# Classification of Supernovae

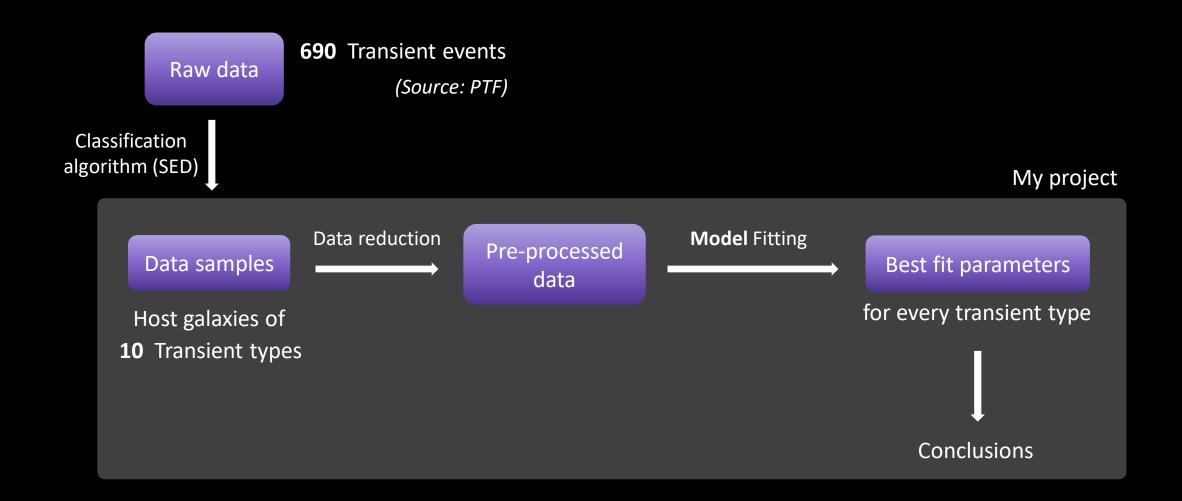
**Amit Gal** 

## Goal

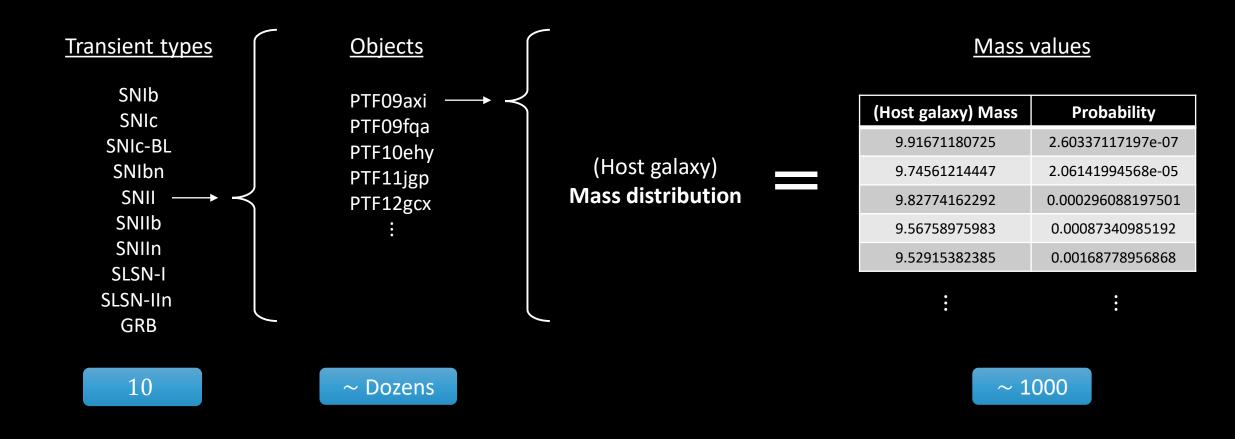
Which supernovae **types** can we classify by their host galaxy mass distributions?



