

The GaRSG Project: Red Supergiants on the Brink of Core-Collapse

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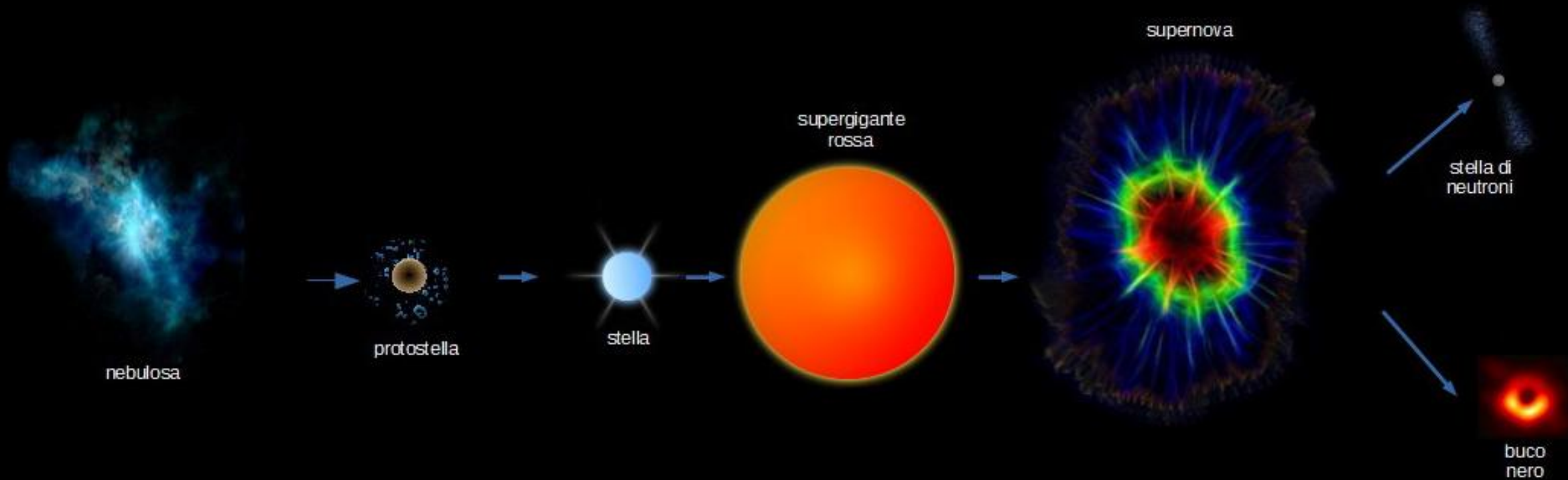


