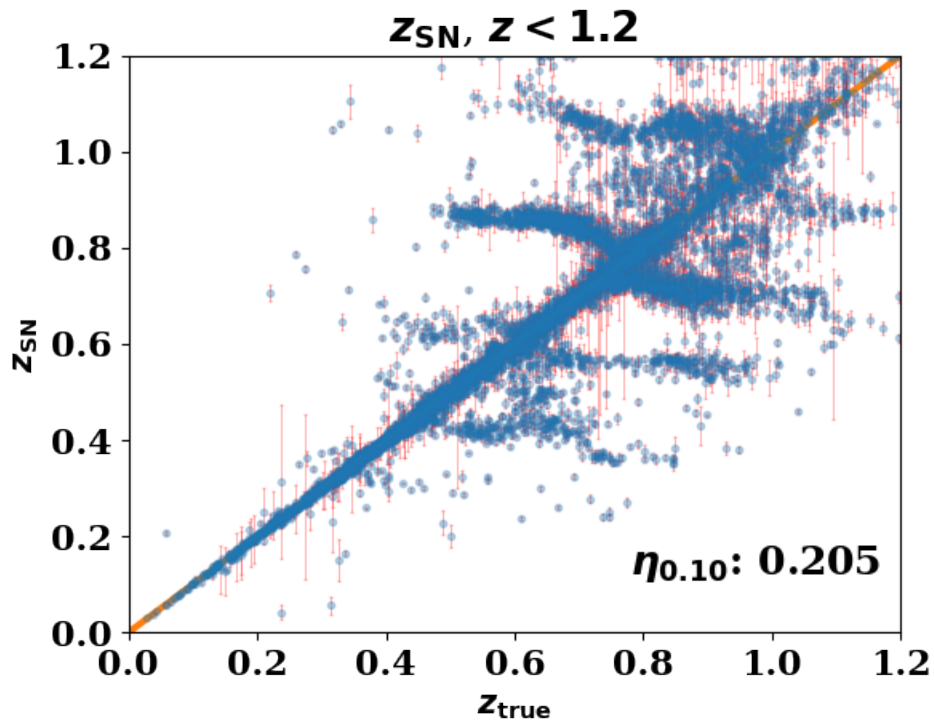
- Using LCFIT+z: Fits LC parameters +  $z$  simultaneously

# SN photo- $z$ + Astro- $z$ ( $z_{\text{SN+DCR}}$ )

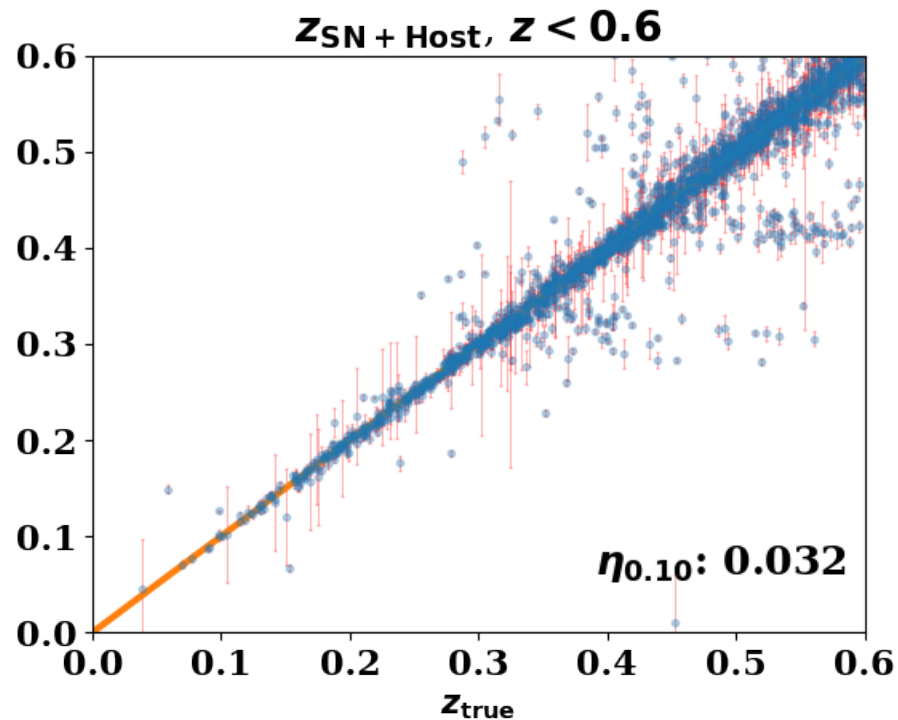
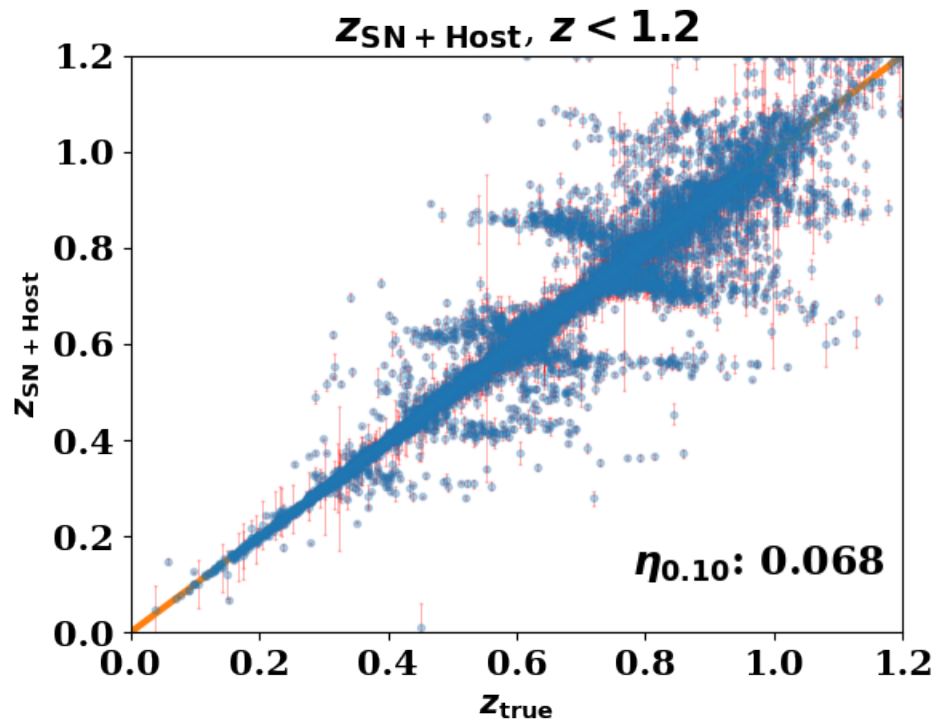




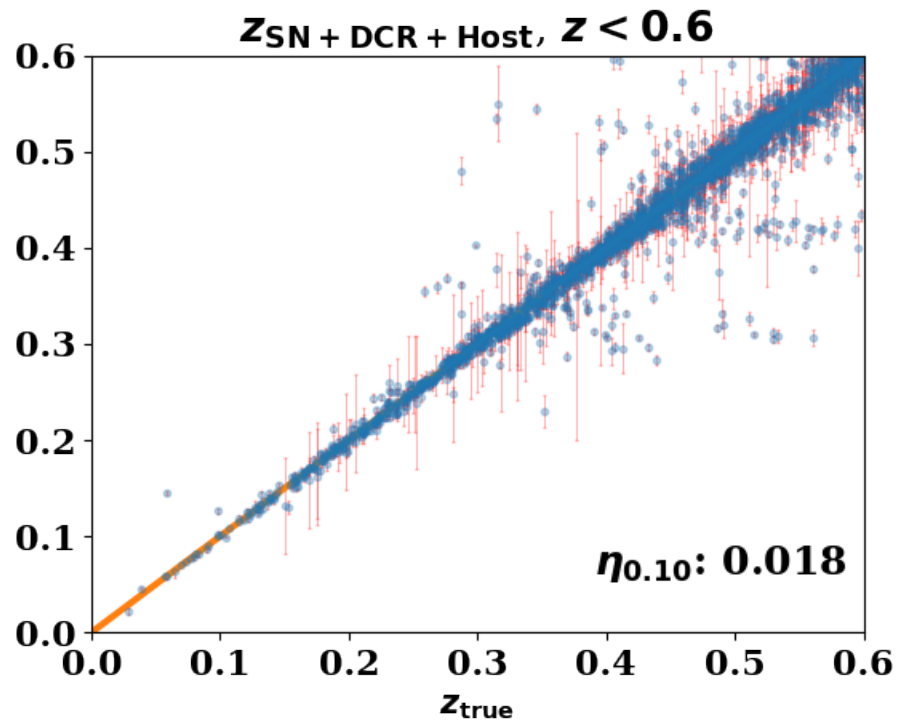
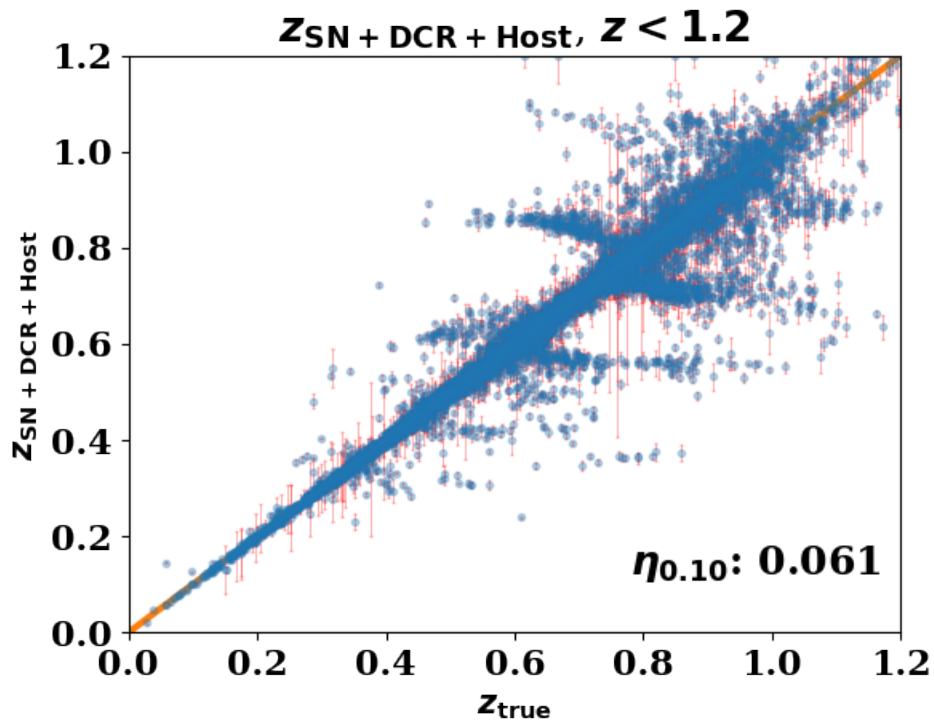
# SN photo- $z$ ( $z_{\text{SN}}$ )



