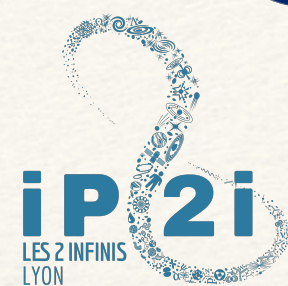




IN2P3

Institut national de physique nucléaire
et de physique des particules



pip install **skysurvey**

Fast python package to simulate *targets* observed by *surveys*

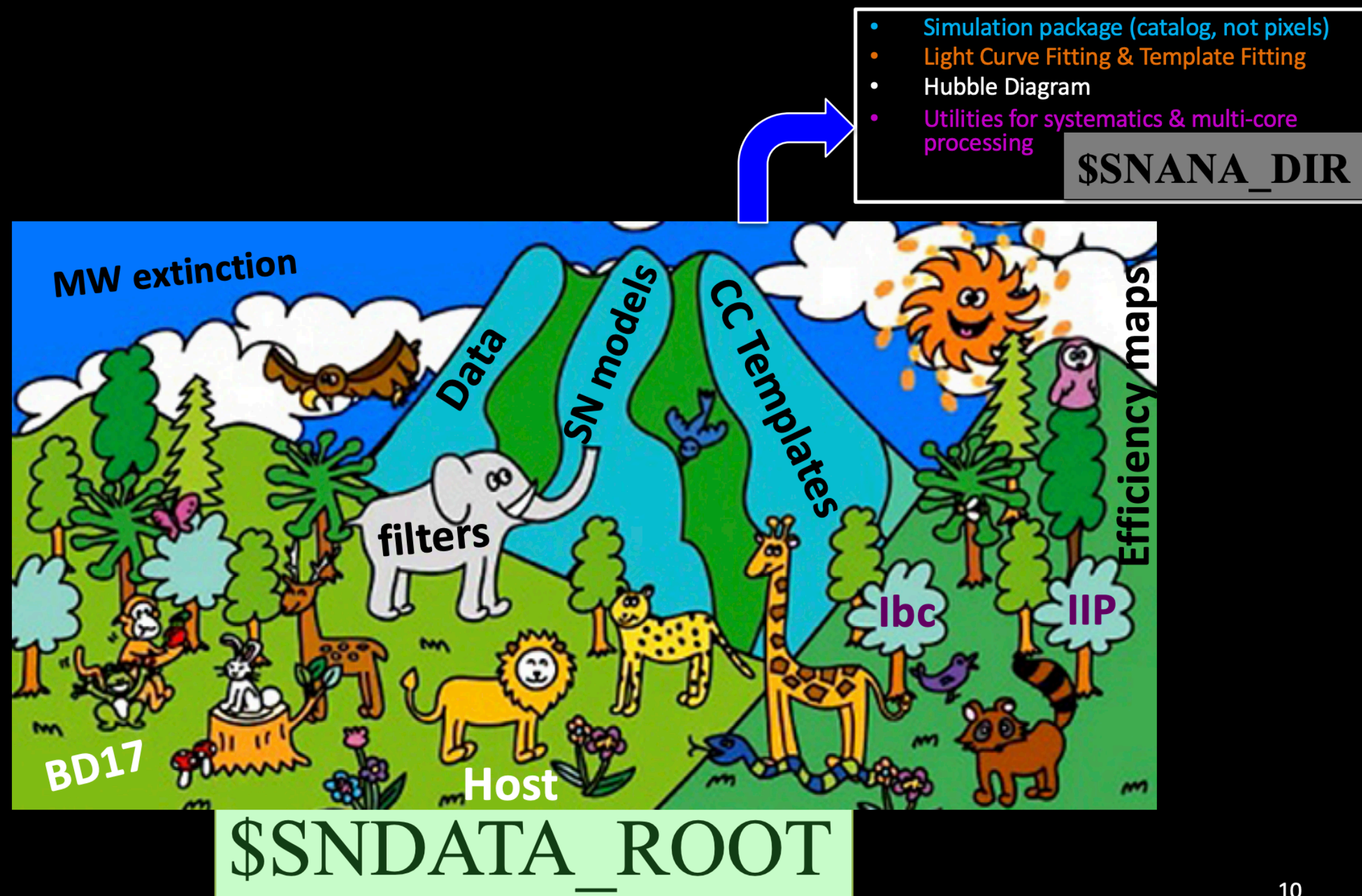
& see also pip install **modeldag**

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Simulation: SNANA

Source: SNANA tutorial

Architecture: Environment



10

Became the central tools SN cosmo:

Simulate Transient

Simulate effect of selection function

Correct for bias selection

Fit cosmology

Scolnic+18 ; Brout+22 ; Riess+22 ; Popovic 2024 ; Vincenzi+24

Preparing:

LSST, Roman

Plasctic / Elasticc

Simulation: SNANA

Became the central tools SN cosmo:

Simulate Transient

Simulate effect of selection function

Correct for bias selection

Fit cosmology

Scolnic+18 ; Brout+22 ; Riess+22 ; Vincenzi+24

SNANA is central for cosmological analyses...

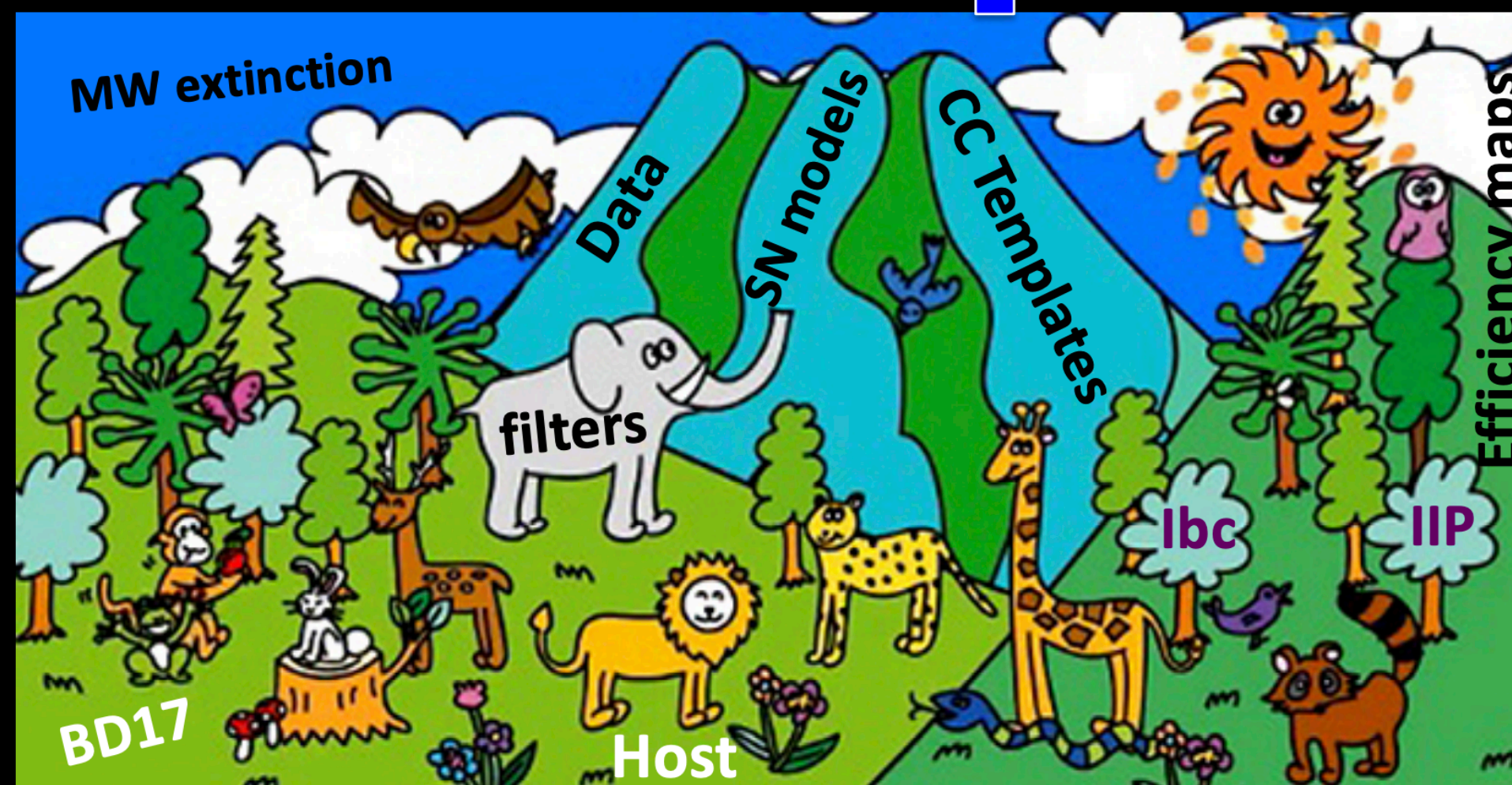
Yet, hard to install and complex to use.