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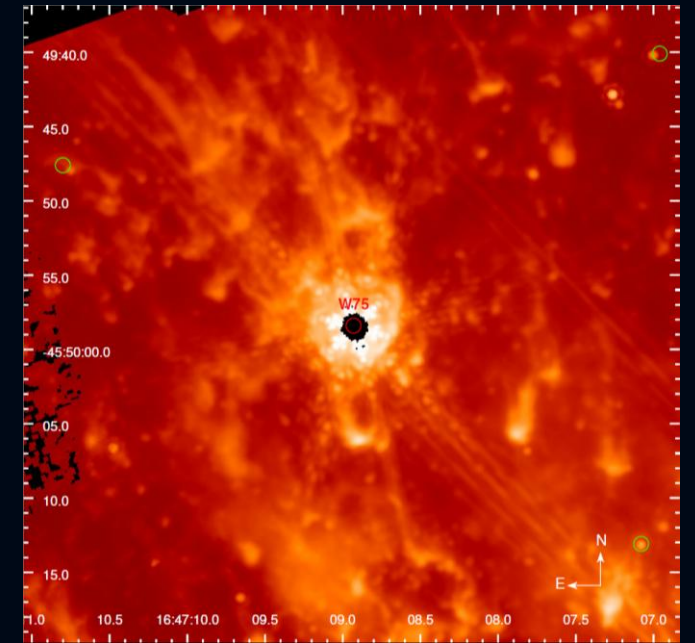
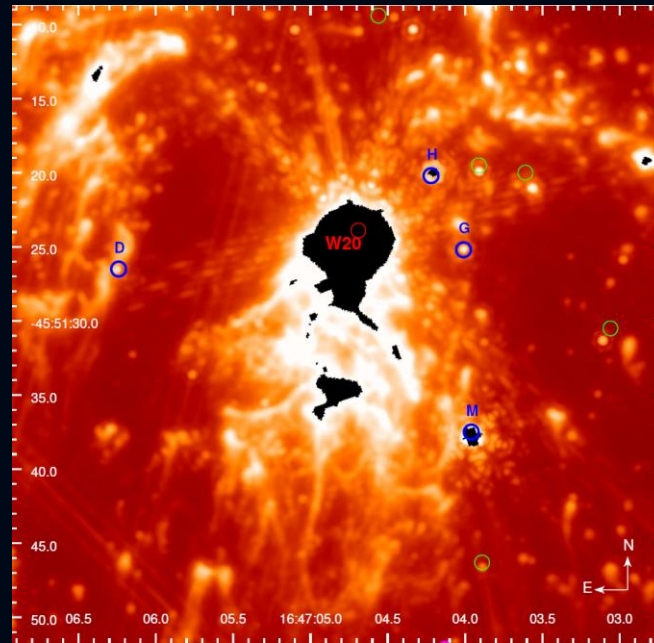
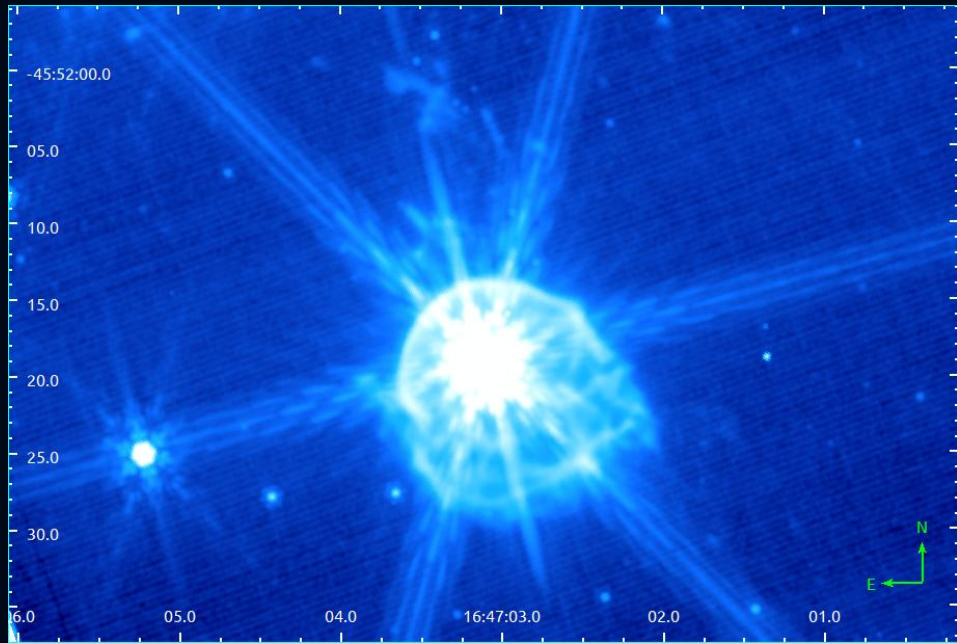
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INAF School of AI

STELLAR EVOLUTION



WHAT ARE WE LOOKING FOR?

MASS-LOSS EVENTS OF RED SUPERGIANTS



Circumstellar Medium (CSM) around Red Supergiants with JWST
(Guarcello+25, *Astronomy & Astrophysics*)