# SN photo- $z$ + Host photo- $z$ ( $z_{\text{SN+Host}}$ )



# SN photo- $z$ + Host photo- $z$ + Astro- $z$ ( $z_{\text{SN}+\text{DCR}+\text{Host}}$ )



# # Events with LCFIT+z convergence

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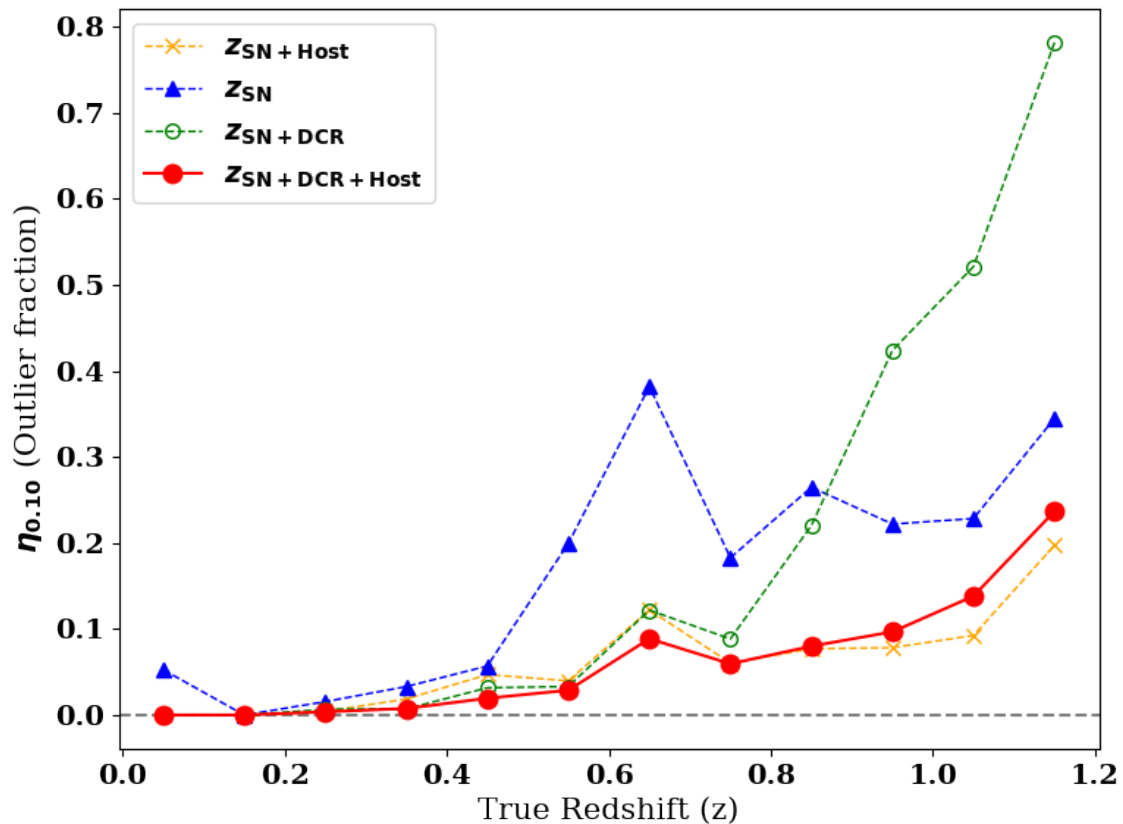
Number of LCFIT+z events with:

Type	No cuts	Default cuts	Default + FITPROB cuts
$z_{\text{SN}}$	12,589	7,295	6,657
$z_{\text{SN+Host}}$	12,324	7,176	6,615
$z_{\text{SN+DCR}}$	12,246	7,660	7,255
$z_{\text{SN+DCR+Host}}$	12,779	7,568	7,215

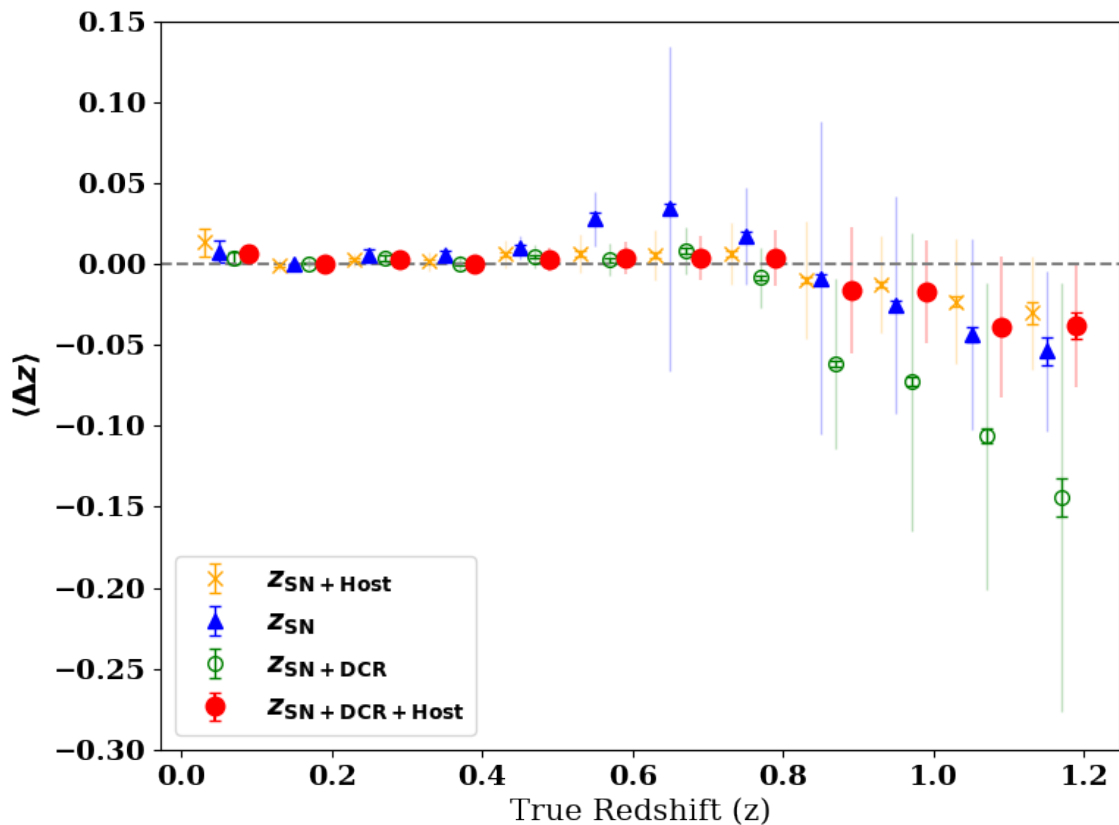
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- Combining with  $z_{\text{DCR}}$  recovers about 600 candidates or 9%!

# Outlier Fractions by True Redshift



# Bias and MAD Deviation by True Redshift



# Systematic Effects

Type	Bias (CDF50)	Bias (PDFPeak)	$\eta_{0.10}$ (CDF50)	$\eta_{0.10}$ (PDFPeak)	$\sigma_{\text{MAD}}$ (CDF50)	$\sigma_{\text{MAD}}$ (PDFPeak)
Perfect	-0.003±0.000	-0.003±0.000	0.016	0.016	0.008	0.008
$\sigma_{\text{sys}} = 2$ mas	-0.007±0.000	-0.003±0.000	0.015	0.016	0.009	0.009
$\sigma_{\text{sys}} = 5$ mas	-0.010±0.000	-0.004±0.000	0.028	0.024	0.011	0.011
$\sigma_{\text{sys}} = 10$ mas	-0.019±0.000	-0.007±0.000	0.065	0.046	0.019	0.015
$c$ marginalized	-0.004±0.000	-0.004±0.000	0.018	0.018	0.007	0.007
$x_1$ marginalized	-0.003±0.000	-0.003±0.000	0.015	0.015	0.008	0.008
realistic $T_{\text{obs}}$	-0.002±0.000	-0.002±0.000	0.041	0.041	0.010	0.010
All syst. for realistic	-0.010±0.001	-0.004±0.001	0.064	0.073	0.016	0.014

**Table 9.** Bias, outlier fractions, and  $\sigma_{\text{MAD}}$  for various systematics. The impact of  $\sigma_{\text{sys}} = 2$  or 5 mas, marginalization over  $c$  and  $x_1$  is small, while using realistic  $T_{\text{obs}}$  results in more degradation. Larger  $\sigma_{\text{sys}}$  results in substantially larger degradation compared to other systematics. ‘All syst. for realistic’ means using realistic  $T_{\text{obs}}$ , marginalizing over  $c$  and  $x_1$ , and assuming  $\sigma_{\text{sys}} = 5$  mas.

# Conclusion

- **Developed methodology** for SN Ia astro- $z$ 's using DCR
- **Reasonable**  $z_{\text{DCR}}$  **at**  $z_{\text{true}} < 0.6$  for LSST-like simulations
- **Combining**  $z_{\text{DCR}}$  with  $z_{\text{Host}}$  and  $z_{\text{SN}}$  substantially **improves**  $z_{\text{SN}}$



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- Will implement astro- $z$ 's into the LSST SN Ia photo- $z$  pipeline
  
- 1st time astro- $z$ 's were applied to SNe Ia and constructed the full PDF for astro- $z$ 's
- Can be adapted for other sources with distinct emission lines like SNe II



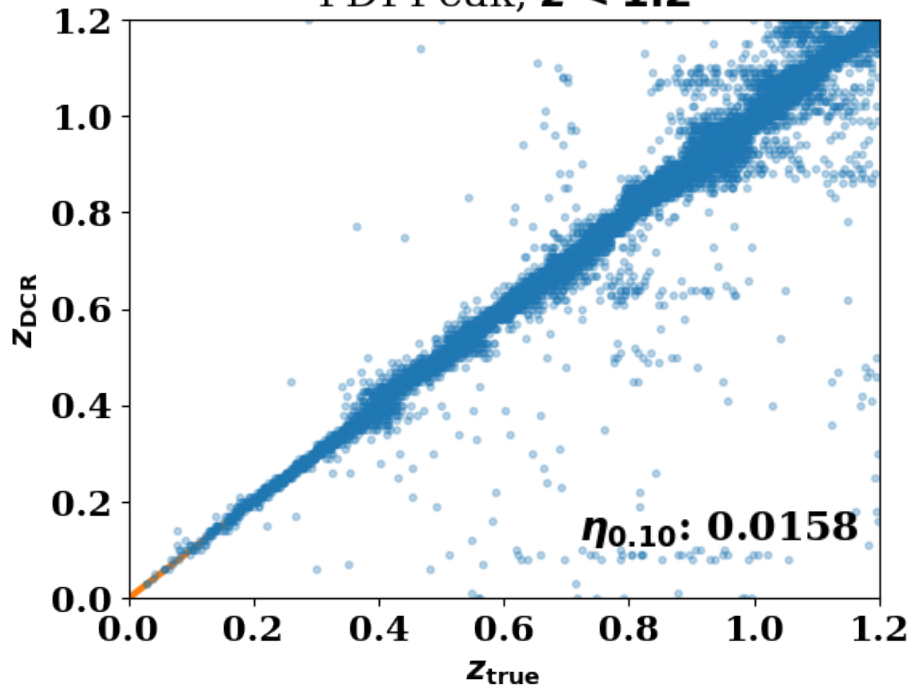
Thank you very much!  
Questions?