Source: SNANA tutorial

Architecture: Environment

- Simulation package (catalog, not pixels)
- Light Curve Fitting & Template Fitting
- Hubble Diagram
- Utilities for systematics & multi-core processing

`$$SNANA_DIR`



`$$SNANA_ROOT`

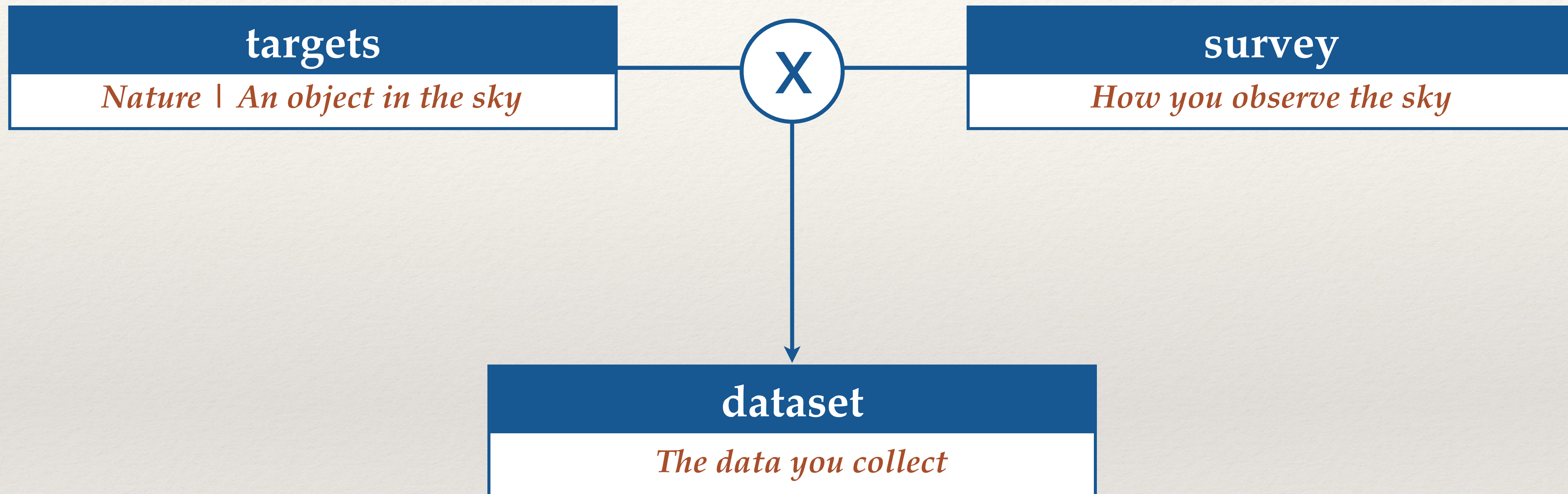
10

`pip install skysurvey`

skysurvey | concept

aims to replace SNANA “step 1: simulation”

See also [simsurvey](#) | [snsim](#)



skysurvey | *goal: easy to use*

targets

Nature | An object in the sky

```
import skysurvey
snia = skysurvey.SNeIa.from_draw(10_000, zmax=0.3,
                                tstart = "2018-04-01",
                                tstop = "2020-12-01")
```

Lots of pre-built targets, *easy* to customize

X

survey

How you observe the sky

```
ztf = skysurvey.ZTF.from_logs() # login requires
lsst = skysurvey.LSST.from_opsim('baseline_v3.3_10yrs.db')
```

Easy to build you own survey

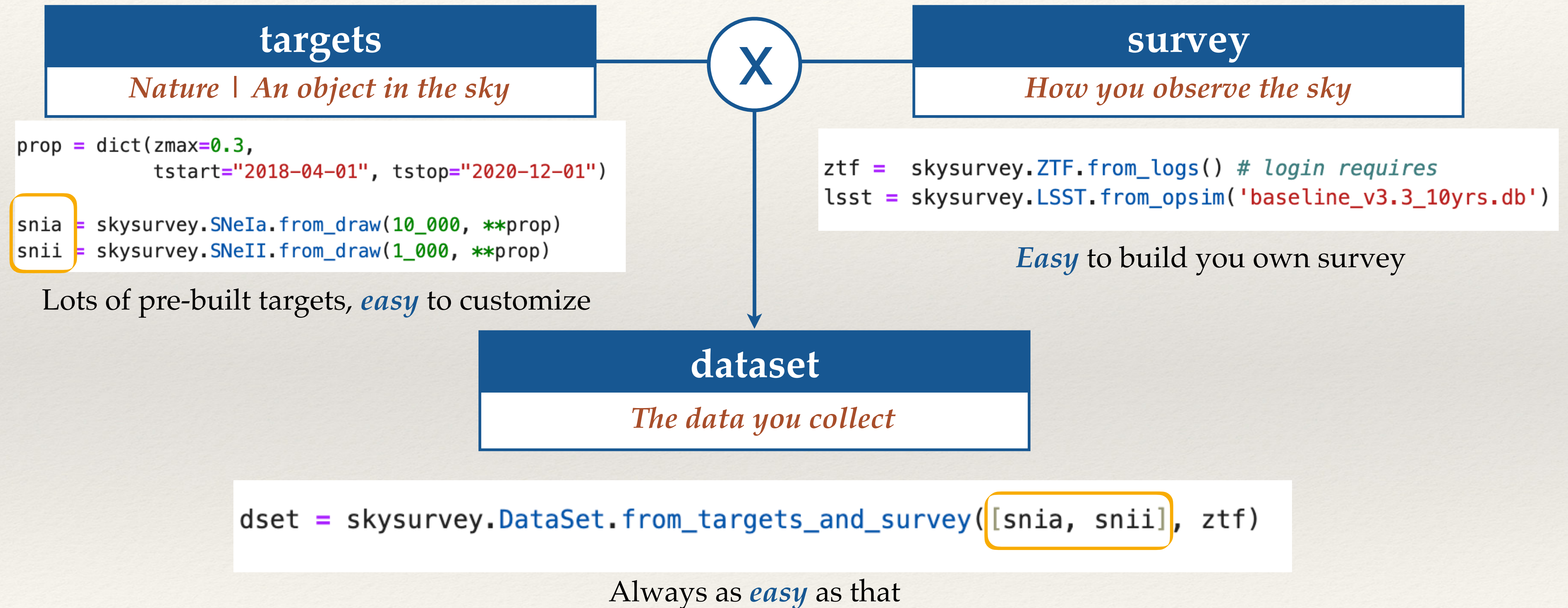
dataset

The data you collect

```
dset = skysurvey.DataSet.from_targets_and_survey(snia, ztf)
```

Always as *easy* as that

skysurvey | *goal: easy to use*



skysurvey | *based within the python environment*



targets
Nature | An object in the sky

survey
How you observe the sky



snia.data

	z	x1	c	t0	ra	dec	magabs	mwebv	magobs	x0	template
0	0.07705	-0.030	1.221450	58487.214844	336.697937	7.605778	-15.358439	0.134286	22.429092	0.000017	salt2
1	0.06865	-0.700	-0.027583	58659.839844	60.706551	55.346775	-19.233326	0.719195	18.290983	0.000766	salt2
2	0.06905	0.320	0.351301	58782.136719	23.629887	-0.468616	-18.197527	0.036892	19.339998	0.000292	salt2

survey.data

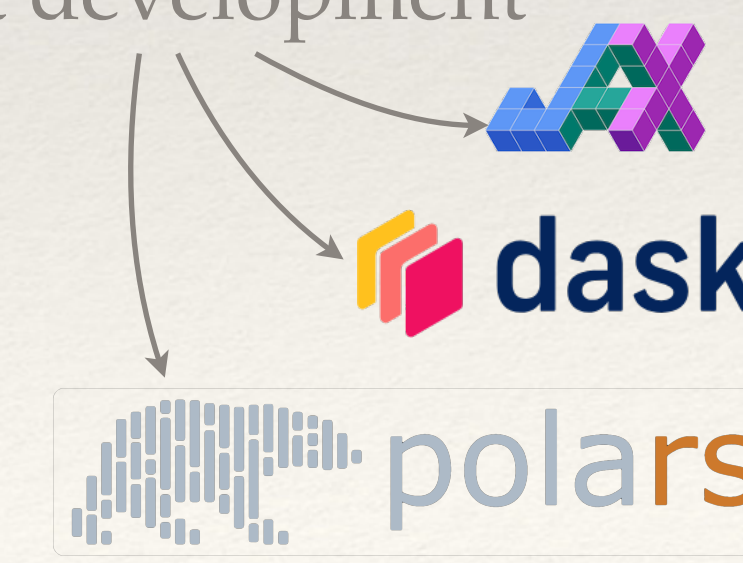
	gain	zp	skynoise	mjd	band	ra	dec	fieldid_survey	fieldid
1867451	1	30	234.6674346923828	58800.0078125	desr	18.292482376098633	1.2328510284423828	11331	227158
1867554	1	30	234.6674346923828	58800.0078125	desr	18.292482376098633	1.2328510284423828	11331	227159
1867634	1	30	234.6674346923828	58800.0078125	desr	18.292482376098633	1.2328510284423828	11331	227160

dataset
The data you collect

dset.data

targets .index	survey .index	fieldid	rcid	time	band	flux	fluxerr	zp	zpsys
5821	26009040	1	3	59104.390625	ztrf	-39.302373	36.143482	25.233383	ab
	26054356	1	3	59106.328125	ztfgr	8.518026	14.717361	24.559206	ab
	26331019	1	3	59112.300781	ztfgr	24.598649	20.259558	26.025703	ab

