

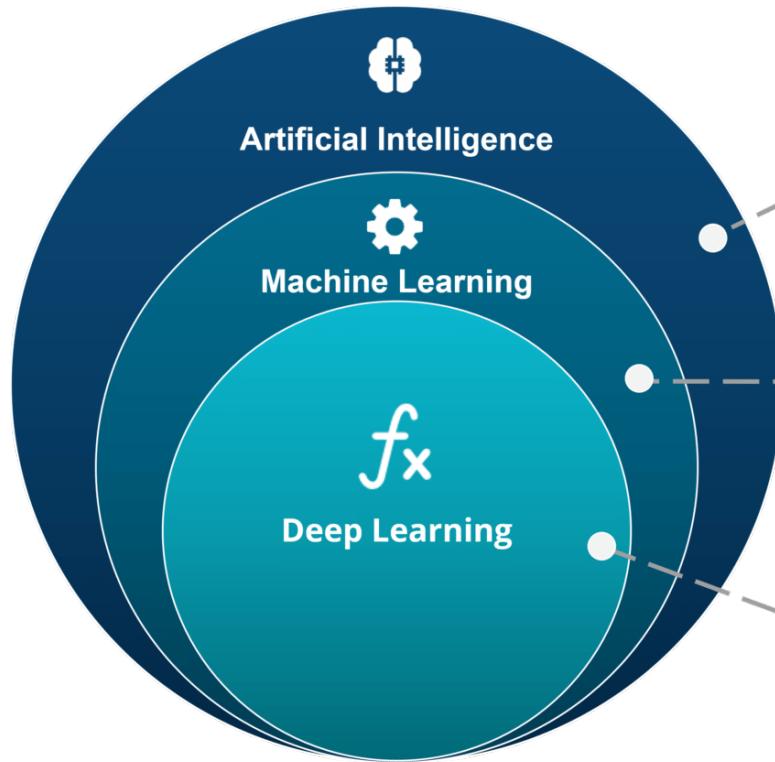
Inteligencia artificial aplicada a astronomía

Paula Sánchez Sáez

ESO Fellow

Miembro del Proyecto ALeRCE

Que es la inteligencia artificial?



ARTIFICIAL INTELLIGENCE

A technique which enables machines to mimic human behaviour

MACHINE LEARNING

Subset of AI technique which use statistical methods to enable machines to improve with experience

DEEP LEARNING

Subset of ML which make the computation of multi-layer neural network feasible

ML vs técnicas tradicionales

Programación tradicional

```
In [1]: import numpy as np

def produce_a_SN(star_mass):
    if star_mass > 8:
        return(True)
    else:
        return(False)

a = 10
print(produce_a_SN(a))
```

True

Machine Learning

```
In [1]: import numpy as np

def produce_a_SN(star_mass):
    if star_mass > 8:
        return(True)
    else:
        return(False)

a = 10
print(produce_a_SN(a))
```