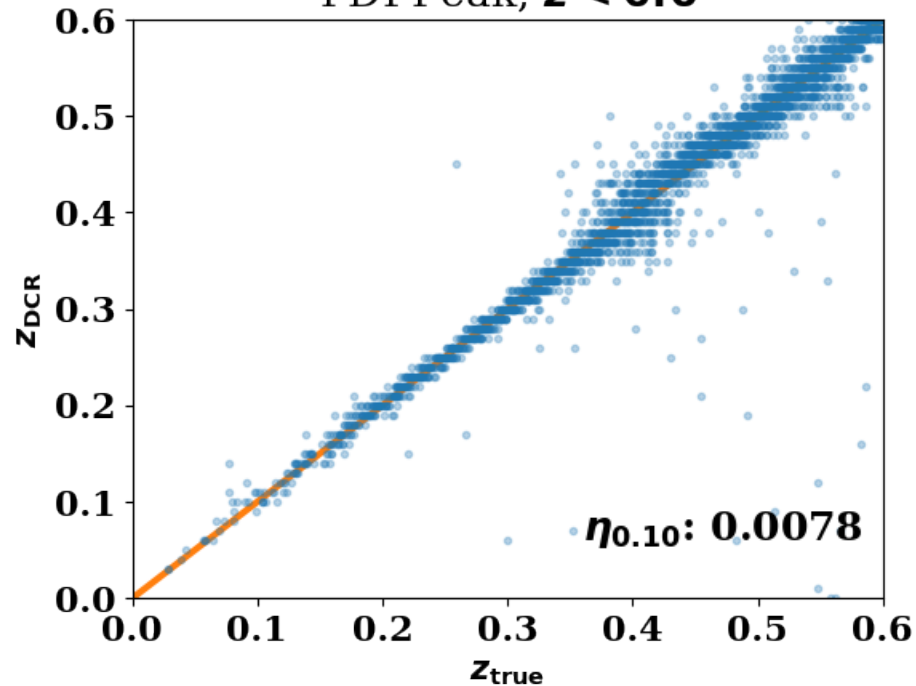
# Extra Slides

# Perfect SIMs

PDFPeak,  $z < 1.2$



PDFPeak,  $z < 0.6$



- $\sigma_{\text{stat}} = \sigma_{\text{sys}} = 0$  & assuming we know  $T_{\text{obs}}$  (epoch),  $x_1$ ,  $c$

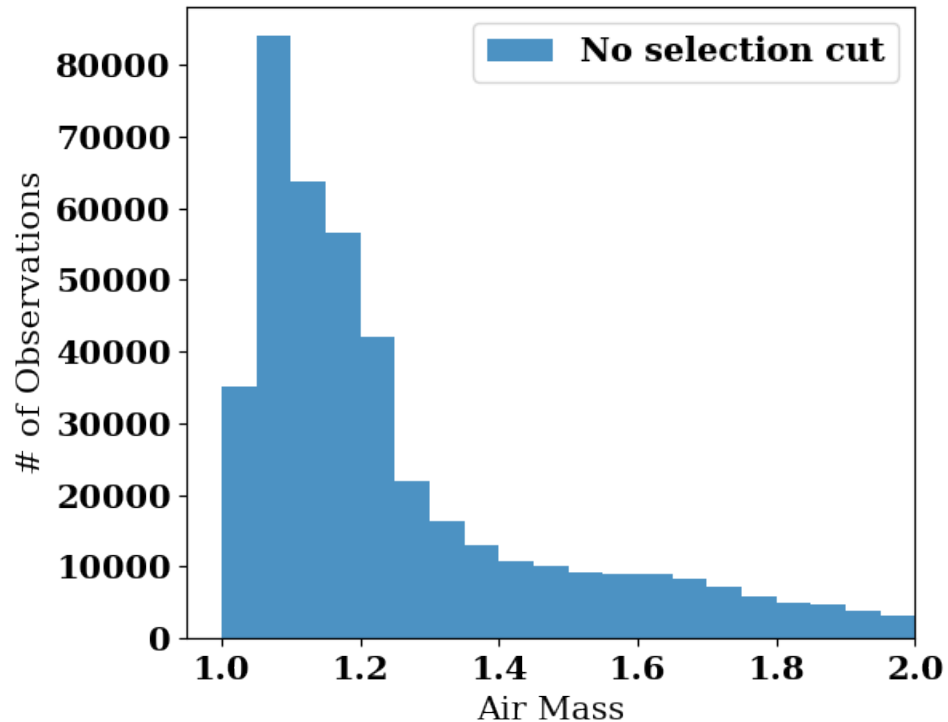
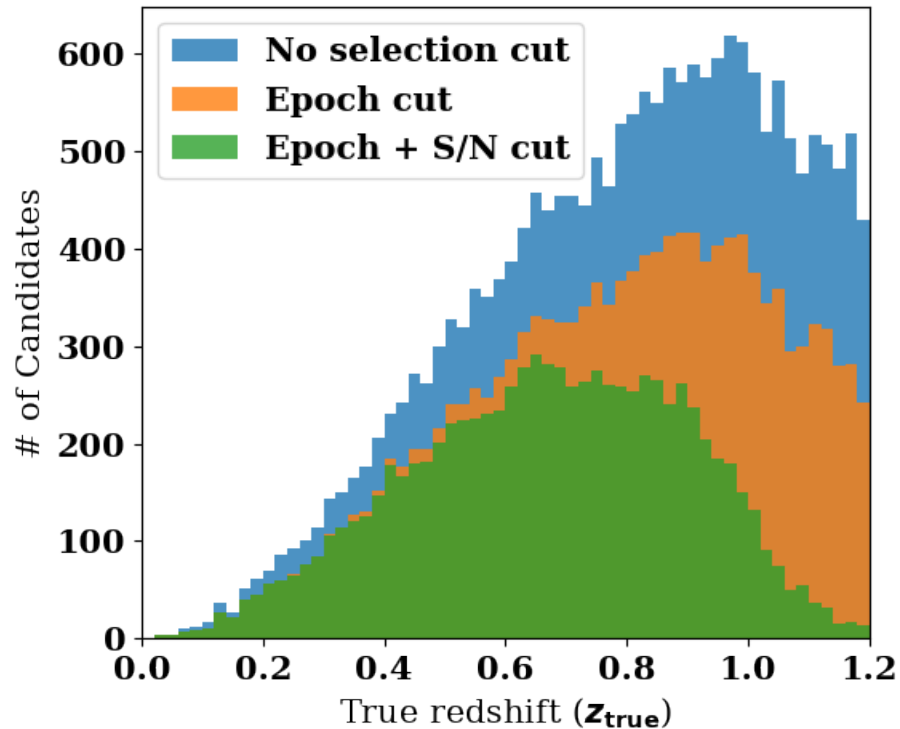
# Various Metrics

- low- $z$ :  $z_{\text{true}} < 0.6$ , high- $z$ :  $z_{\text{true}} > 0.6$

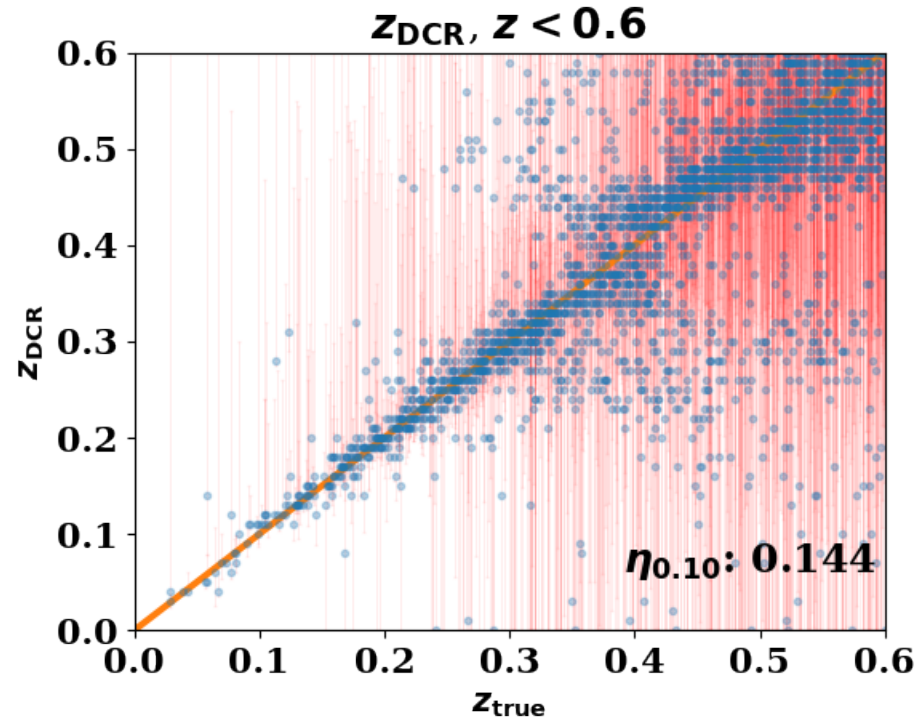
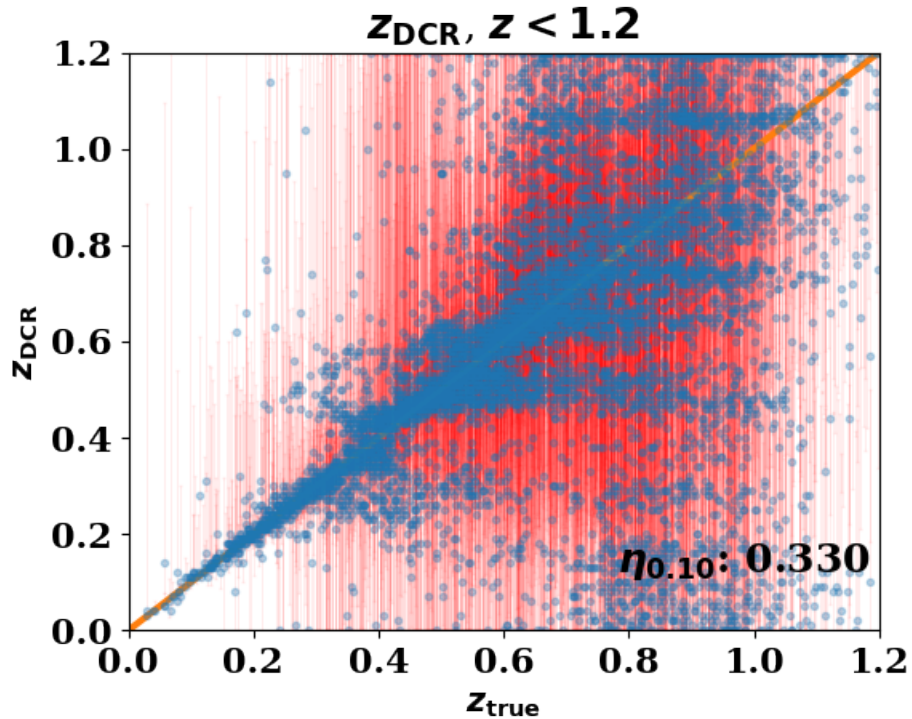
Type	Bias (low- $z$ )	Bias (high- $z$ )	$\eta_{0.10}$ (low- $z$ )	$\eta_{0.10}$ (high- $z$ )	$\sigma_{\text{MAD}}$ (low- $z$ )	$\sigma_{\text{MAD}}$ (high- $z$ )
$z_{\text{DCR}}$	$0.006 \pm 0.002$	$-0.047 \pm 0.002$	0.144	0.439	0.031	0.121
$z_{\text{Host}}$	$0.012 \pm 0.003$	$-0.052 \pm 0.002$	0.421	0.268	0.123	0.066
$z_{\text{DCR+Host}}$	$0.005 \pm 0.001$	$-0.025 \pm 0.002$	0.151	0.212	0.039	0.057
$z_{\text{SN}}$	$0.014 \pm 0.001$	$0.002 \pm 0.001$	0.094	0.265	0.007	0.064
$z_{\text{SN+Host}}$	$0.005 \pm 0.001$	$-0.004 \pm 0.001$	0.032	0.089	0.008	0.024
$z_{\text{SN+DCR}}$	$0.003 \pm 0.001$	$-0.036 \pm 0.001$	0.023	0.224	0.007	0.044
$z_{\text{SN+DCR+Host}}$	$0.003 \pm 0.001$	$-0.008 \pm 0.001$	0.018	0.087	0.007	0.023

**Table 8.** Bias and error on the mean,  $\eta_{0.10}$ , and  $\sigma_{\text{MAD}}$  for low- $z$  ( $z_{\text{true}} < 0.6$ ) and high- $z$  ( $z_{\text{true}} > 0.6$ ) for various combinations of the three methods discussed in this work. For  $z_{\text{DCR}}$ ,  $z_{\text{Host}}$ , and  $z_{\text{DCR+Host}}$ , we show the PDFPeak values.  $z_{\text{DCR}}$  shows good performance at low- $z$  while  $z_{\text{Host}}$  shows good performance at high- $z$ , which leads to  $z_{\text{SN+DCR}}$  being significantly improved at low- $z$  compared to  $z_{\text{SN}}$  and  $z_{\text{SN+Host}}$  being significantly improved at high- $z$ .  $z_{\text{SN+DCR+Host}}$  show similar performance to  $z_{\text{SN+DCR}}$  and  $z_{\text{SN+Host}}$  at low- $z$  and high- $z$  respectively.