In test development



skysurvey | DataFrame

targets

```
import skysurvey
snia = skysurvey.SNeIa.from_draw(10_000, zmax=0.3,
                                tstart = "2018-04-01",
                                tstop = "2020-12-01")
```

survey

```
ztf = skysurvey.ZTF.from_logs()
```

47 M entries | 760 k exposures

dataset

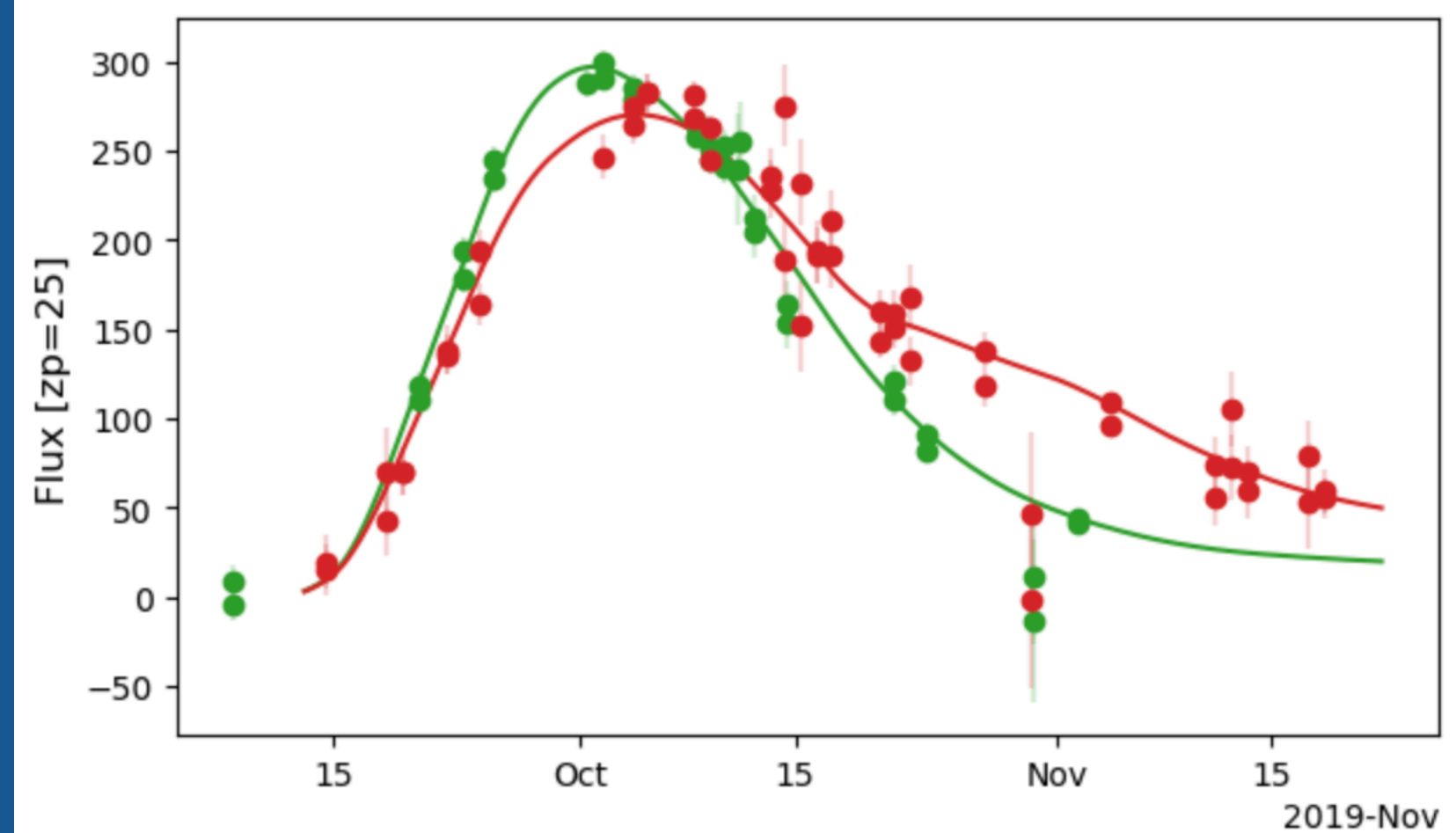
```
dset = skysurvey.DataSet.from_targets_and_survey(snia, ztf)
```

dset.data

	survey.index	fieldid	rcid	time	band	flux	fluxerr	zp	zpsys
5821	26009040	1	3	59104.390625	ztfr	-39.302373	36.143482	25.233383	ab
	26054356	1	3	59106.328125	ztfg	8.518026	14.717361	24.559206	ab
	26331019	1	3	59112.300781	ztfg	24.598649	20.259558	26.025703	ab
	26338533	1	3	59112.359375	ztfr	-3.302169	36.260182	26.127666	ab
	27121580	1	3	59128.250000	ztfr	-20.931607	46.604773	25.959711	ab
...
9727	30211656	1408	16	59203.179688	ztfr	14.887892	60.705860	26.204699	ab
	30211715	1408	16	59203.179688	ztfr	90.188222	61.295684	26.208666	ab
	30841192	1408	16	59221.167969	ztfr	34.804501	22.752532	26.142845	ab
	36571545	1408	16	59392.414062	ztfr	-9.636968	41.250546	26.149637	ab
	36571604	1408	16	59392.414062	ztfr	-18.475494	44.445759	26.151228	ab

8429351 rows x 8 columns

```
fig = dset.show_target_lightcurve(index=347, phase_window=[-40,100])
```



skysurvey | *easy & fast*

targets 0.5s

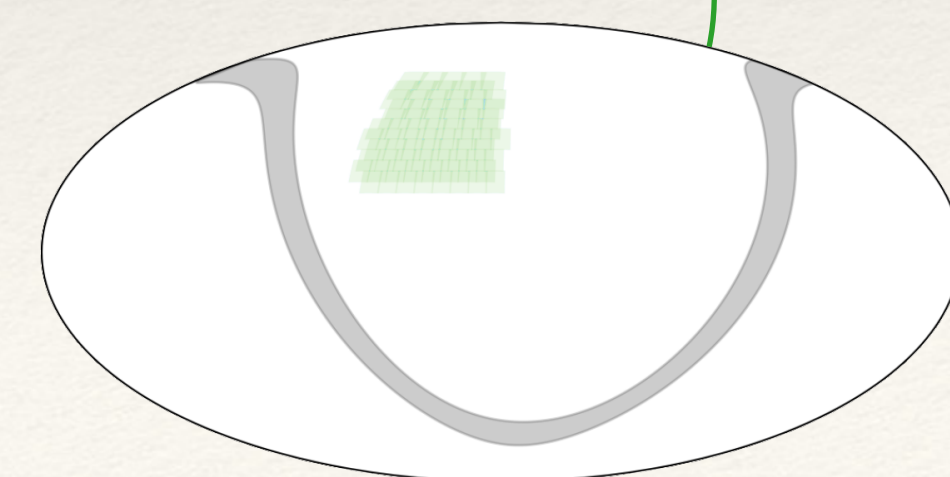
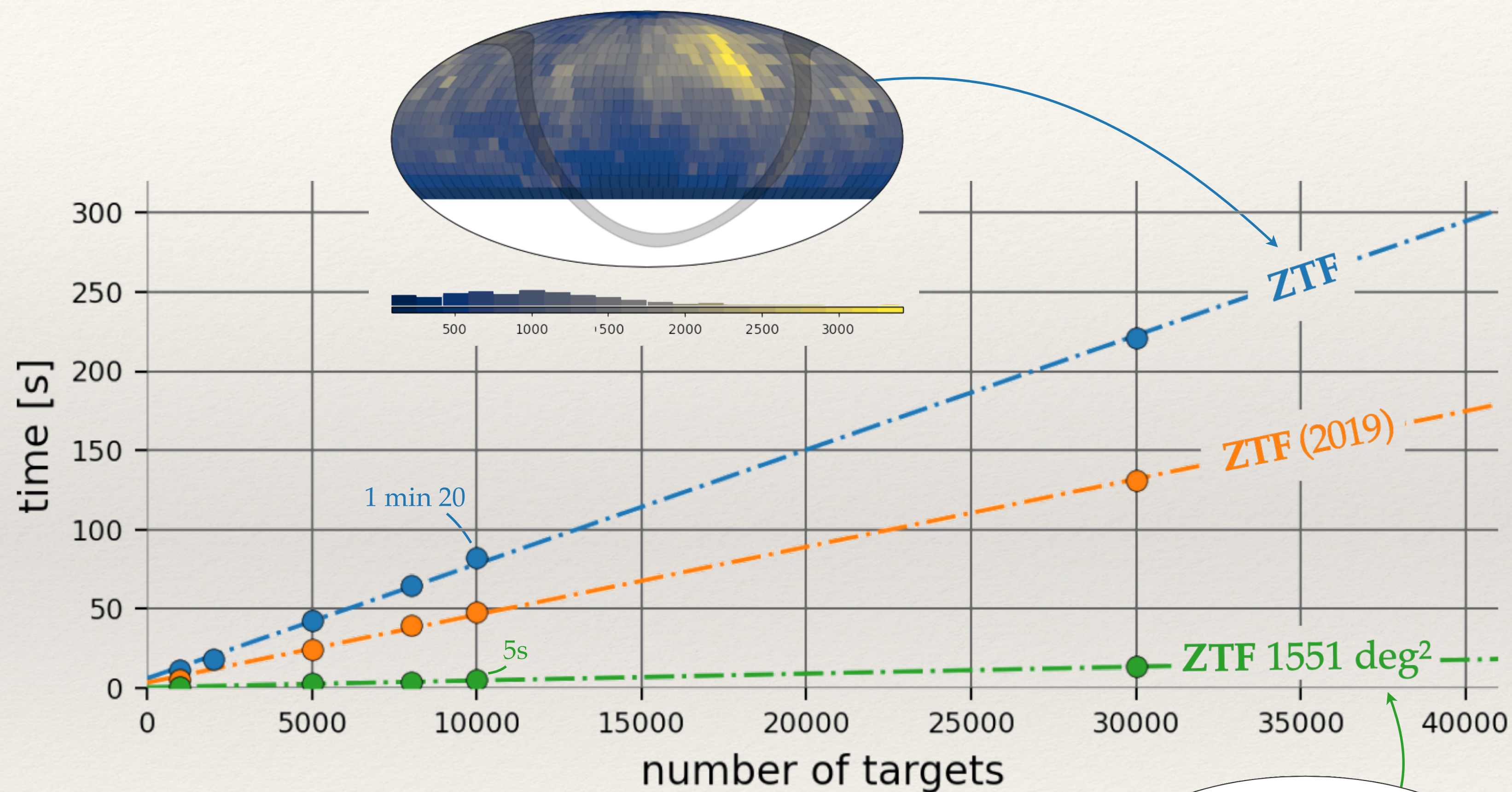
```
import skysurvey
snia = skysurvey.SNeIa.from_draw(10_000, zmax=0.3,
                                tstart = "2018-04-01",
                                tstop = "2020-12-01")
```