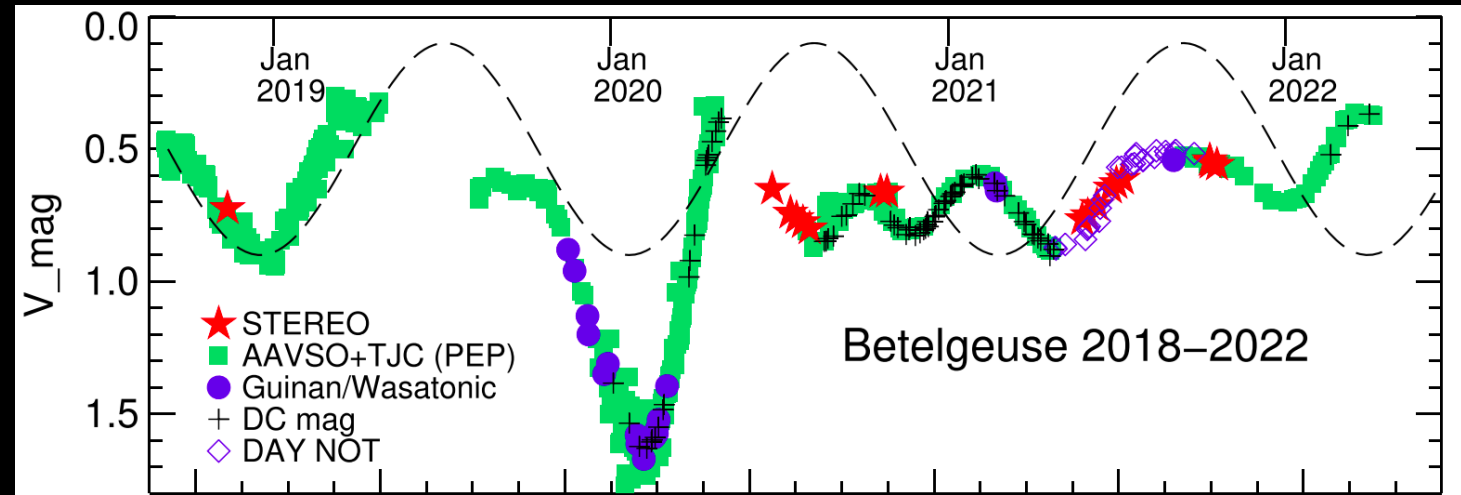
RSGs can undergo brief but intense mass-loss episodes on timescales ranging from thousands of years to centuries, years, months, and even days before the supernova.

PRE-SUPERNOVA OUTBURST: THE CASE OF BETELGEUSE

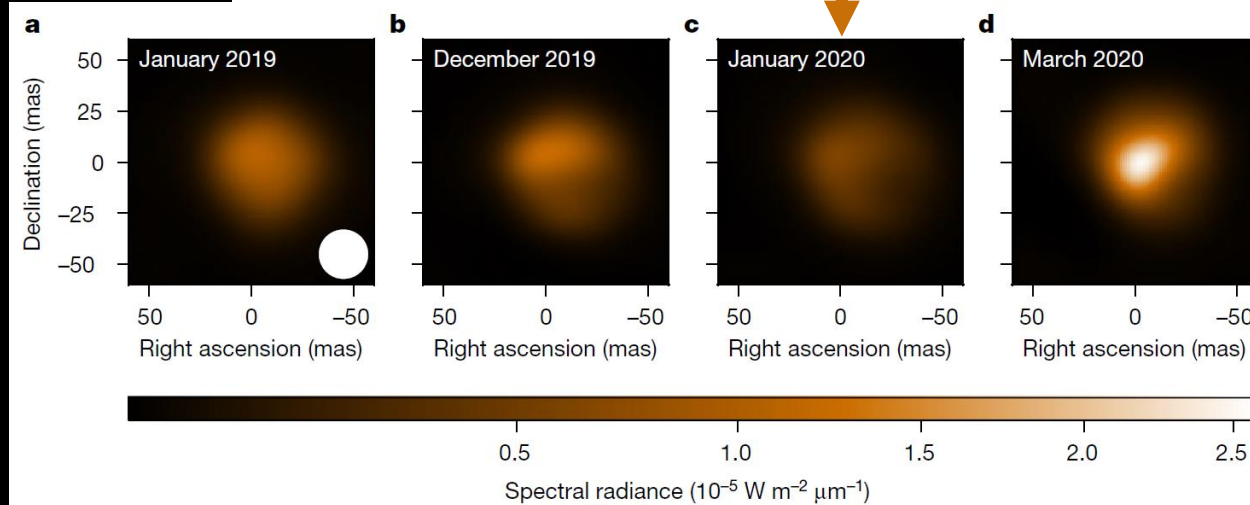
Energy injection in the envelope produces detectable modifications in the light curves.