# Project outline



## Data organization



690 objects with (host galaxy) mass distributions

# Data reduction – Sampling (bootstrapping)

Data reduction

#### For each object

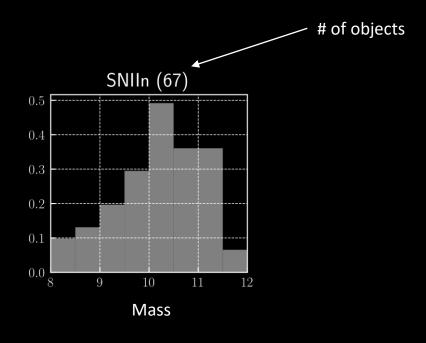
#### PTF09axi

(Host galaxy) Mass	Probability
9.91671180725	2.60337117197e-07
9.74561214447	2.06141994568e-05
9.82774162292	0.000296088197501
9.56758975983	0.00087340985192
9.52915382385	0.00168778956868

random sampling of one mass value

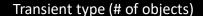


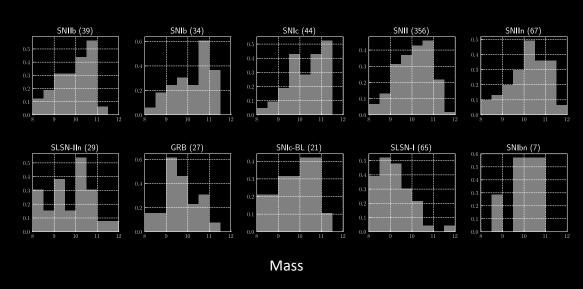
#### For 1 transient type



# Data reduction – Sampling (bootstrapping)

## Mass distributions of host galaxies

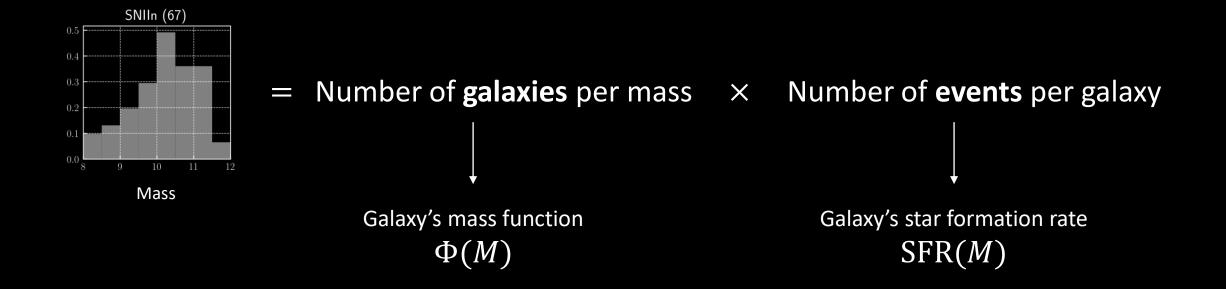




= 1 bootstrap sample

# Model

# Mass distribution of host galaxies



# Mass function $\Phi(M)$

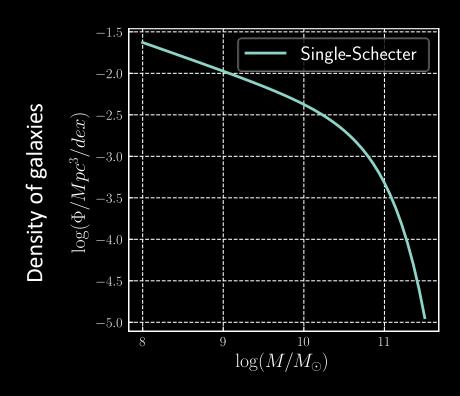
Number **density** of galaxies as a function of their **mass**.

## The single-Schechter function (1976)

$$\Phi(M)dM = \ln(10) \cdot \Phi^* \cdot 10^{(1+\alpha)(\log M - \log M^*)} \cdot \exp(-10^{(\log M - \log M^*)}) dM$$

$\log M^*$	$\Phi^*$	α
10.81	$11.34 \cdot 10^{-4}$	-1.34

Muzzin et al. (2013)



Mass

# Star formation main-sequence (SFR)

The relationship between a galaxy's **star formation rate** (SFR) and its stellar **mass**.