# Redshift and AM distributions (maybe get rid of this)



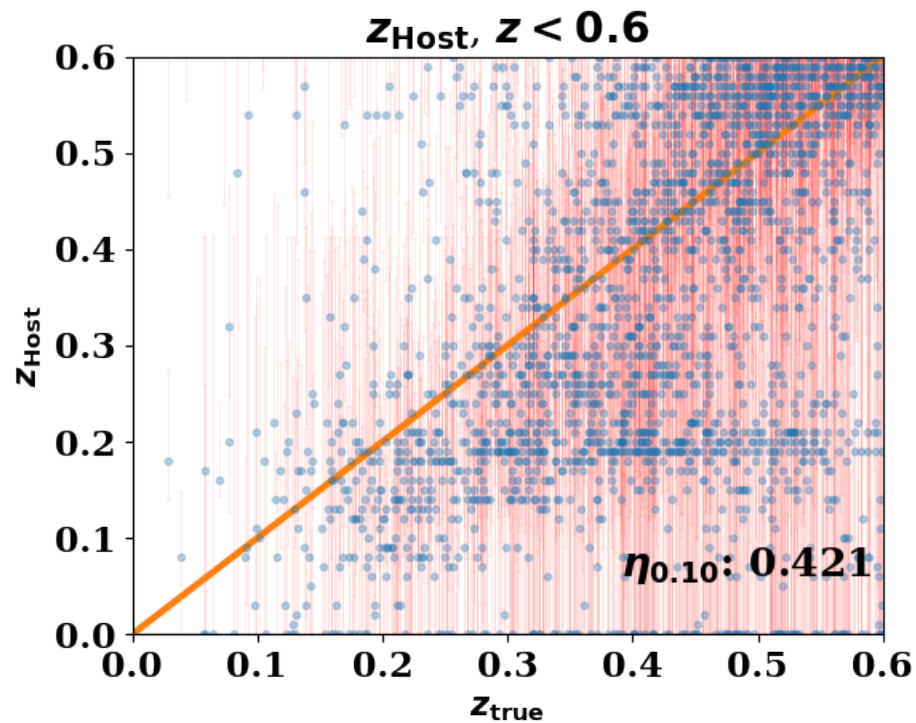
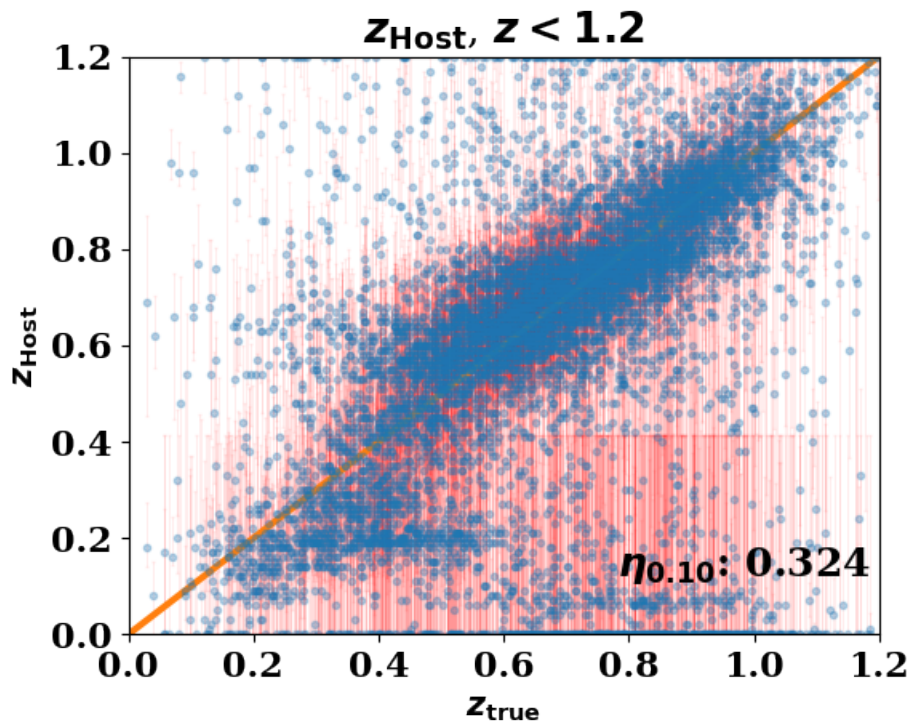
# Astro-z ( $z_{\text{DCR}}$ )



- $\sigma_{\text{sys}} = 5 \text{ mas}$ , Realistic  $T_{\text{obs}}$  (epoch), marginalized over  $x_1, c$ , better at low- $z$

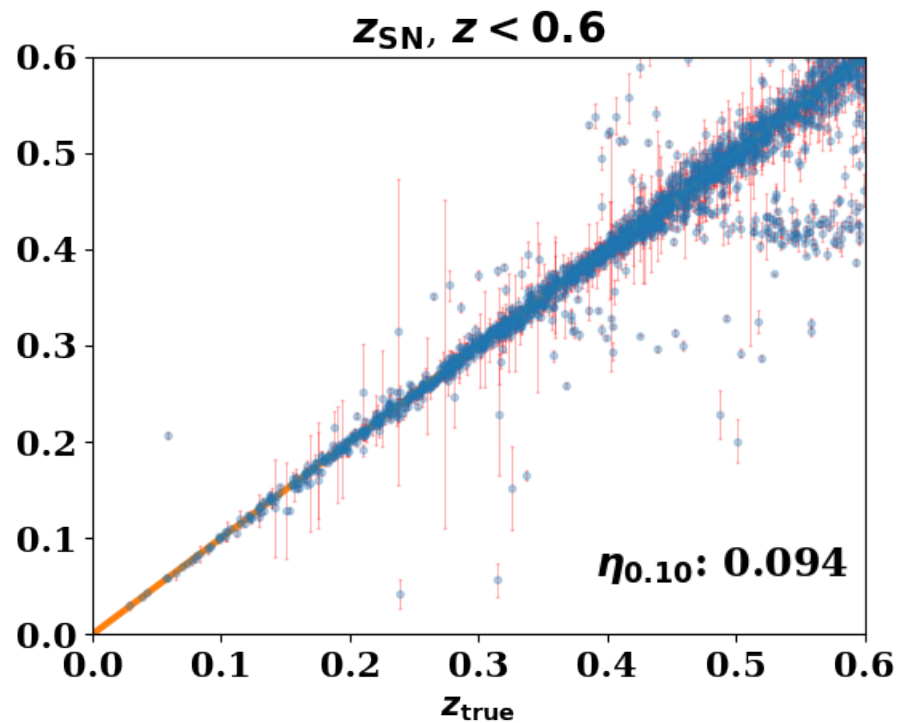
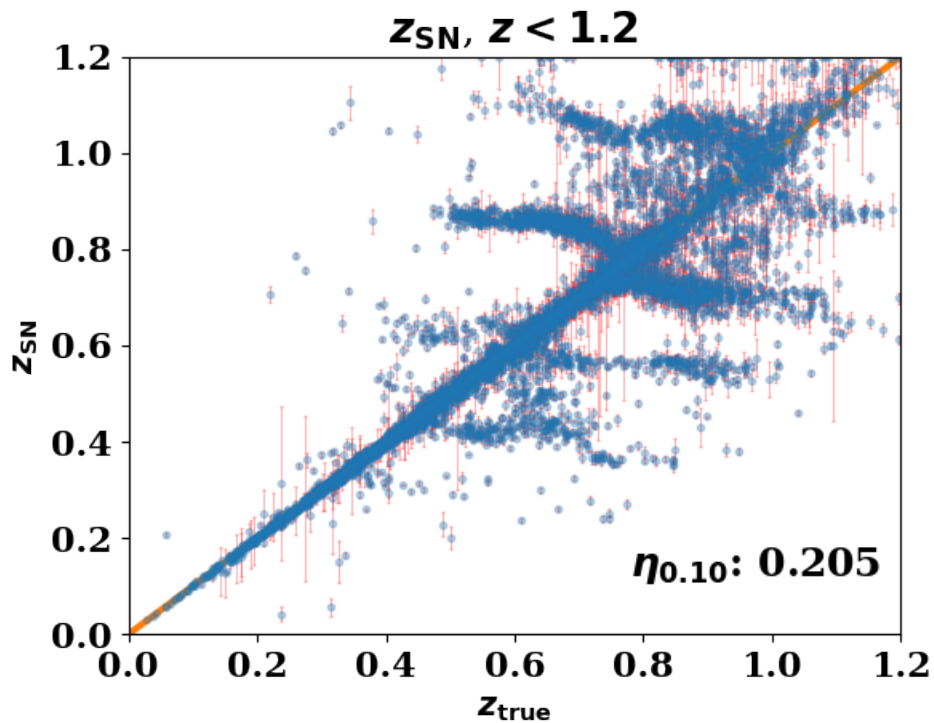