(Jacobson-Galán+22, Davies+22, Morozova+18,20)

Dupree+22, *The Astrophysical Journal*



JD - 2458000



Montargès+21, *Nature*

GaIRSG CATALOGUE AND ONGOING OBSERVING CAMPAIGN



STAR CATALOGUE

Lauriano et al. 2026, in press,
*A New Catalogue of Galactic Red
Supergiants for direct detection of
episodic mass-loss events*

• 2.6 m INAF-VST (Paranal, Chile)

Focused on **field RSGs** since 2023
5 ± 2 days cadence photometry in i,z

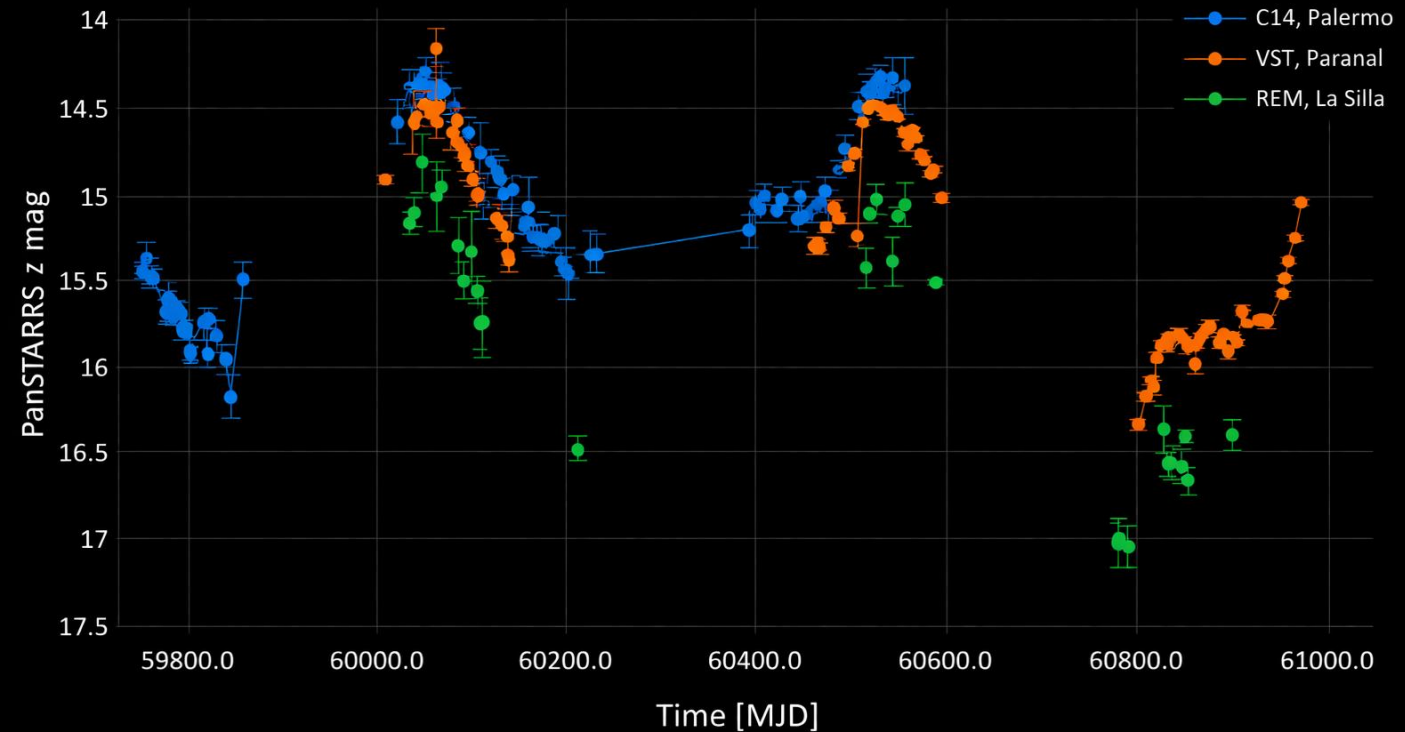
• 60 cm INAF-REM (La Silla, Chile)

Monitoring of **RSGs in clusters** since 2023
2-day cadence in g,r,i,z and J,H,K

CLASSIFICATION LIGHT CURVES

ACTUALLY MORE THAN 2,000 LIGHT CURVES JUST FOR GALACTIC TARGETS!