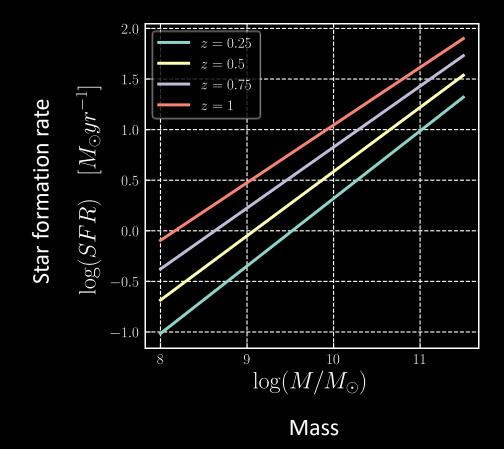
$$\log(\mathbf{SFR}) = \alpha(z)(\log(\mathbf{M}) - 10.5) + \beta(z)$$

$$\alpha(z) = 0.70 - 0.13z$$

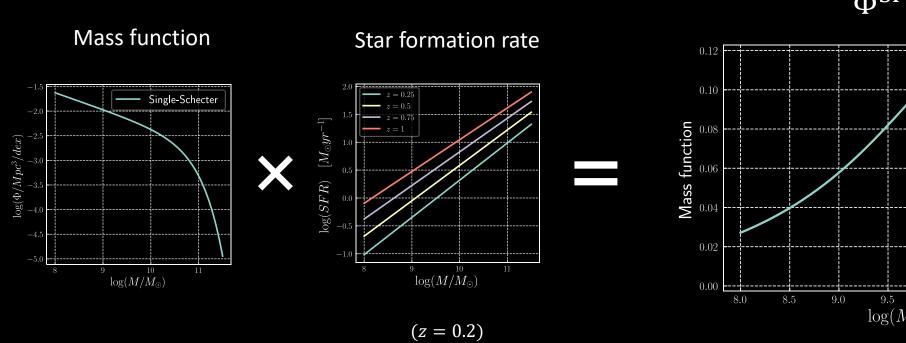
$$\beta(z) = 0.38 + 1.14z - 0.19z^2$$

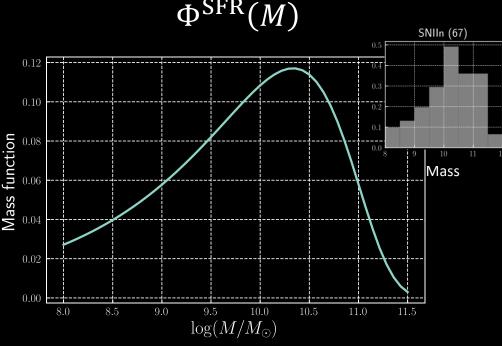
z - redshift

Whitaker et al. (2014)



# The mass function weighted by SFR





How to differentiate between transient types?

# Star formation efficiency function

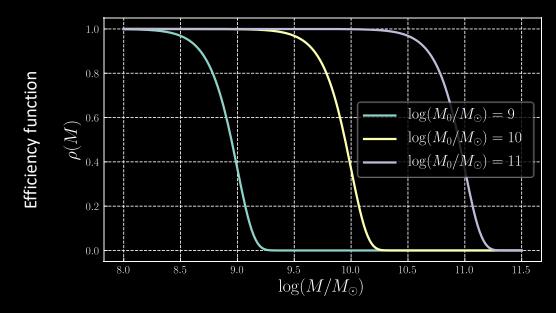
$$\Phi^{\rm SFR}(M) \longrightarrow \underbrace{\rho(M; M_0, \beta.C)}_{\rm efficiency} \cdot \Phi^{\rm SFR}(M)$$

$$\rho(M; M_0, \beta, C) = C \exp \left\{ -\left(\frac{M}{M_0}\right)^{\beta} \right\}$$