survey 1s

```
ztf = skysurvey.ZTF.from_logs()
47 M entries | 760 k exposures
```

dataset 80s

```
dset = skysurvey.DataSet.from_targets_and_survey(snia, ztf)
```



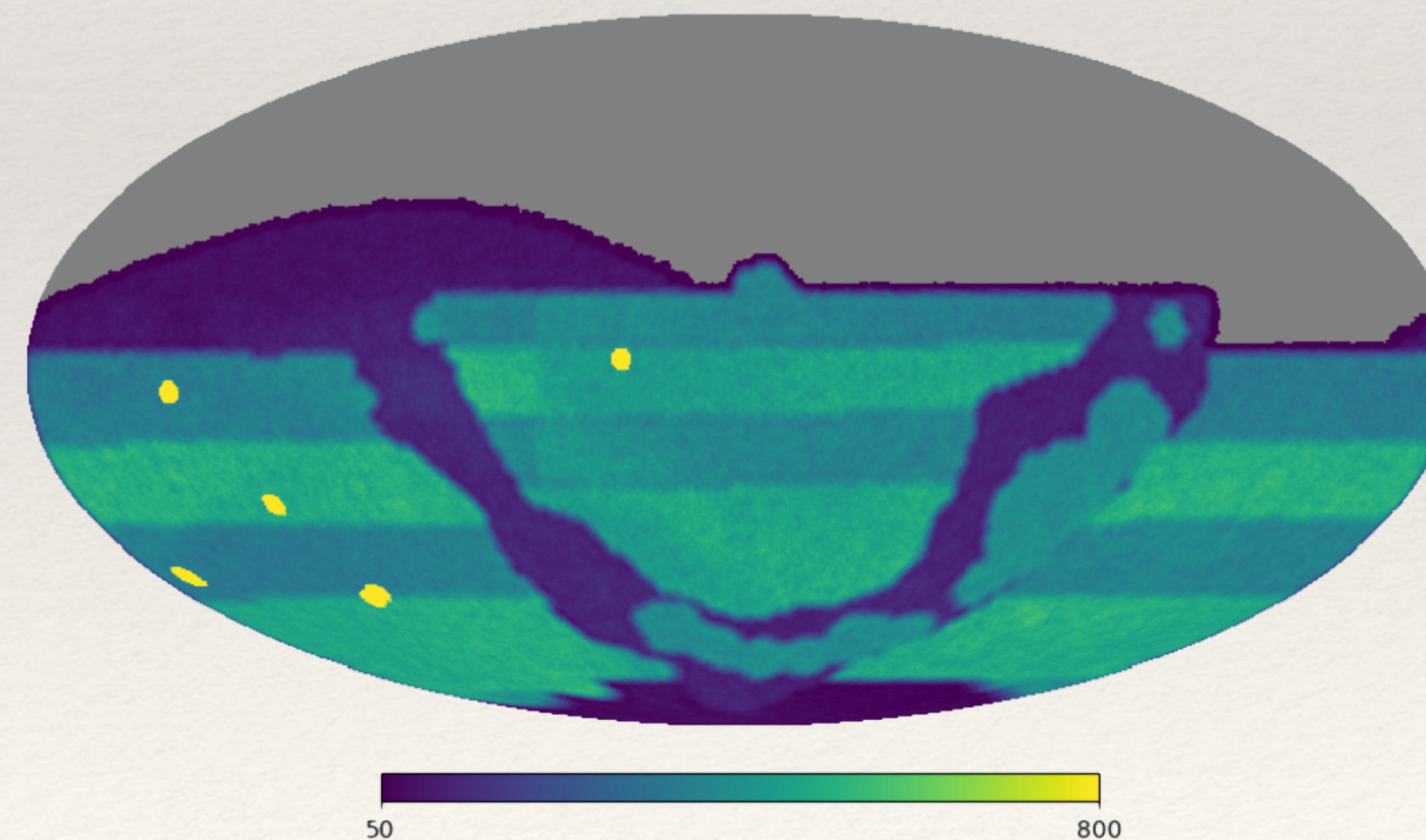
```
survey 1min20  
  
import skysurvey  
lsst = skysurvey.LSST.from_opsim('baseline_v3.3_10yrs.db',  
                                sql_where="night<365*5")
```

119 962 089 lines DataFrame

```
targets 1s  
  
tstart, tstop = lsst.get_timerange()  
snia = skysurvey.SNeIa.from_draw(100_000,  
                                 template="salt2-extended",  
                                 zmax=0.7,  
                                 tstart=tstart, tstop=tstop)
```

```
dataset 3min20  
  
dset = skysurvey.DataSet.from_targets_and_survey(snia, lsst)
```

100 000 SNe Ia | 5 years of LSST | 3min



```

survey 1min20
import skysurvey
lsst = skysurvey.LSST.from_opsim('baseline_v3.3_10yrs.db',
                                sql_where="night<365*5")
    
```

