# Exploration of rapidly evolving transients with the Subaru/HSC transient survey Seiji Toshikage, Masaomi Tanaka, Ichiro Takahashi (Tohoku Univ),

Naoki Yasuda, Nao Suzuki (Kavli IPMU) Tominaga Nozomu, Takashi J. Moriya (NAOJ) and HSC transient working group

### Summary

- A number of Supernovae and peculiar transients (SuperLuminous, rapidly evolving, etc.) are discovered by recent transient surveys - a method for transients classification without spectroscopic data is needed
- Subaru/Hyper Suprime-Cam survey discovered 1824 Supernovae in the COSMOS field
- We developed a machine learning multiclass classifier for type Ia, Ibc, II supernovae and rapidly evolving transients
- New rapidly evolving transients were identified at redshift z = 0.70 1.0
- Event rate of rapidly evolving transients is estimated ~ 2% of Core-Collapse Supernovae
  - type Ibn supernovae ~ 0.4% of CCSNe  $\rightarrow \geq 2\%$  of He star experience eruptive mass loss before the explosion

### **Rapidly evolving transients (RETs)**

#### • optical transients with a short timescale

- time above half max  $\leq 10$  day
- unsolved mechanism
- Subdominant types of RETs (1) **AT2018cow like** 2 Type Ibn (CSM) ③ Type IIb (SCE) by Ho et al. 2021



## Subaru/HSC transient survey

- The Hyper Suprime-Cam-SSP transient survey program
  - 2016.11 2017.04 - COSMOS UD/D field  $1.77/5.78 \, deg^2$ - g, r, i, z band
- 1824 Supernovae spec-z 759 objects photo z 957 objects No redshift 108 objects



### Method - Classification of rapidly evolving transients -

#### $(\mathbf{1})$ Datasets

training data set by simulation

| la       | SALT2 model template         |
|----------|------------------------------|
| lbc • II | observational template       |
| RETs     | Arnett82 semi-analytic model |

#### test data set by observation COSMOS UD 879 SNe





**2** Gaussian Process Regression (GPR)

connect sparse data with smooth curves

### **3** Feature extraction

60 features per one object e.g. peak flux

decline time from peak to half

flux



③ Feature extraction

peak

**(4)** classification by machine learning

Random Forest (supervised learning) Validation with simulation data

Accuracy 0.96(all) 0.97 (RETs)

classification of real data

#### 66/879 objects classified as RETs



### **Results - New rapidly evolving transients**

66 RET candidates by machine Learning selection by the light curve

## **Discussion - phase diagram and event rate**

Estimation of event rate with HSC RETs

**9 RETs identified** (4 objects reported by Tampo+ 20)

#### Light curve comparison with RET samples





 $r = \sum_{i}^{N} \frac{(1+z_i)}{\epsilon T_i V_{max,i}} \simeq 6000 \text{ yr}^{-1} \text{ Gpc}^{-3}$ 

~ 2% of Core-Collapse Supernovae

 $lbn \simeq 1150 \text{ yr}^{-1} \text{ Gpc}^{-3}$  $\simeq$  0.4% of CCSNe

Ib  $\simeq 20\%$  of CCSNe Shivers et al. 2017  $\rightarrow$  2% of He star eruptive mass loss