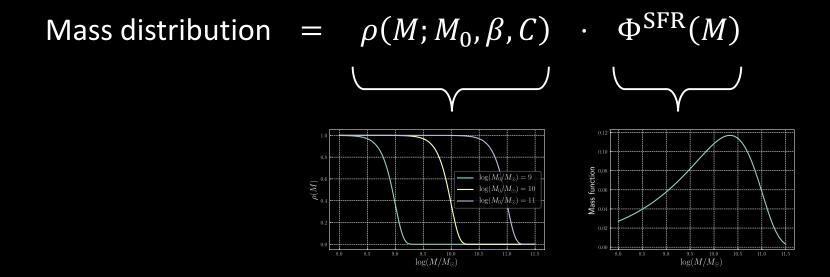
 $M_0$  cut-off mass

 $\beta$  cut-off strength

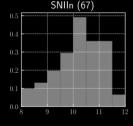
C fudge factor



# Final expression of the model



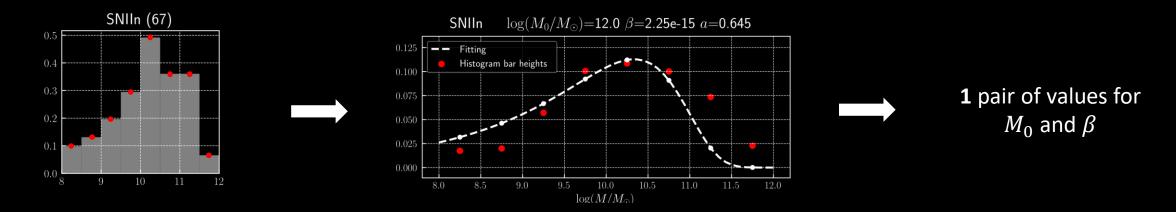
Model **fitting** to



with 3 free parameters:  $M_0$ ,  $\beta$ , C

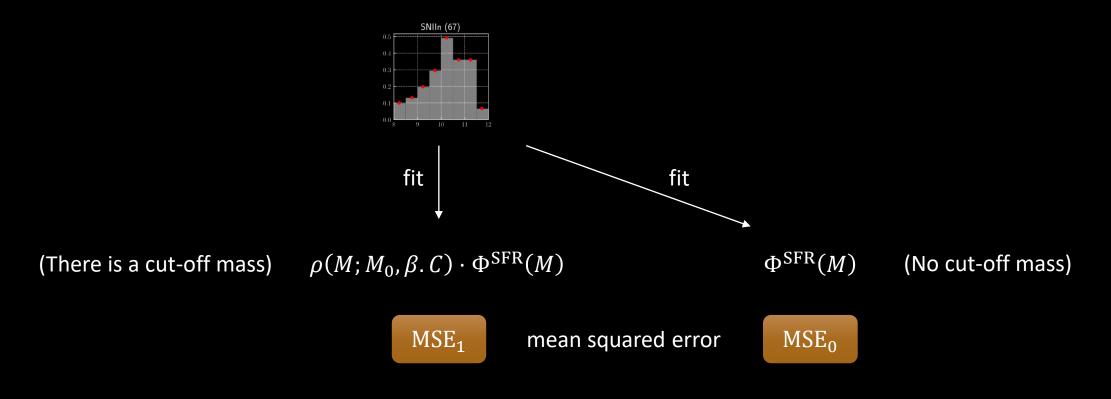
# Model fitting

# Fitting the cut-off mass $M_0$ and strength eta



with random bin centers

# Labeling unsuccessful fitting



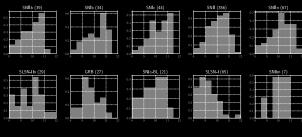
Successful fitting

 $MSE_1 < MSE_0$ 

# Using 1 bootstrap sample

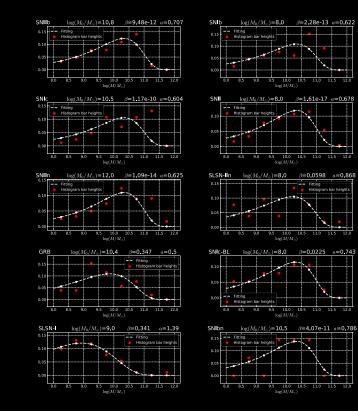
## 1 bootstrap sample

#### Transient type (# of objects)



Mass

## Fitting for every transient type



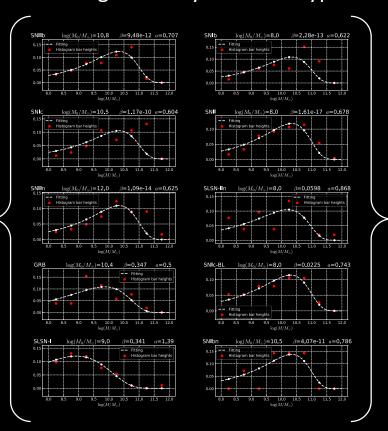
For every transient type:

f 1 pair of values for  $M_0$  and eta

# Using **N** bootstrap sample

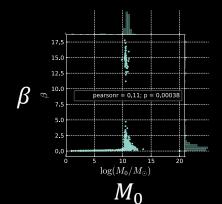
# Transient type (# of objects) SNIB (39) SNIB (39) SNIB (39) SSNIB (39) SS