# Host photo- $z$ ( $z_{\text{Host}}$ )

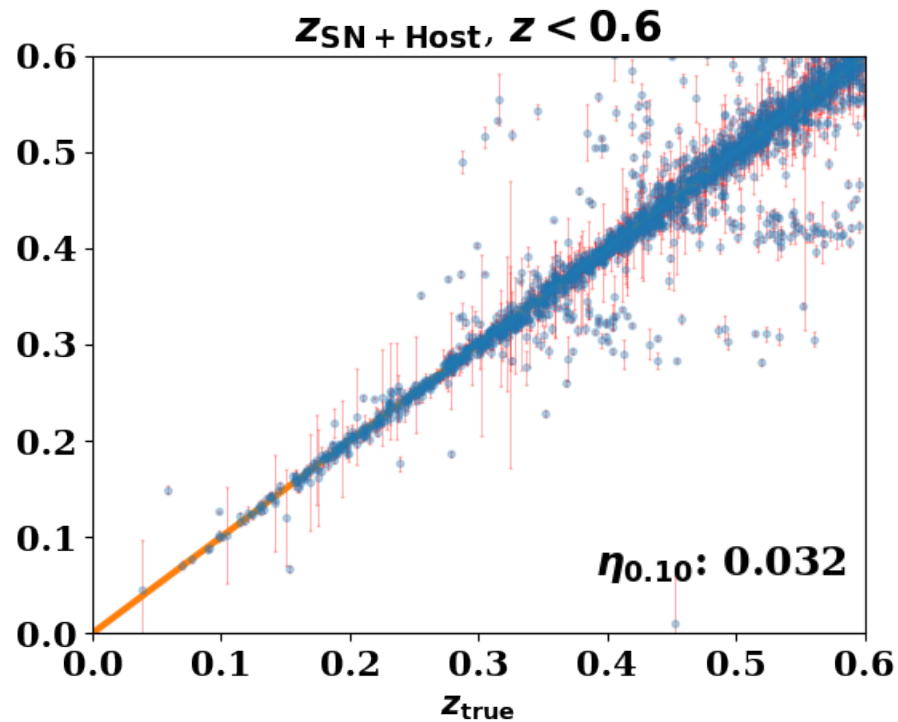
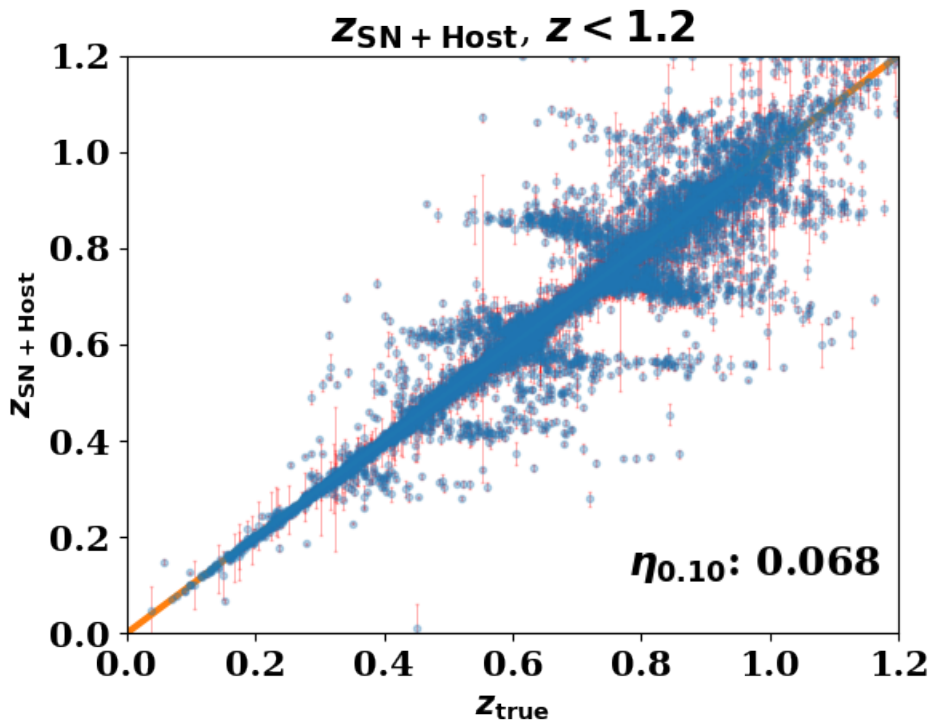


- Better performance at high- $z$

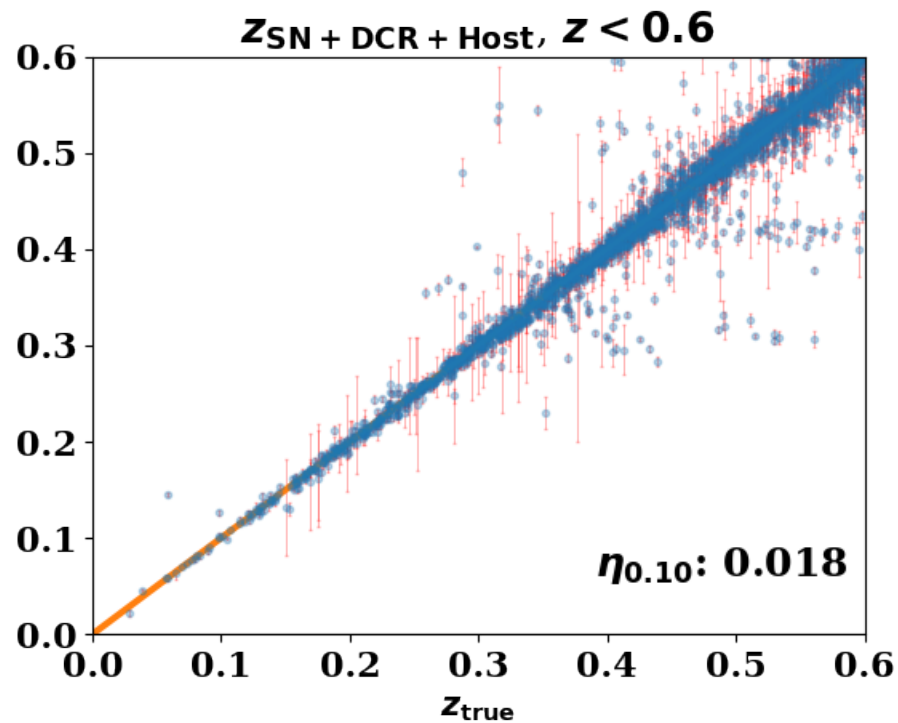
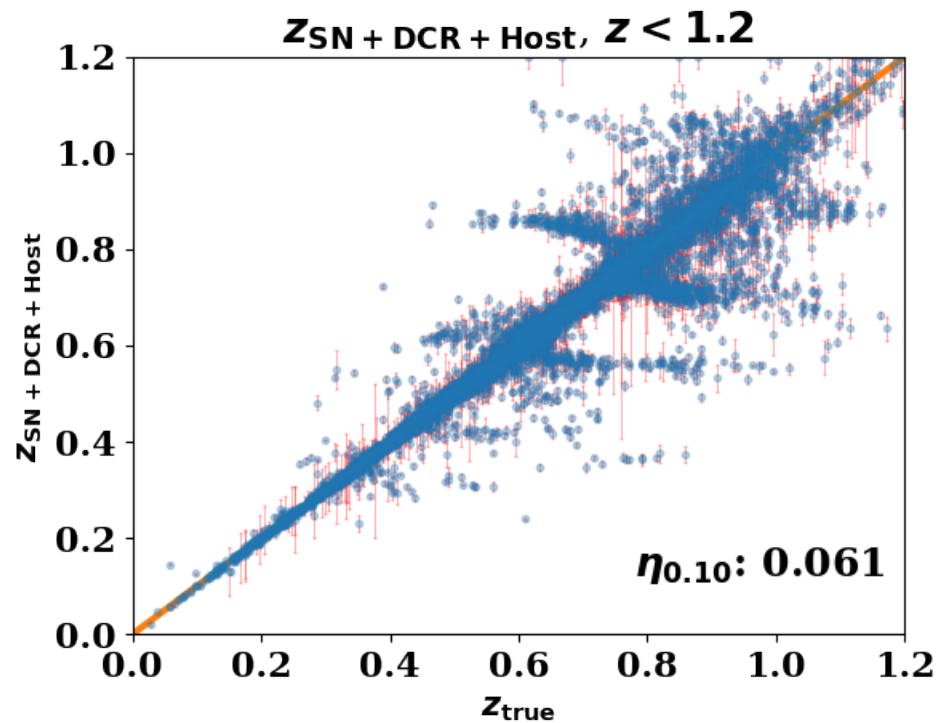
# SN photo- $z$ ( $z_{\text{SN}}$ )



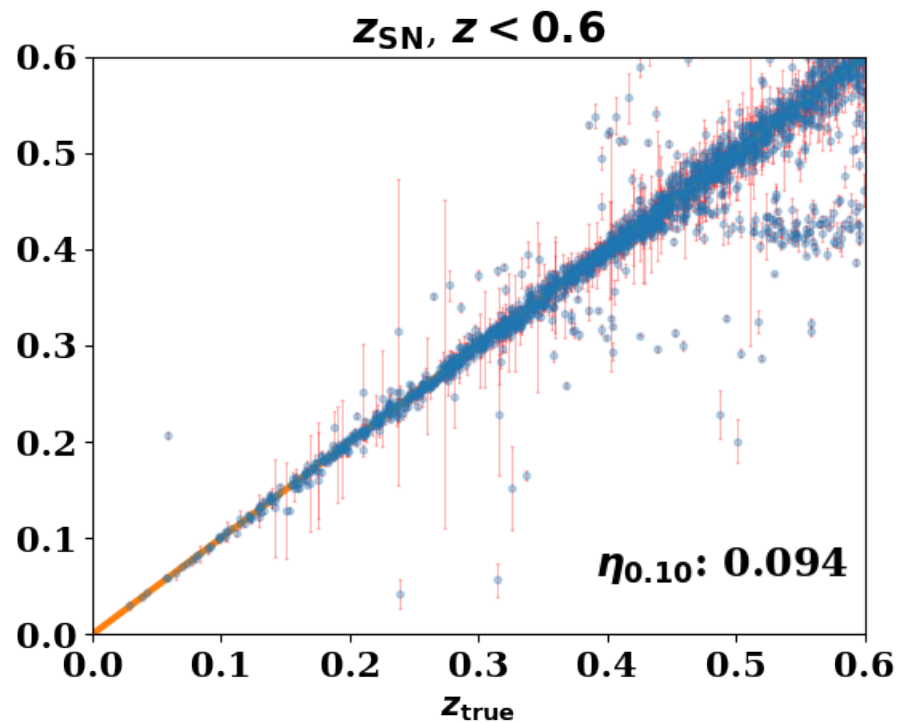
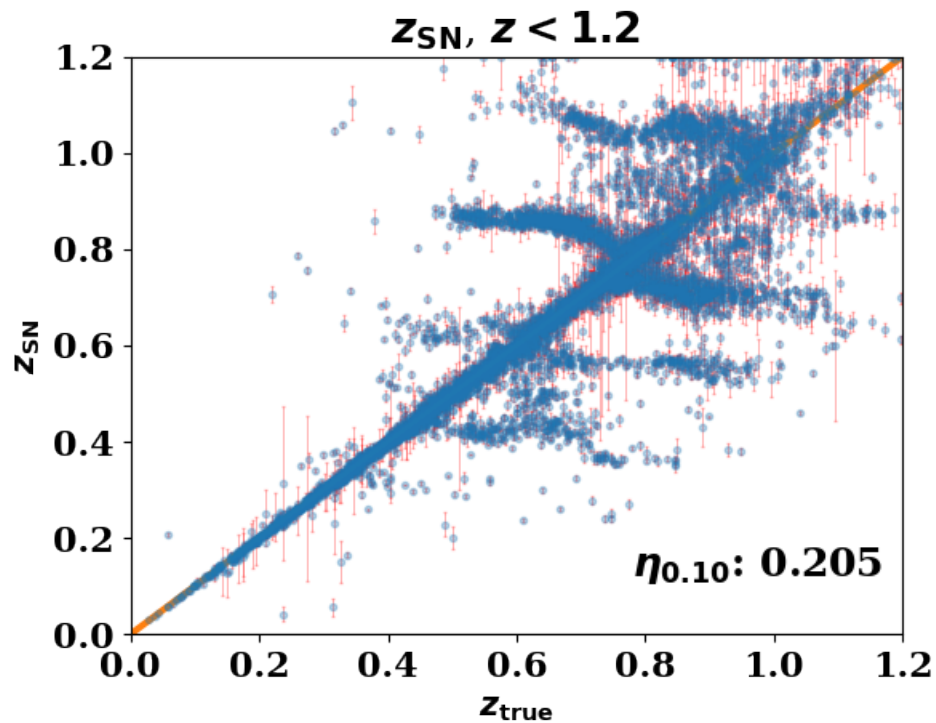
# SN photo- $z$ + Host photo- $z$ ( $z_{\text{SN+Host}}$ )