True

ML vs técnicas tradicionales

Programación tradicional

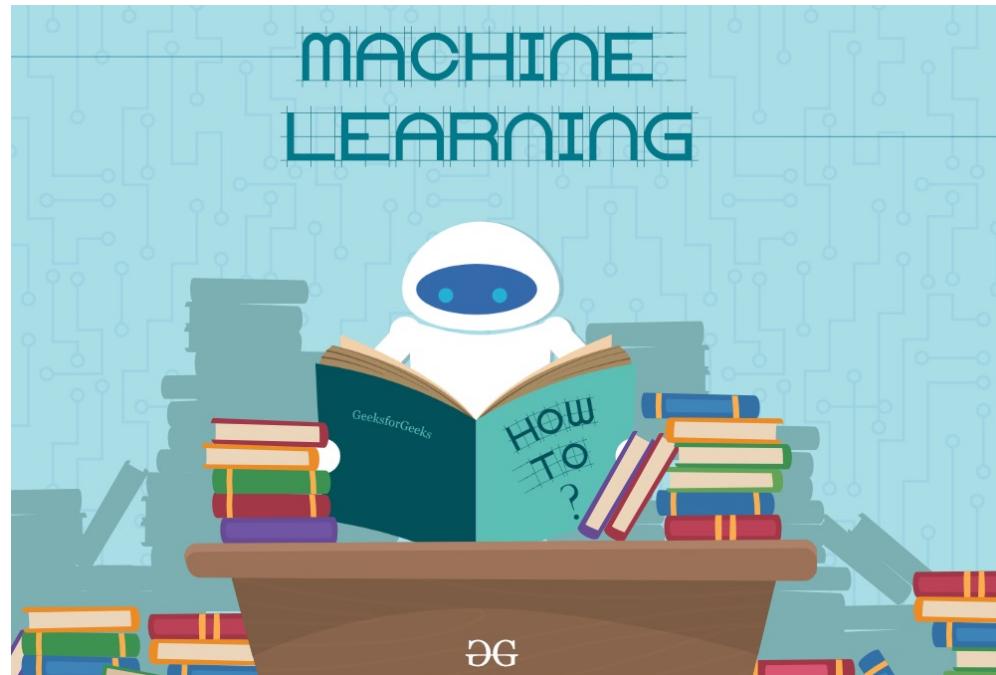
In [1]:

```
import numpy as np

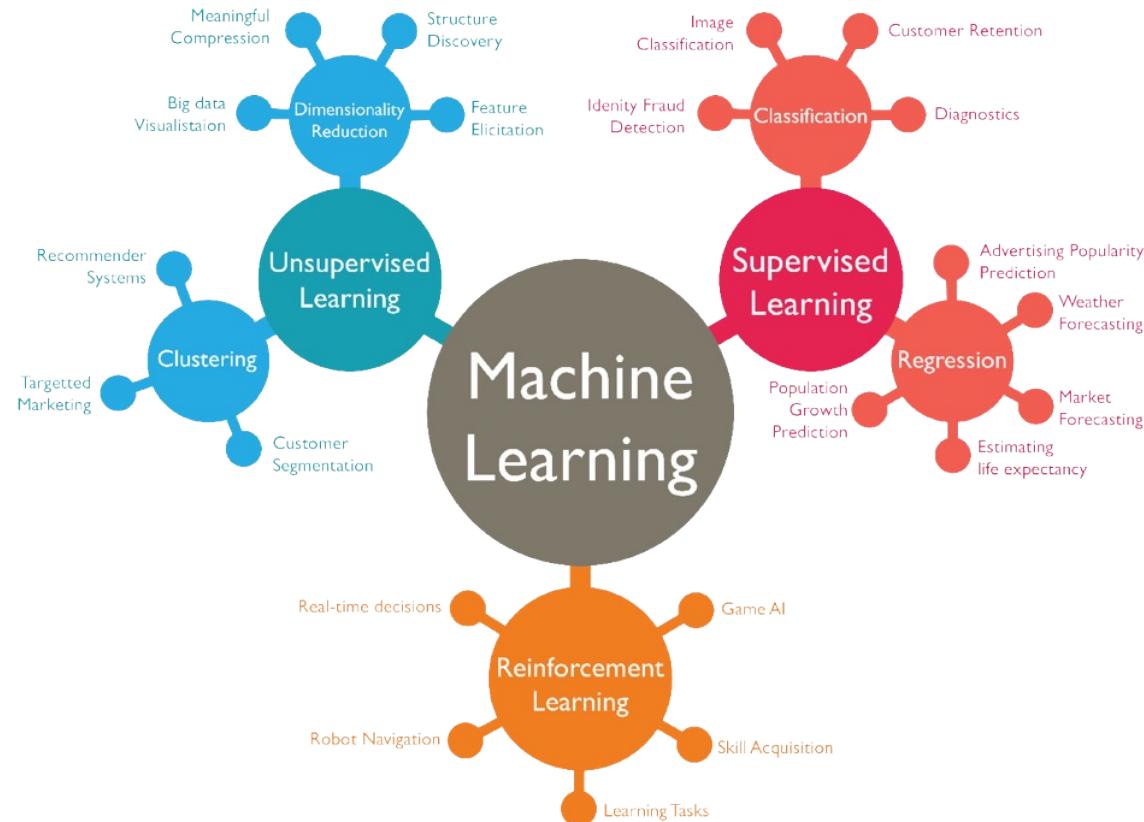
def produce_a_SN(star_mass):
    if star_mass > 8:
        return(True)
    else:
        return(False)

a = 10
print(produce_a_SN(a))
```