Light curve in z band of Star 2MASS J18443885-0326135



VARIABILITY

INTRINSIC

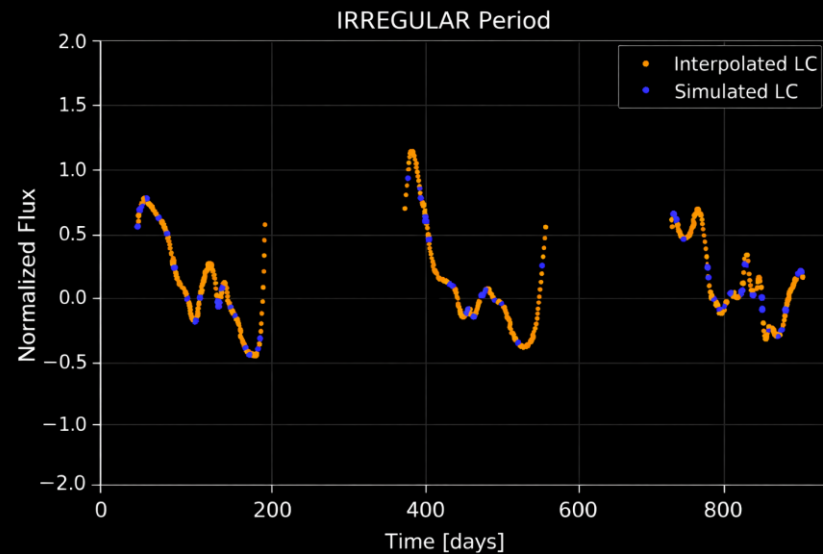
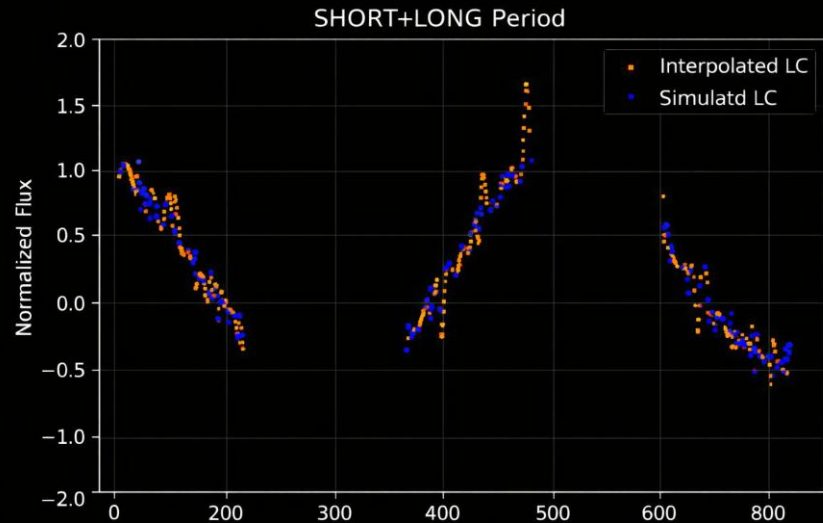
EXTRINSIC

- Radial Pulsations;
- Photospheric convection;
- **Nuclear Flashes;**
- ...

- Effect from a binary companion;
- ...

MACHINE LEARNING CLASSIFIER

Simulated light curves of five intrinsic variability classes: from (semi)regular to irregular periodicity and amplitude.



We classified simulated light curves by comparing different supervised ML algorithms (RandomForest, XGBoost, Traditional NN, CNN,...)

Best results was returned from *XGBoost*

train upsampled

| | | | | | |
|-----------|----------|-----------|---------|--------|--------|
| Constant | 254 | 0 | 17 | 0 | 0 |
| Irregular | 0 | 715 | 0 | 1 | 0 |
| LSPonly | 60 | 0 | 530 | 2 | 0 |
| SP+LSP | 3 | 20 | 34 | 156 | 61 |
| SPonly | 8 | 10 | 4 | 76 | 49 |
| | Constant | Irregular | LSPonly | SP+LSP | SPonly |

Actual

Predicted

FUTURE AIMS

- Improving the photometry of light curves and then performing **variability classification on real data** using ML algorithms (*XGBoost, RNNs ?*);
- **Anomaly Detection** using ML algorithms (*Variational Autoencoders ?*) in (semi)regular and irregular RSGs, with an ad-hoc alert system;
- **Expanding the sample** from GalRSG to ExtraGalRSG in order to monitor tens of thousands of RSGs instead of hundreds;
- Images and spectroscopic **follow-up** of putative ongoing outburst events.