119 962 089 lines DataFrame

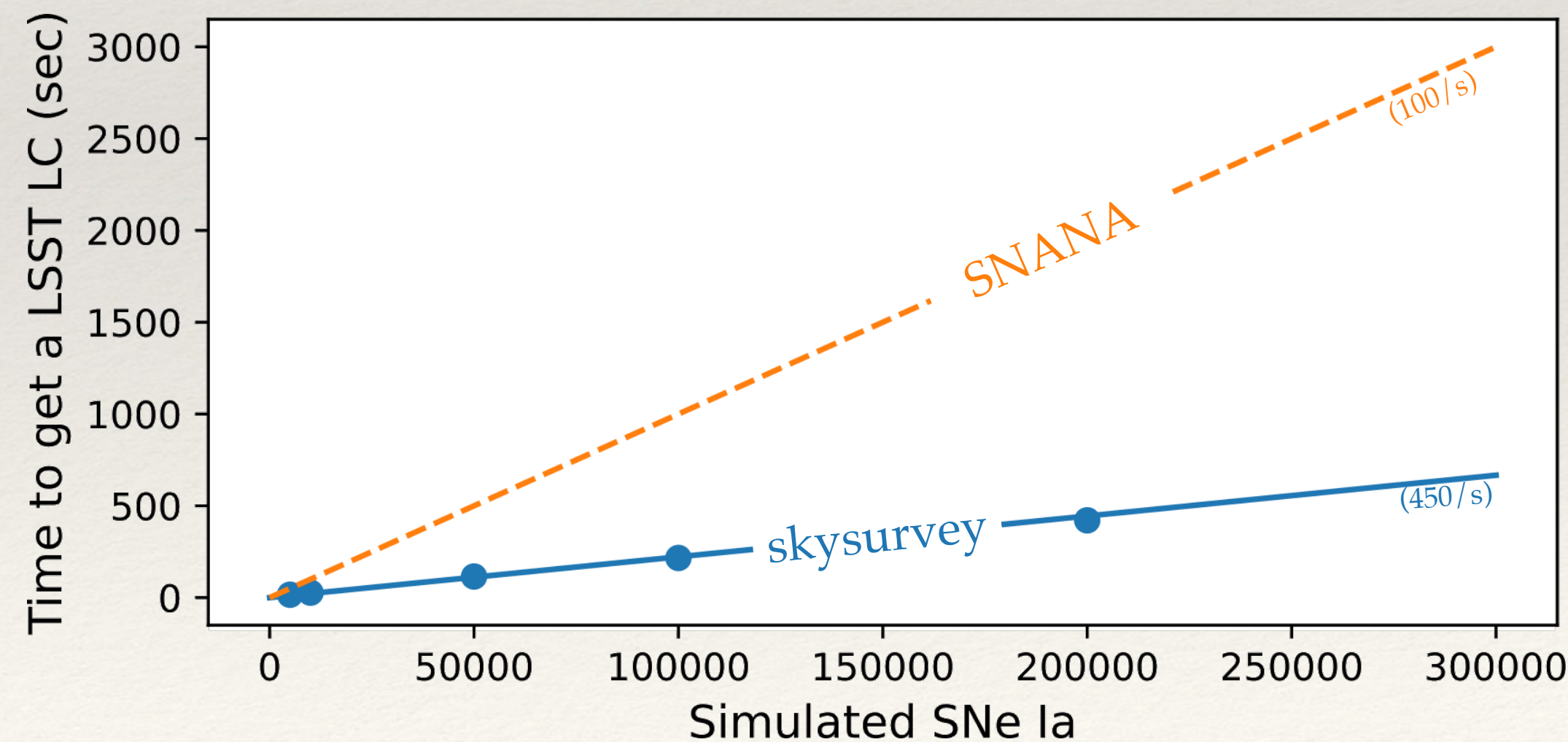
```

targets 1s
tstart, tstop = lsst.get_timerange()
snia = skysurvey.SNeIa.from_draw(100_000,
                                template="salt2-extended",
                                zmax=0.7,
                                tstart=tstart, tstop=tstop)
    
```

```

dataset 3min20
dset = skysurvey.DataSet.from_targets_and_survey(snia, lsst)
    
```


100 000 SNe Ia | 5 years of LSST | 3min



Source: SNANA tutorial

SIM CPU Proc-Time

U Chicago Research Computing Center:
Sep 2018 for PLAsTiCC



- 117 million light curves generated in 8 hr on 40 cores → 100/sec

skysurvey: 450/sec | 1 laptop (M1)

pip install skysurvey

```
survey 1min20  
import skysurvey  
lsst = skysurvey.LSST.from_opsim('baseline_v3.3_10yrs.db',  
                                sql_where="night<365*5")
```

Customize these as you want!

```
targets 1s  
tstart, tstop = lsst.get_timerange()  
snia = skysurvey.SNeIa.from_draw(100_000,  
                                 template="salt2-extended",  
                                 zmax=0.7,  
                                 tstart=tstart, tstop=tstop)
```

This will work

```
dataset 3min20  
dset = skysurvey.DataSet.from_targets_and_survey(snia, lsst)
```

100 000 SNe Ia | 5 years of LSST | 3min

Easy to build *complex* transient

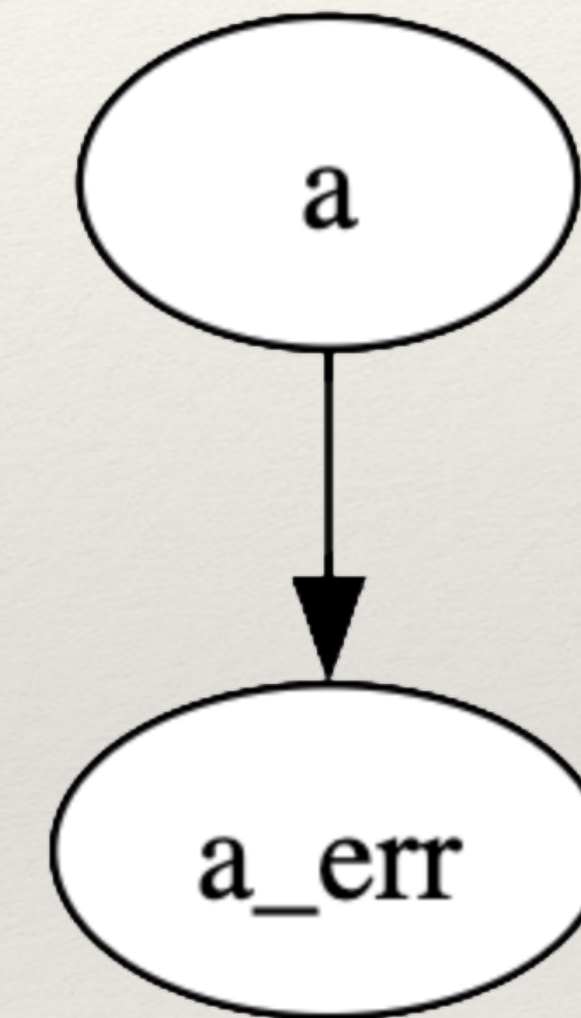
```
import numpy as np
import modeldag

def foo(value, scale=0.05, floor=0.2):
    """ the quadratic sum between value*scale and floor """
    return np.sqrt( (value*scale)**2 + floor**2)

model = {"a": {"func": np.random.normal,
              "kwargs": {"loc":10, "scale":2}},
         "a_err": {"func": foo,
                  "kwargs": {"value":"@a"} # leaving other param unchanged
                  }
         }

mdag = modeldag.ModelDAG(model)
mdag.visualize()
```