True

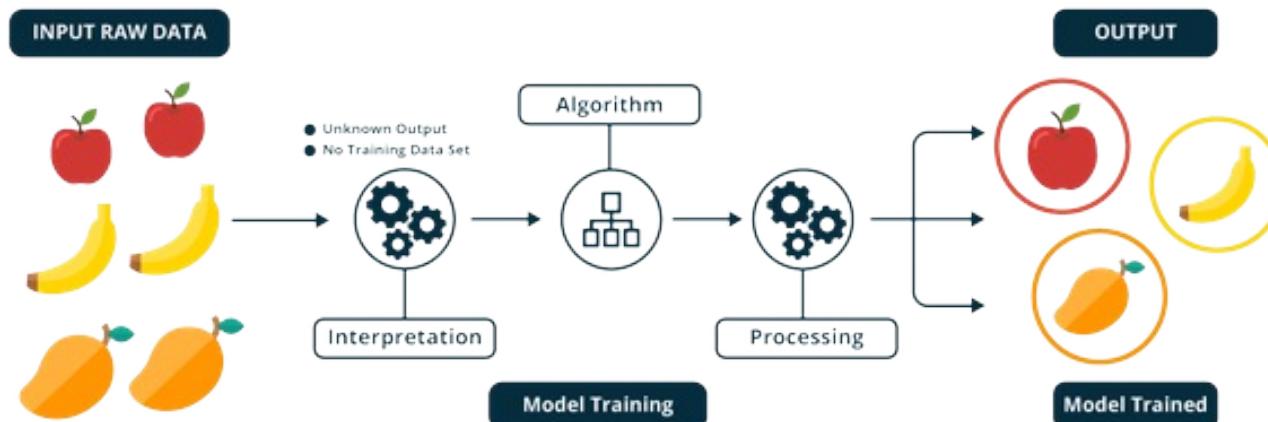


Técnicas de ML



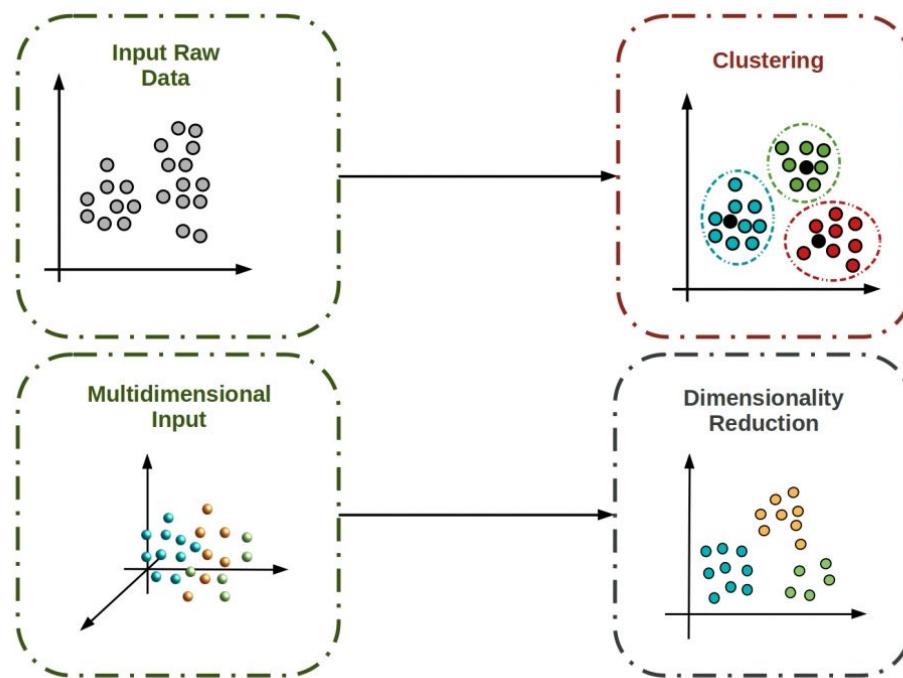
Técnicas de ML: aprendizaje supervisado

Input: etiquetas + atributos



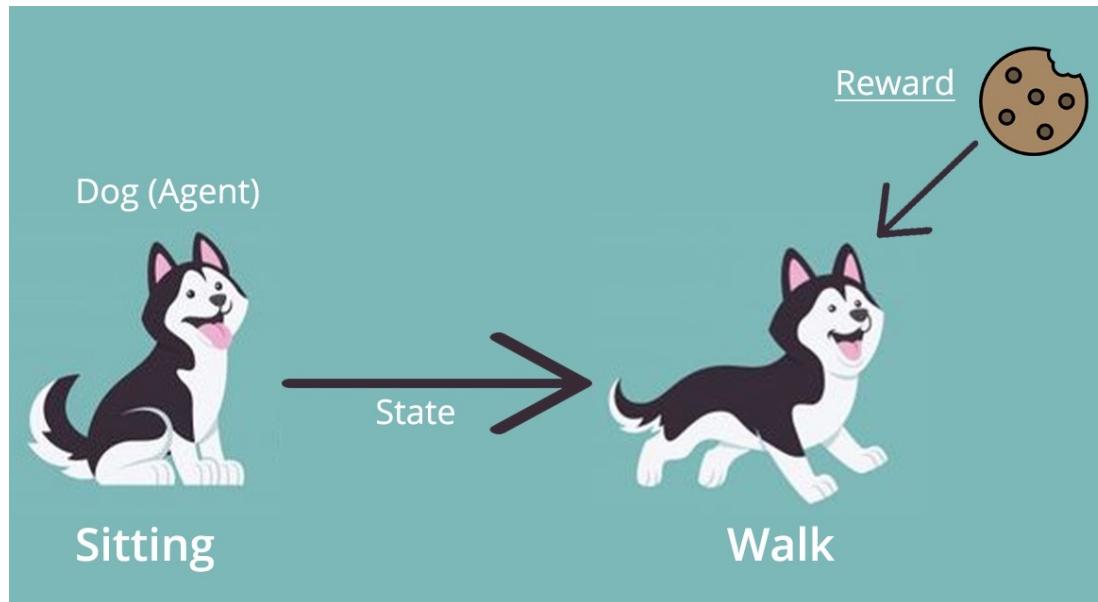
Técnicas de ML: aprendizaje no supervisado