# SN photo- $z$ + Host photo- $z$ + Astro- $z$ ( $z_{\text{SN+DCR+Host}}$ )

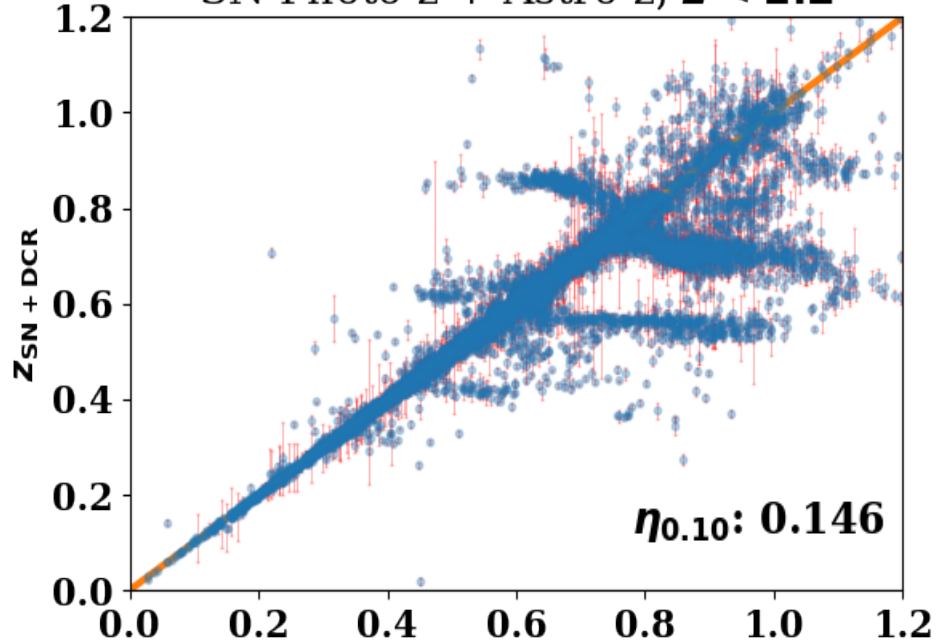


# SN photo- $z$ ( $z_{\text{SN}}$ )

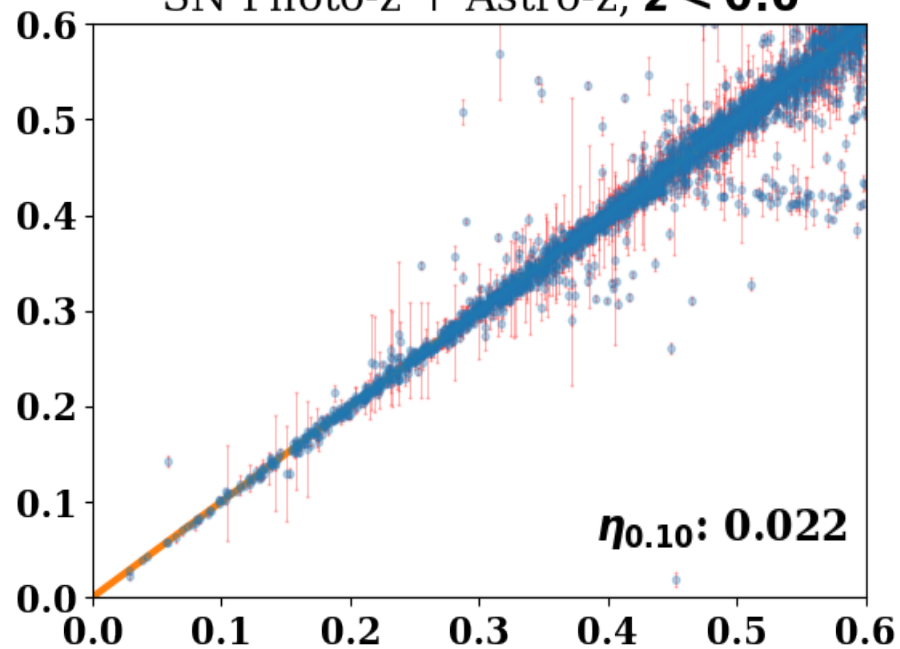


# SN photo- $z$ + Astro- $z$ ( $z_{\text{SN+DCR}}$ )

SN Photo- $z$  + Astro- $z$ ,  $z < 1.2$

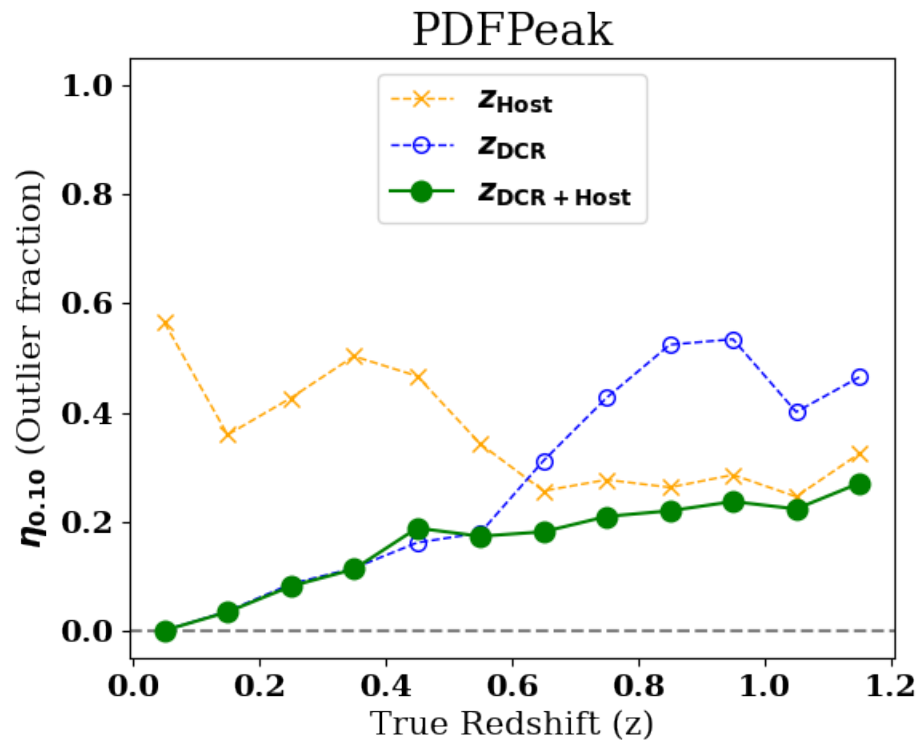
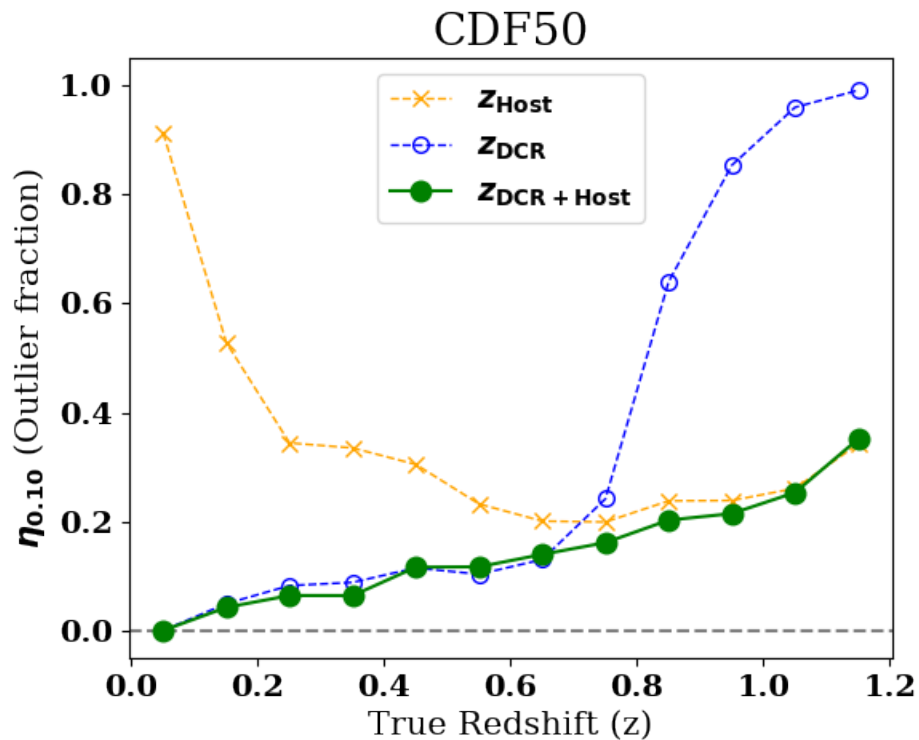


SN Photo- $z$  + Astro- $z$ ,  $z < 0.6$

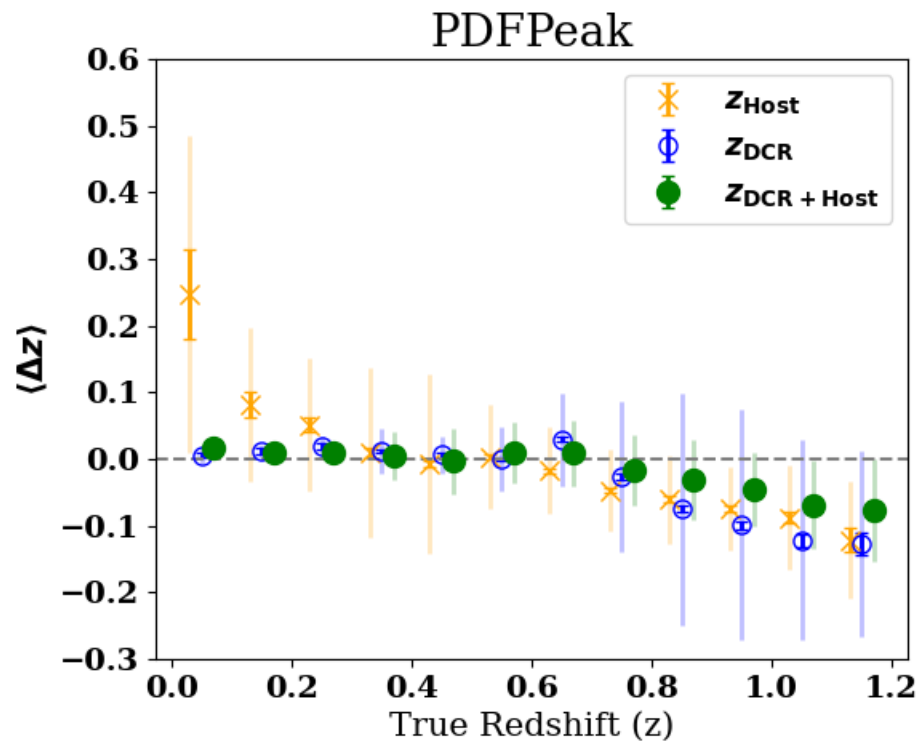
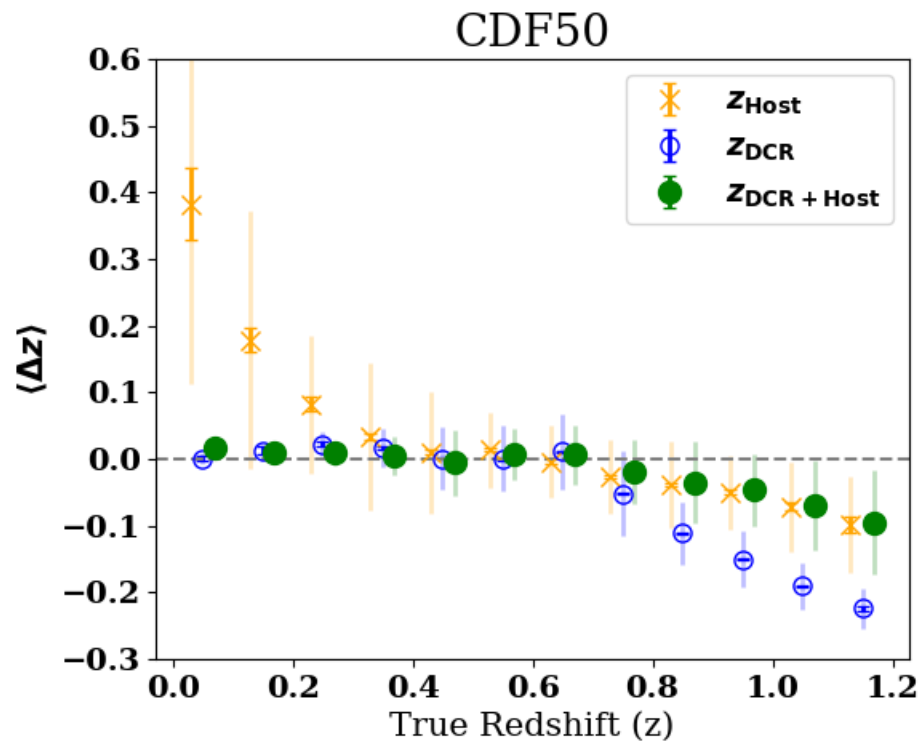