DAG



Easy to build *complex* transient

DAG

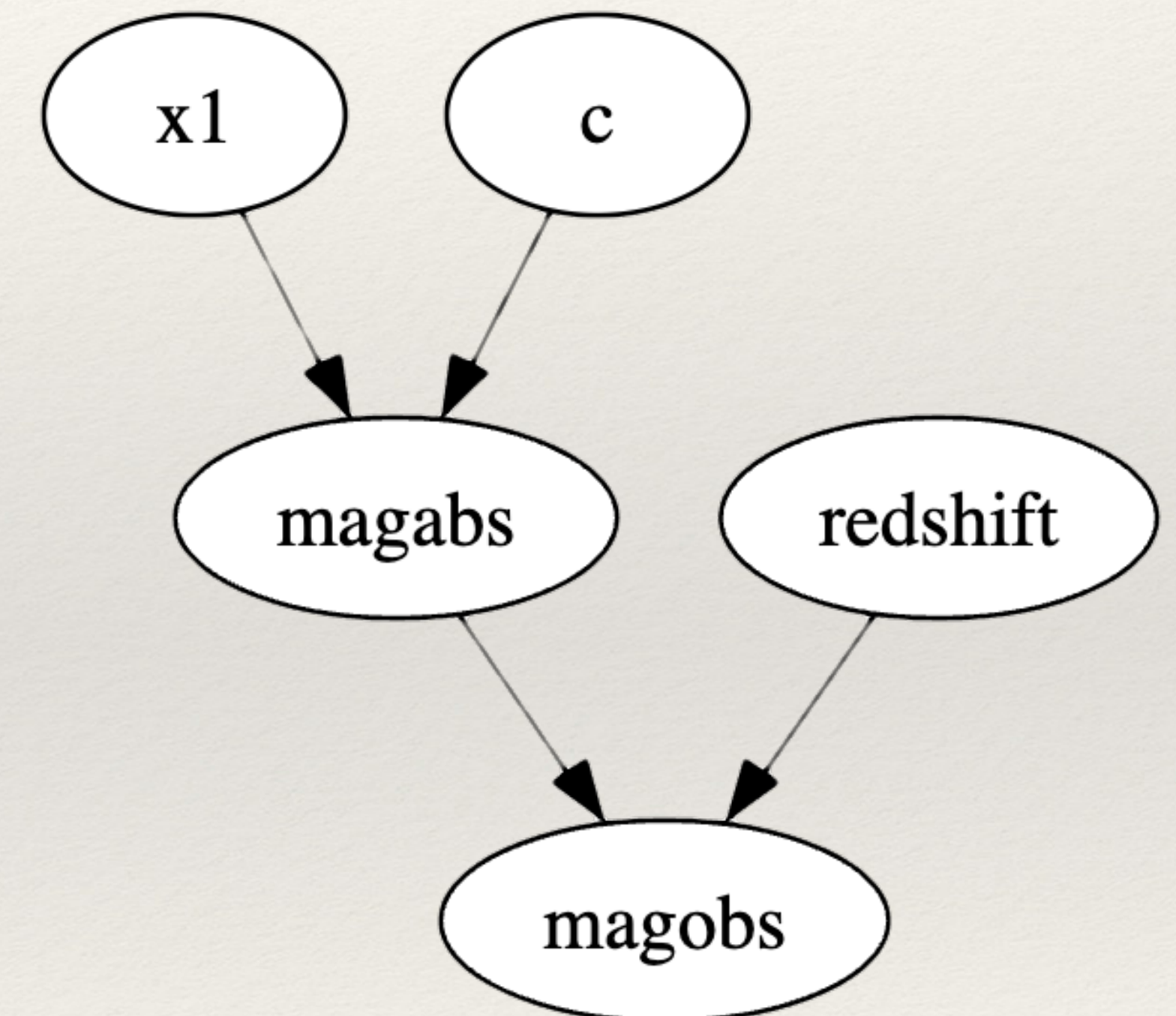
```
# ----- #
# Your functions #
# ----- #
def tripp98(x1, c, alpha=-0.14, beta=3.3, sigmaint=0.1, mabs=-19.3):
    """ This is the Tripp 98 relation """
    magabs = x1*alpha + c*beta + mabs # natural SN Ia magnitude (non-standardized)
    magabs_scattered = np.random.normal(loc=magabs, scale=sigmaint) # add intrinsic scatter
    return magabs_scattered

def magabs_to_magobs(magabs, redshift, cosmology=Planck18):
    """ distance_modulus(cosmo) = m - M """
    mu = Planck18.distmod(redshift).value
    return mu + magabs

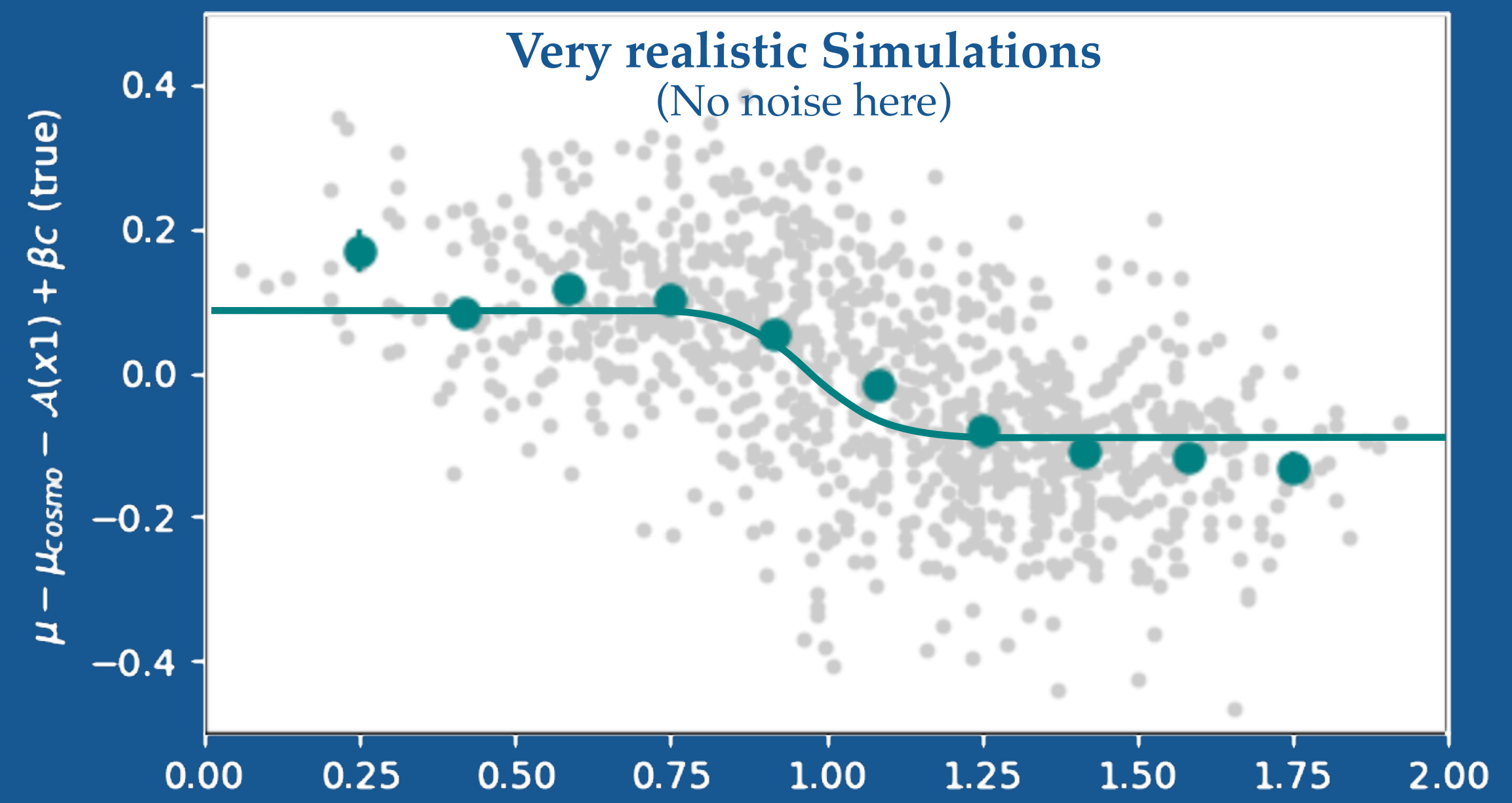
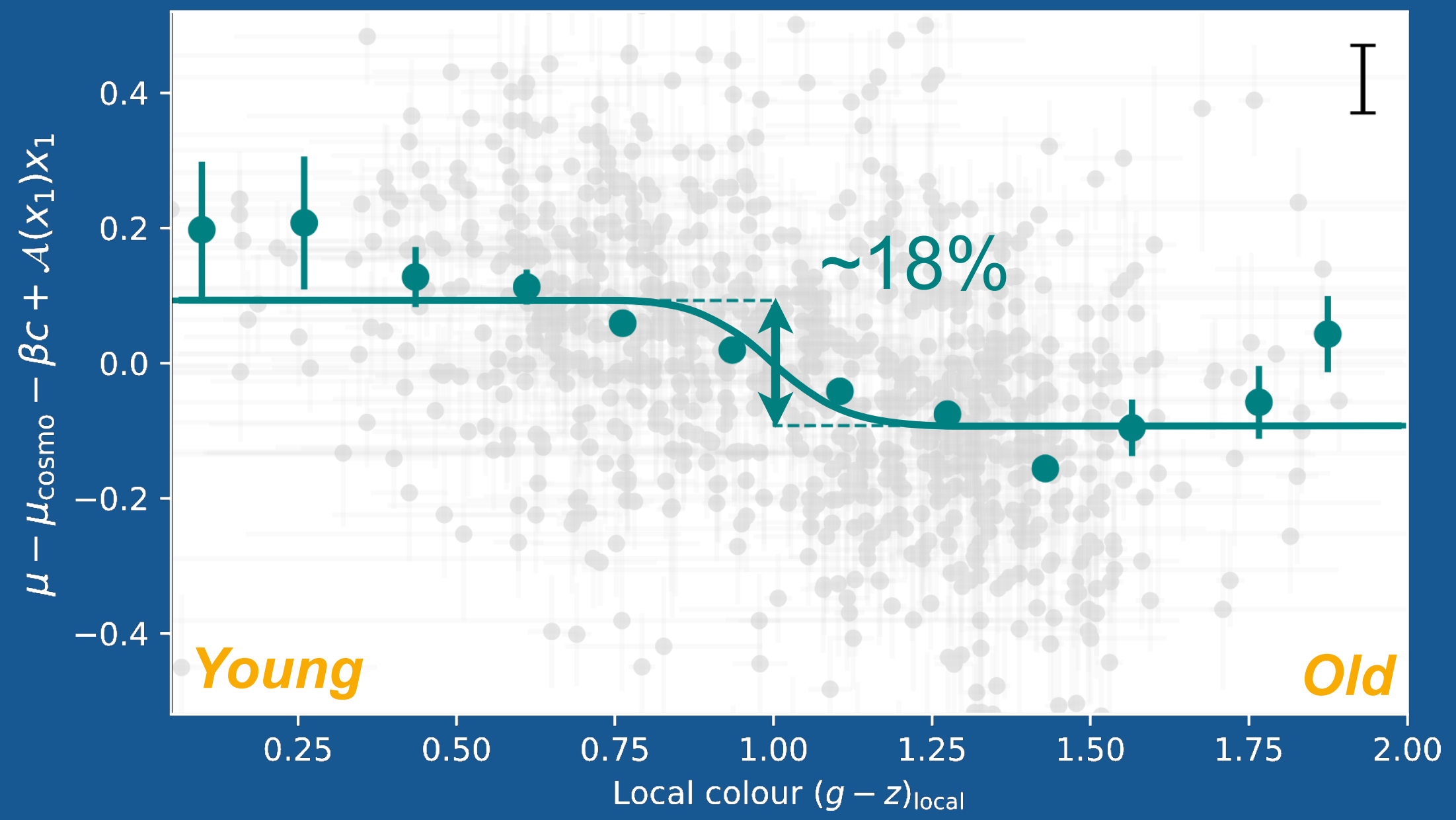
# ----- #
# Create the dict #
# ----- #
modeldict = {"x1": {"func": np.random.normal, "kwargs": {"loc": 0, "scale": 1}},
            "c": {"func": stats.lognorm.rvs, "kwargs": {"s": 0.7, "loc": -0.2, "scale": 0.2}},
            "magabs": {"func": tripp98,
                      "kwargs": {"x1": "@x1", "c": "@c"}},
            "redshift": {"func": draw_from_rate},
            "magobs": {"func": magabs_to_magobs,
                      "kwargs": {"magabs": "@magabs", "redshift": "@redshift"}},
            }

# ----- #
# Simulate #
# ----- #
mdag = modeldag.ModelDAG(modeldict)
data_sim = mdag.draw(1_000)
```