## Fitting for every transient type



For every transient type:

**N** pairs of values for  $M_0$  and  $\beta$ 



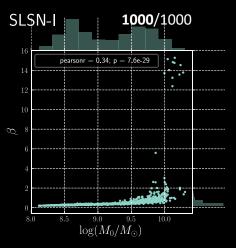
# Model fitting - Results

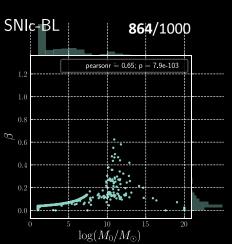
## Fitting results

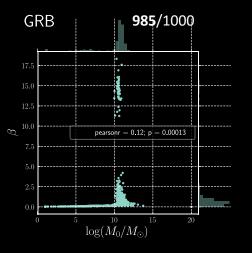
successful/total

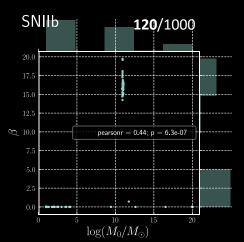
**1000** bootstrap samples

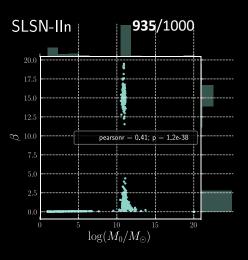


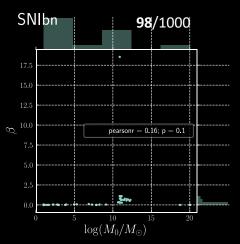






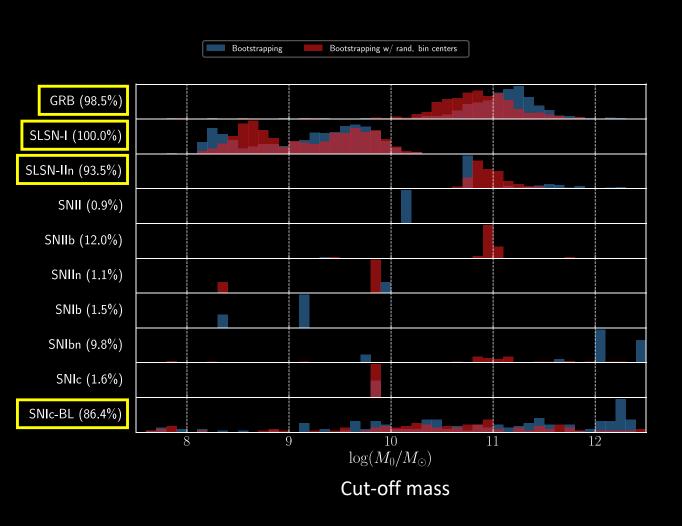


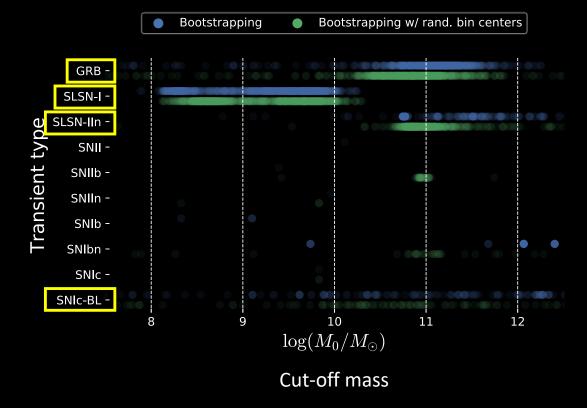




**Cut-off mass** 

# Fitting results — Cut-off mass $M_0$





## Summary

Data analysis – creating python package and notebook

## Results:

– Only 4 transient types show prominent suppression at high mass galaxies:

SLSN-I, SLSN-II, GRB, SNIc-BL

Cut-off metallicity can be calculated using mass-metallicity relation

## Further directions:

- Deriving hypothesis testing  $H_0$ ,  $H_1$  will have different  $M_0$  values, where  $\ln \mathcal{L}(M_0) = \sum_{i=1}^n \ln \Phi^{\rm SFR}(m_i | M_0)$
- Using different data luminosity instead of mass
- Re-estimation when new data arrives