● Pearson Correlation: 0.154

# Outlier Fractions by True Redshift



# Bias and MAD Deviation by True Redshift

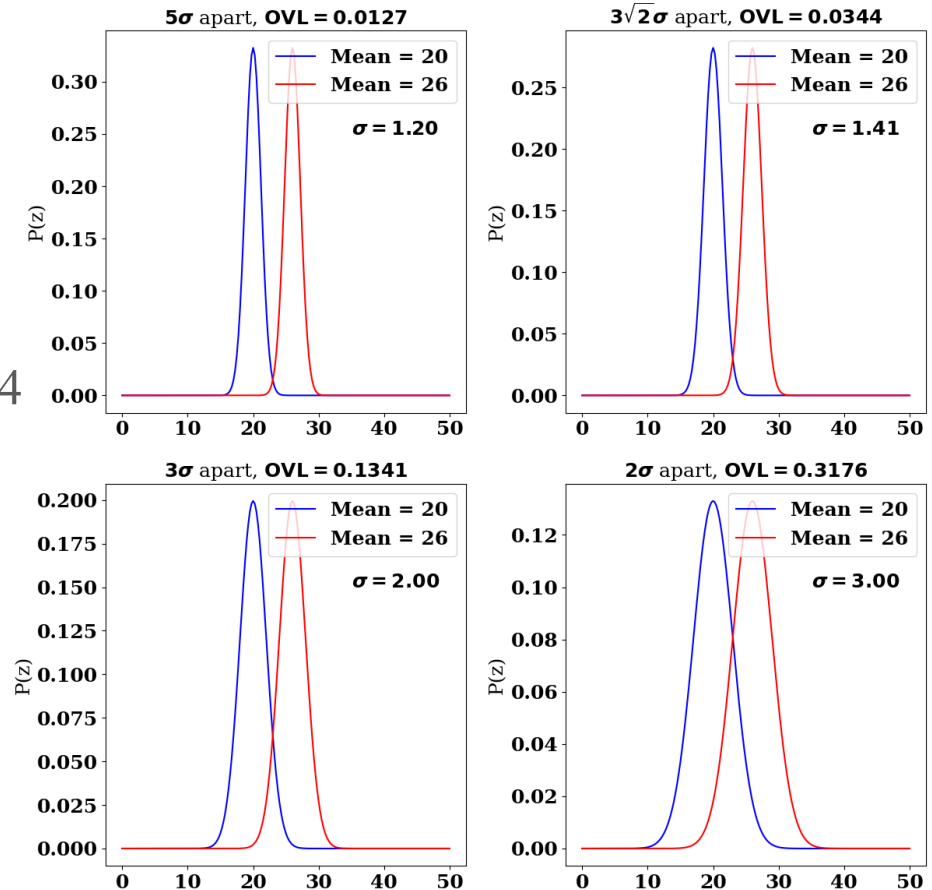




# Treatment of Incompatible Priors

$$\text{OVL}(p_0, p_1) = \int \min(p_0(x), p_1(x)) dx$$

➤ Only combine priors when  $\text{OVL} > 0.0344$



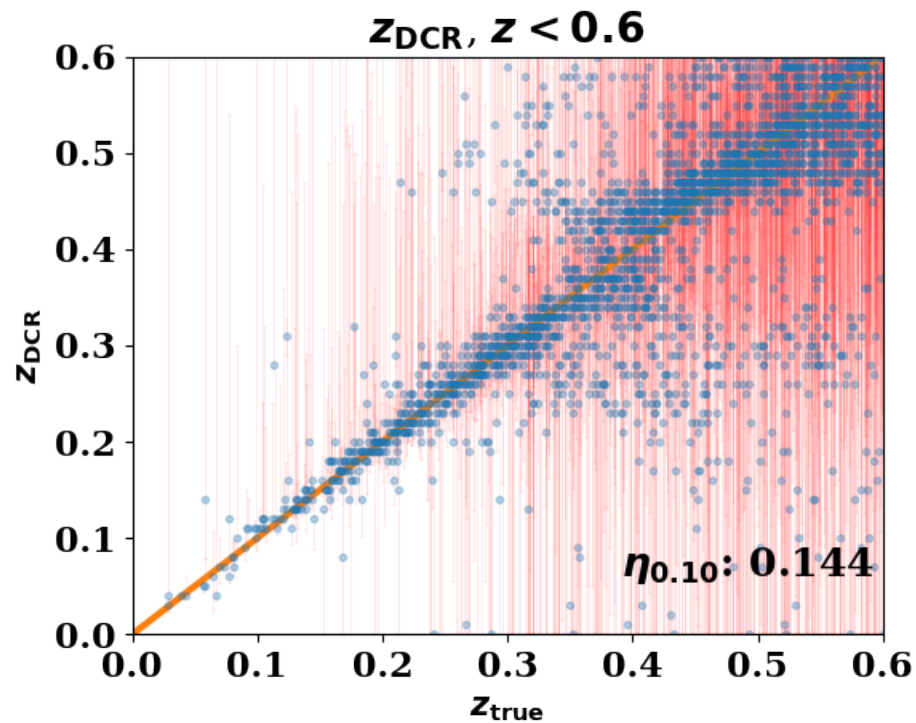
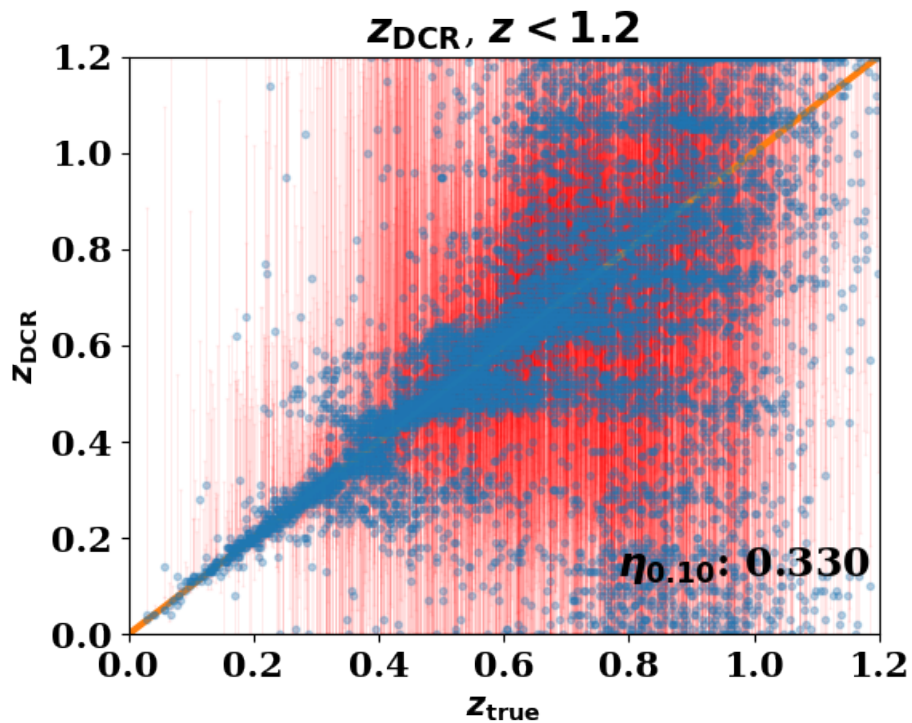
# Various Metrics

- low- $z$ :  $z_{\text{true}} < 0.6$ , high- $z$ :  $z_{\text{true}} > 0.6$

Type	Bias (low- $z$ )	Bias (high- $z$ )	$\eta_{0.10}$ (low- $z$ )	$\eta_{0.10}$ (high- $z$ )	$\sigma_{\text{MAD}}$ (low- $z$ )	$\sigma_{\text{MAD}}$ (high- $z$ )
$z_{\text{DCR}}$	$0.007 \pm 0.002$	$-0.046 \pm 0.002$	0.144	0.438	0.031	0.121
$z_{\text{Host}}$	$0.012 \pm 0.003$	$-0.052 \pm 0.002$	0.421	0.268	0.123	0.066
$z_{\text{DCR+Host}}$	$0.005 \pm 0.001$	$-0.024 \pm 0.002$	0.151	0.212	0.039	0.057
$z_{\text{SN}}$	$0.014 \pm 0.001$	$0.002 \pm 0.001$	0.094	0.265	0.007	0.064
$z_{\text{SN+Host}}$	$0.005 \pm 0.001$	$-0.004 \pm 0.001$	0.032	0.089	0.008	0.024
$z_{\text{SN+DCR}}$	$0.003 \pm 0.001$	$-0.036 \pm 0.001$	0.022	0.224	0.007	0.044
$z_{\text{SN+DCR+Host}}$	$0.003 \pm 0.001$	$-0.008 \pm 0.001$	0.019	0.087	0.007	0.023