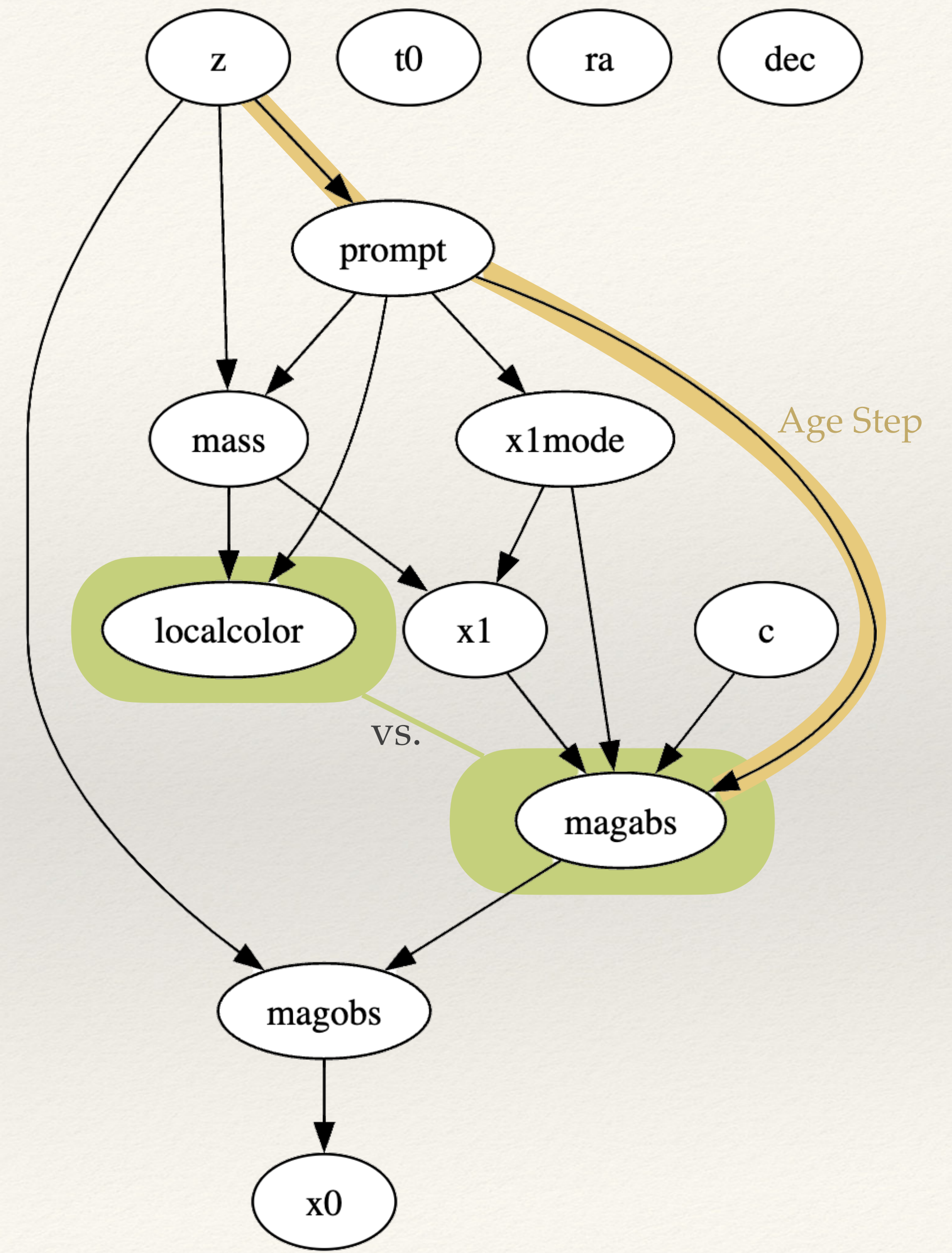
```
mdag.visualize()
```



● All data - - - Step model — Step model incl. (g - z) error



skysurvey | modeldag



snia.data

	z	x1	c	t0	ra	dec	magabs	mwebv	magobs	x0	template
0	0.07705	-0.030	1.221450	58487.214844	336.697937	7.605778	-15.358439	0.134286	22.429092	0.000017	salt2
1	0.06865	-0.700	-0.027583	58659.839844	60.706551	55.346775	-19.233326	0.719195	18.290983	0.000766	salt2
2	0.06905	0.320	0.351301	58782.136719	23.629887	-0.468616	-18.197527	0.036892	19.339998	0.000292	salt2

sn_07pk.data

	z	t0	magabs	ra	dec	magobs	amplitude	template
0	0.04905	56041.375000	-18.313118	213.854660	-78.848862	18.451269	1.207582e-15	v19-2007pk-corr
1	0.03315	56130.003906	-16.217342	310.827576	17.159157	19.671362	3.925357e-16	v19-2007pk-corr
2	0.04845	56179.964844	-14.626013	337.447815	37.145824	22.110720	4.150845e-17	v19-2007pk-corr

```
snii = skysurvey.SNeII.from_draw(1000)
snii.data
```

	template	z	t0	magabs	ra	dec	magobs	amplitude
0	v19-2016x-corr	0.02965	56173.551302	-17.000019	313.871430	-32.681907	18.640832	3.438910e-15
1	v19-2016bkv-corr	0.04335	56078.601439	-16.390584	294.182082	23.938362	20.096692	8.948369e-15
2	v19-asassn14jb-corr	0.02645	56139.120893	-16.300129	44.165477	-52.273437	19.087617	7.501353e-15

Customization is *easy*

rate, redshift range, absolute magnitude etc. when calling `[from_]draw()`

Pre-Built Transients

Transient template is built on top of `sncosmo`

Any `sncosmo` built-in source is available:

<https://sncosmo.readthedocs.io/en/stable/source-list.html>

Type Ia Supernovae

```
snia = skysurvey.SNeIa()
```

Transient Time Serie

```
sn_07pk = skysurvey.TSTransient("v19-2007pk-corr")
sn_07pk = skysurvey.TSTransient.from_draw(1000,
                                           template="v19-2007pk-corr")
```

Kilonova

```
kilonova = skysurvey.Kilonova()
```

Supernova Types

“Collection
Transient Time Series”

```
snii = skysurvey.SNeII.from_draw(1000)
```

SNeII | SNeIIb | SNeIIIn | SNeIIb | SNeIIc | SNeIIcBL
 23 12 6 12 7 6

Vincenzi+2019

See `skysurvey` doc for how to build a transient blackbody

pip install skysurvey

targets

```
import skysurvey
snia = skysurvey.SNeIa.from_draw(10_000, zmax=0.3,
                                tstart = "2018-04-01",
                                tstop = "2020-12-01")
```

survey

```
import pandas
from shapely import geometry

# footprint | 2-deg radius circle
footprint = geometry.Point(0,0).buffer(2)

# stored observing logs | ra, dec, zp, skynoise, gain, filter ...
data = pandas.read_parquet("observing_logs.parquet")

survey = skysurvey.Survey.from_pointings(data, footprint=footprint)
```

dataset

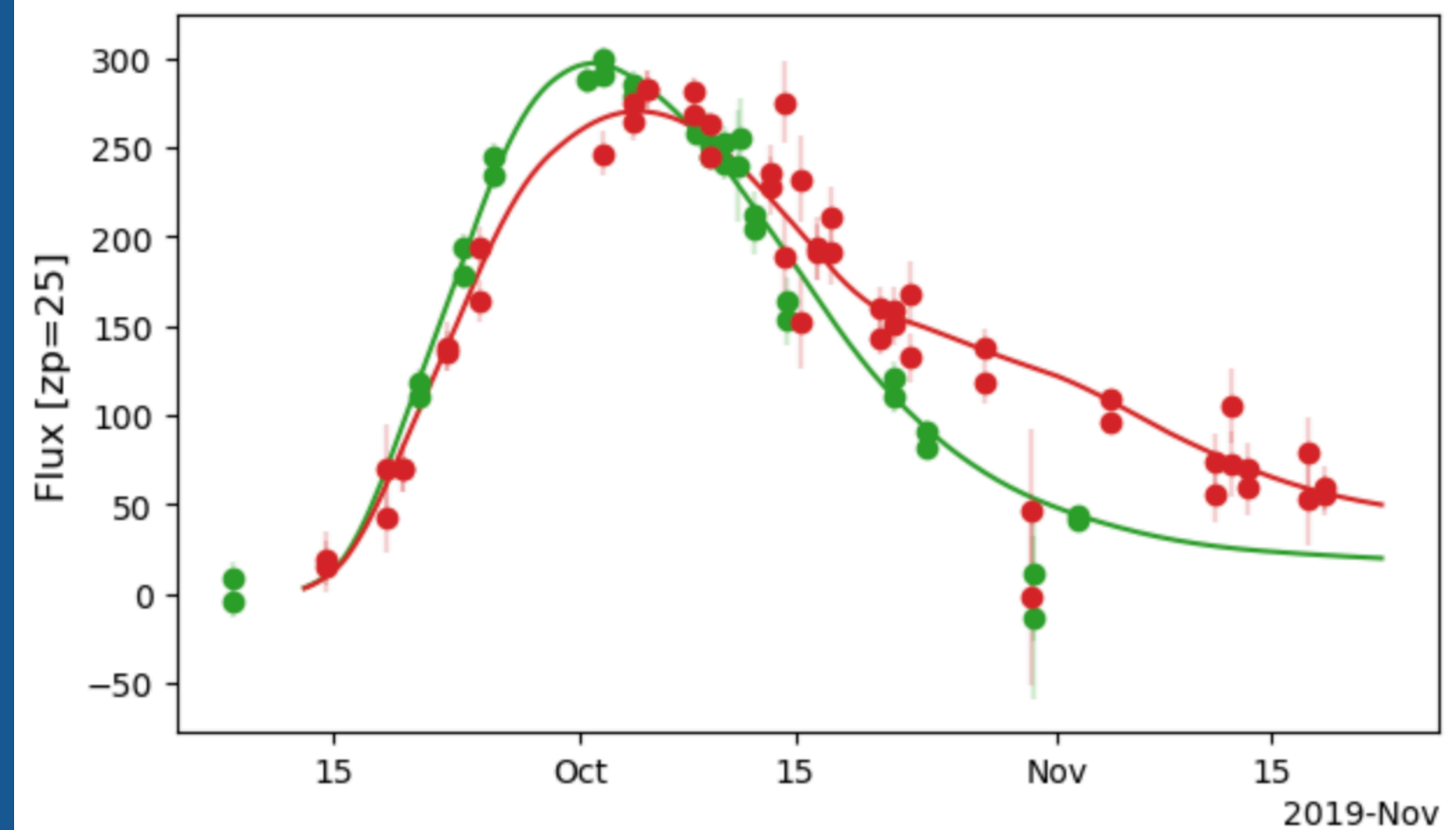
```
dset = skysurvey.DataSet.from_targets_and_survey(snia, ztf)
```

dset.data

	<i>survey</i> index	fieldid	rcid	time	band	flux	fluxerr	zp	zpsys
5821	26009040	1	3	59104.390625	ztf	-39.302373	36.143482	25.233383	ab
	26054356	1	3	59106.328125	ztf	8.518026	14.717361	24.559206	ab
	26331019	1	3	59112.300781	ztf	24.598649	20.259558	26.025703	ab
	26338533	1	3	59112.359375	ztf	-3.302169	36.260182	26.127666	ab
	27121580	1	3	59128.250000	ztf	-20.931607	46.604773	25.959711	ab
...
9727	30211656	1408	16	59203.179688	ztf	14.887892	60.705860	26.204699	ab
	30211715	1408	16	59203.179688	ztf	90.188222	61.295684	26.208666	ab
	30841192	1408	16	59221.167969	ztf	34.804501	22.752532	26.142845	ab
	36571545	1408	16	59392.414062	ztf	-9.636968	41.250546	26.149637	ab
	36571604	1408	16	59392.414062	ztf	-18.475494	44.445759	26.151228	ab

8429351 rows x 8 columns

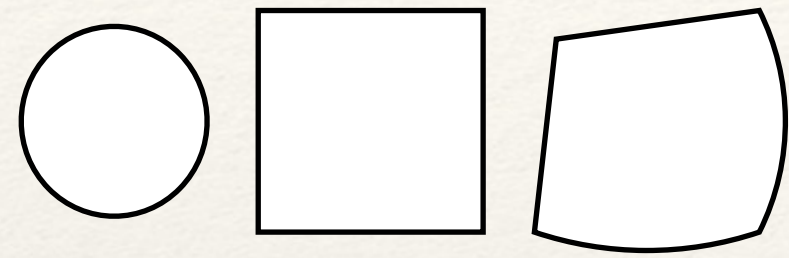
```
fig = dset.show_target_lightcurve(index=347, phase_window=[-40,100])
```



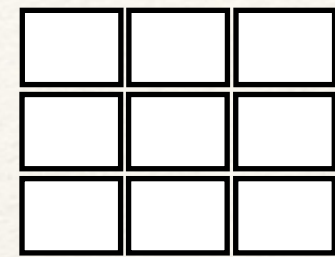
Build your own Survey

footprint | camera footprint in the sky

shapely



Polygon



MultiPolygon



`skysurvey.get_footprint("des")`

data | what was observed when, under which conditions

pandas

ra, dec — Camera pointings

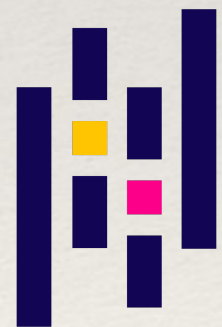
mjd — time in Modified Julian Date

band — name of the filter used

skynoise — level of the sky noise (unit of *zp*)

zp — depth of the image

gain — camera gain (e-/adu), used for noise



```
import numpy as np
from shapely import geometry
from skysurvey.tools import utils

# footprint | 2-deg radius circle
footprint = geometry.Point(0,0).buffer(2)

# fake observations
size = 100_000 # n-pointings
data = {}
data["gain"] = 1
data["zp"] = 30
data["skynoise"] = np.random.normal(size=size, loc=200, scale=20)
data["mjd"] = np.random.uniform(58_800, 59_600, size=size)
data["band"] = np.random.choice(["desg", "desr", "desi"], size=size)

data["ra"], data["dec"] = utils.random_radec(size=size,
                                             ra_range=[10,350],
                                             dec_range=[-50,10])

# Load the survey
survey = skysurvey.Survey.from_pointings(data, footprint=footprint)
```

```
# footprint | 2-deg radius circle
footprint = geometry.Point(0,0).buffer(2)

# fake observations
data = pandas.read_parquet("observing_logs.parquet")

# Load the survey
survey = skysurvey.Survey.from_pointings(data, footprint=footprint)
```

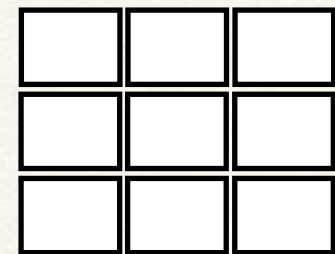
Build your own GridSurvey

footprint | *camera footprint in the sky*

shapely



Polygon



MultiPolygon



skysurvey.get_footprint("des")



fields | *Coordinates of the grid center*

pandas

{**fieldid**: {**ra**: *value*, **dec**: *value*}}

data | *what was observed when, under which conditions*

pandas

ra, dec — Camera pointings

mjd — time in Modified Julian Date

band — name of the filter used

skynoise — level of the sky noise (unit of zp)

zp — depth of the image

gain — camera gain (e-/adu), used for noise



```
import numpy as np
import skysurvey
from shapely import geometry
```

```
# footprint
footprint = geometry.Point(0,0).buffer(2)
```

```
# fields
fields_radec = { 'C1': {'dec': -27.11161, 'ra': 54.274292+180},
                  'C2': {'dec': -29.08839, 'ra': 54.274292+180},
                  'C3': {'dec': -28.10000, 'ra': 52.648417+180},
                  'E1': {'dec': -43.00961, 'ra': 7.8744167+180},
                  'E2': {'dec': -43.99800, 'ra': 9.5000000+180},
                  'S1': {'dec': 0.00000, 'ra': 42.820000+180},
                  'S2': {'dec': -0.988389, 'ra': 41.194417+180},
                  'X1': {'dec': -4.929500, 'ra': 34.475708+180},
                  'X2': {'dec': -6.412111, 'ra': 35.664500+180},
                  'X3': {'dec': -4.600000, 'ra': 36.450000+180}
                }
```

```
# observing logs
size = 100_000
data = {}
data["gain"] = 1
data["zp"] = 30
data["skynoise"] = np.random.normal(size=size, loc=200, scale=20)
data["mjd"] = np.random.uniform(58_800, 59_600, size=size)
data["band"] = np.random.choice(["desg", "desr", "desi"],
                                size=size)
```

```
# - no ra, dec - #
data["fieldid"] = np.random.choice(list(fields_radec.keys()),
                                    size=size)
```

```
# Load a GridSurvey
survey = skysurvey.GridSurvey.from_pointings(data, fields_radec,
                                             footprint=footprint)
```