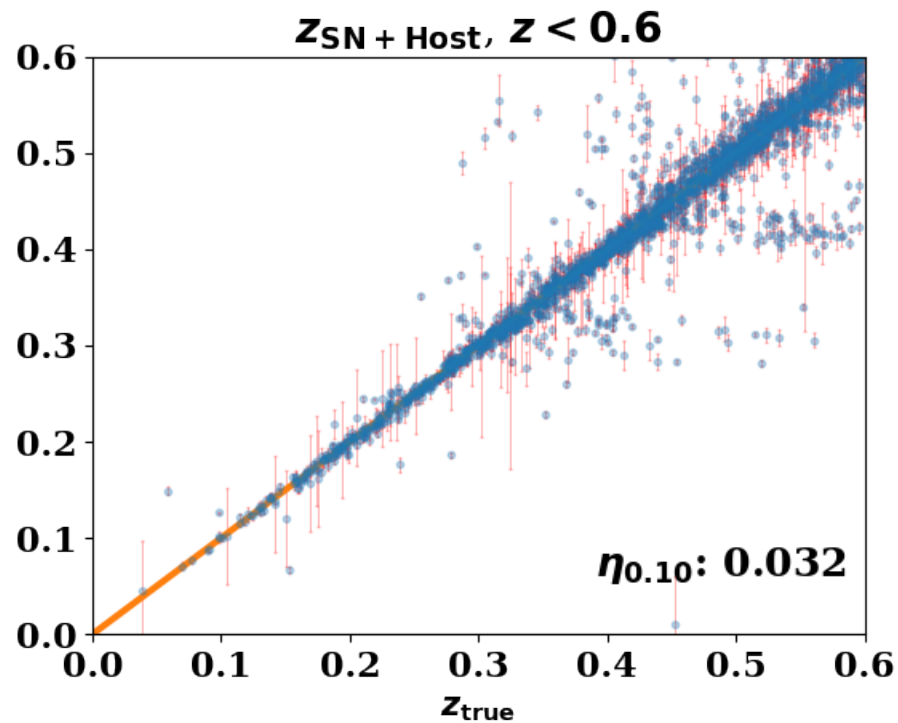
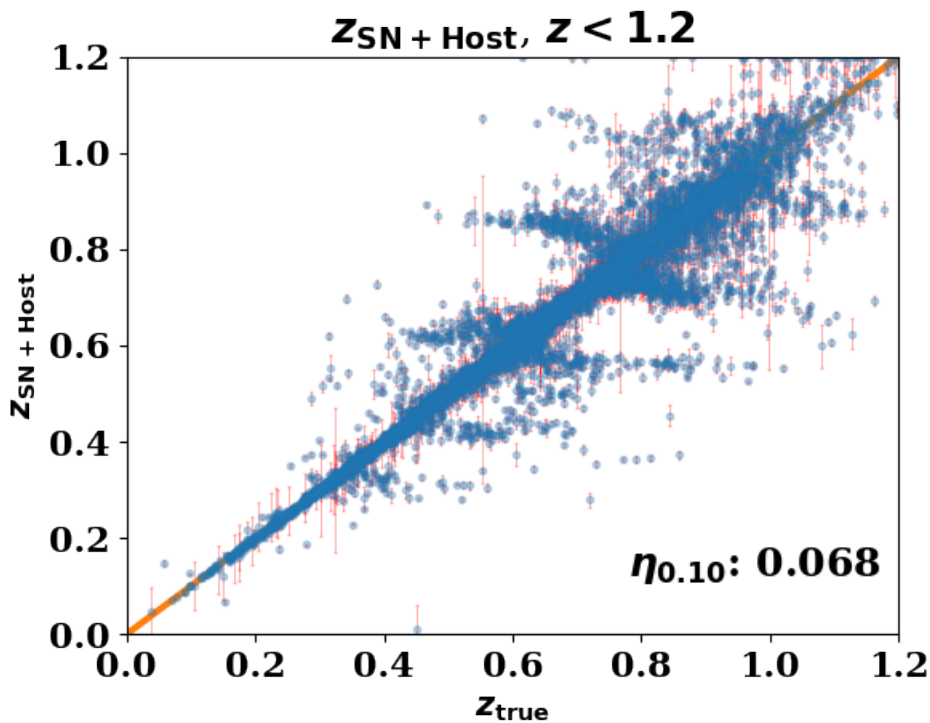
**Table 8.** Bias and error on the mean,  $\eta_{0.10}$ , and  $\sigma_{\text{MAD}}$  for low- $z$  ( $z_{\text{true}} < 0.6$ ) and high- $z$  ( $z_{\text{true}} > 0.6$ ) for various combinations of the three methods discussed in this work. For  $z_{\text{DCR}}$ ,  $z_{\text{Host}}$ , and  $z_{\text{DCR+Host}}$ , we show the PDFPeak values.  $z_{\text{DCR}}$  shows good performance at low- $z$  while  $z_{\text{Host}}$  shows good performance at high- $z$ , which leads to  $z_{\text{SN+DCR}}$  being significantly improved at low- $z$  compared to  $z_{\text{SN}}$  and  $z_{\text{SN+Host}}$  being significantly improved at high- $z$ .  $z_{\text{SN+DCR+Host}}$  show similar performance to  $z_{\text{SN+DCR}}$  and  $z_{\text{SN+Host}}$  at low- $z$  and high- $z$  respectively.

# Astro-z ( $z_{\text{DCR}}$ )



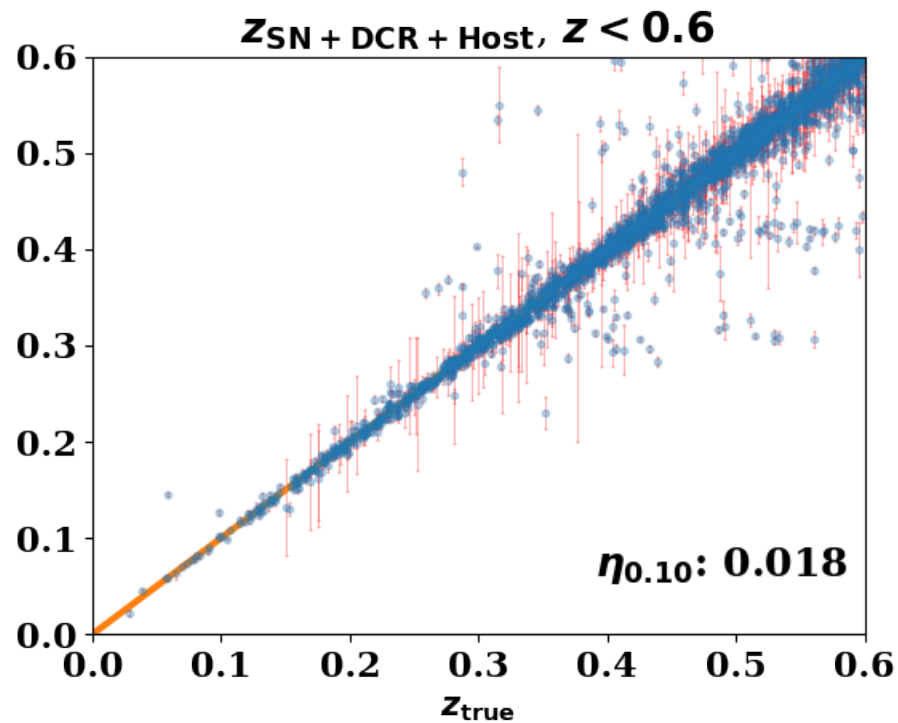
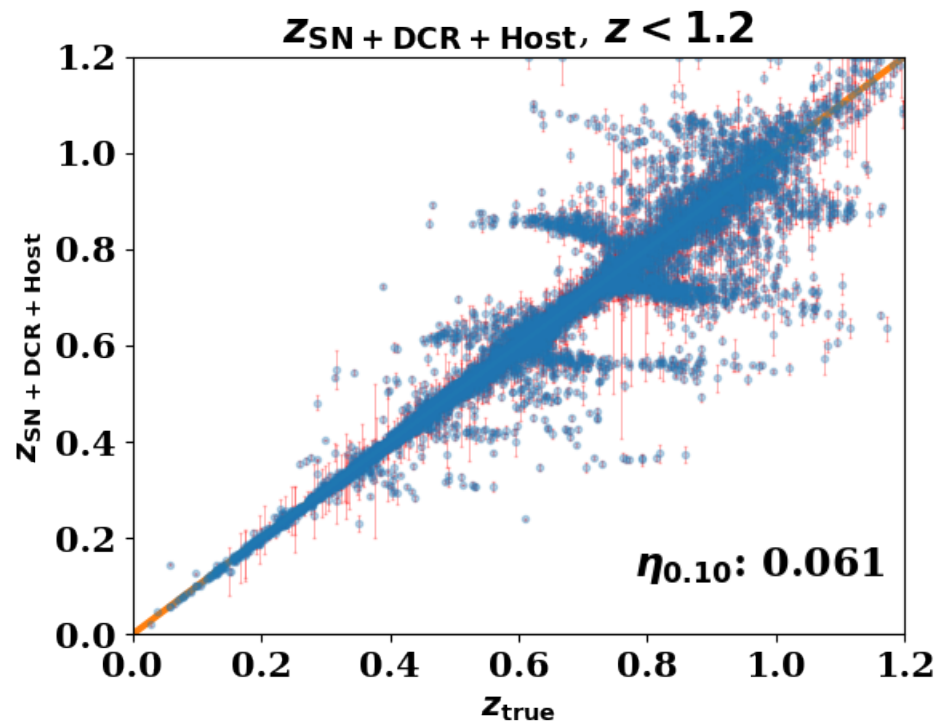
- $\Delta z \equiv \frac{z_{\text{estimate}} - z_{\text{true}}}{1 + z_{\text{true}}}$  &  $\eta_x$ : fraction of candidates where  $|\Delta z| > x$

# SN photo- $z$ + Host photo- $z$ ( $z_{\text{SN+Host}}$ )



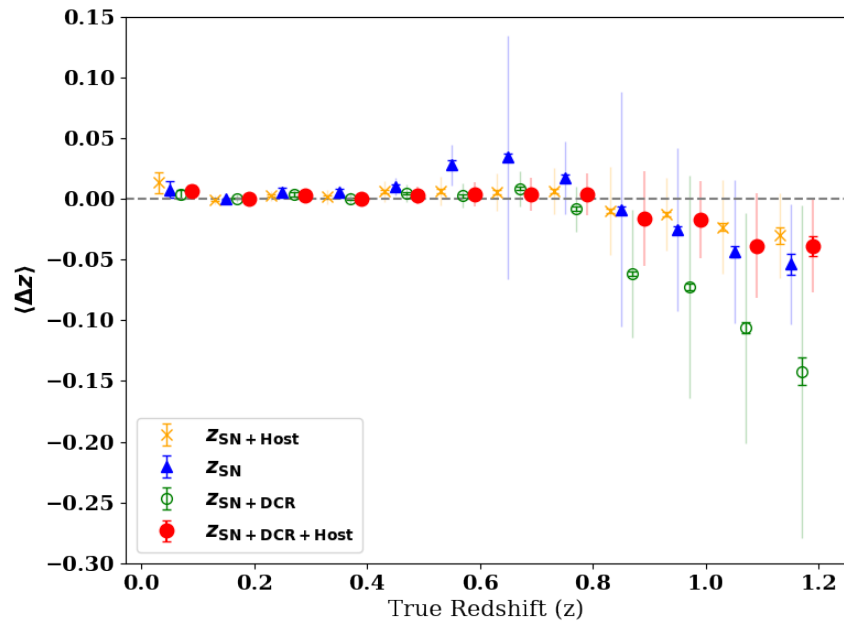
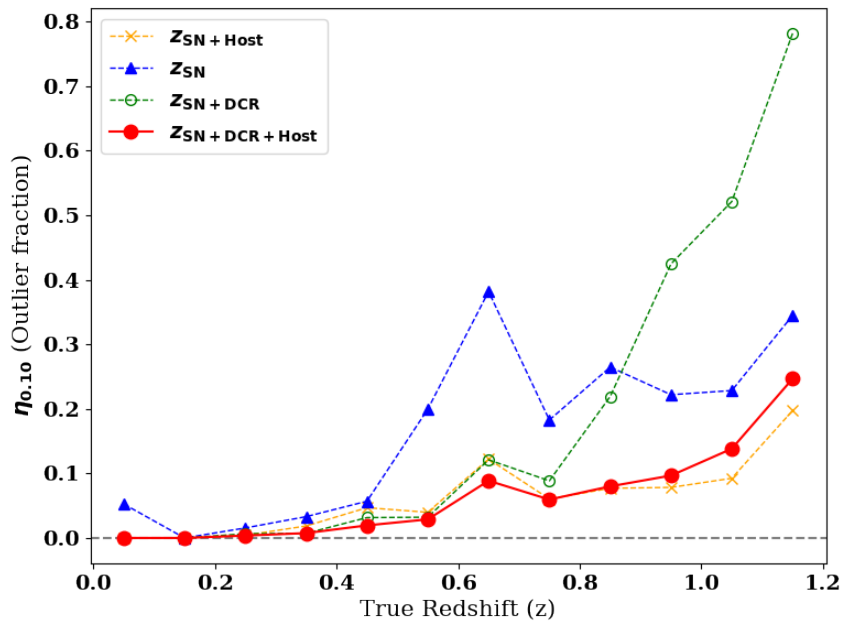
- Using LCFIT+z,  $z_{\text{Host}}$  from RAIL

# SN photo- $z$ + Host photo- $z$ + Astro- $z$ ( $z_{\text{SN+DCR+Host}}$ )



- 8 ~ 9% more LCFIT+z converged candidates that pass selection cut

# Outlier Fractions, Bias and MAD Deviation by True Redshift



●  $\Delta z \equiv \frac{z_{\text{estimate}} - z_{\text{true}}}{1 + z_{\text{true}}}$  &  $\eta_x$ : fraction of candidates where  $|\Delta z| > x$