Input: atributos



Técnicas de ML: aprendizaje por refuerzo

Input: datos dinámicos

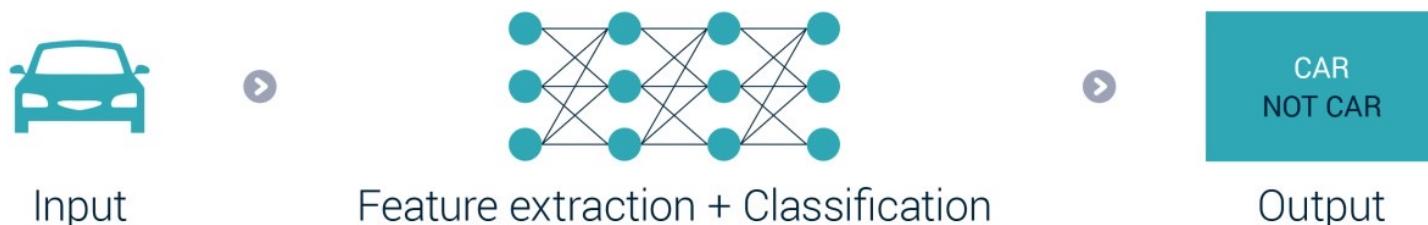


ML vs DL

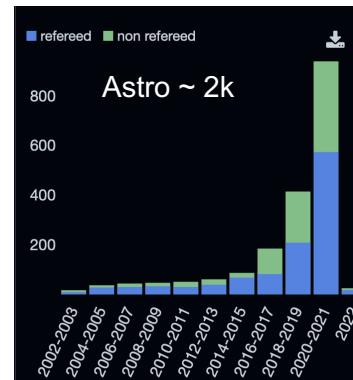
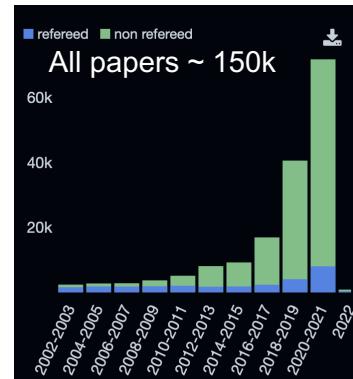
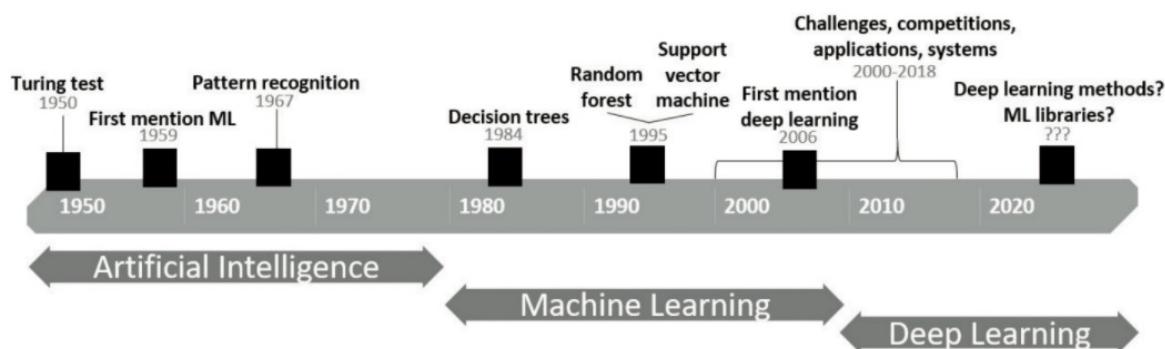
Machine Learning



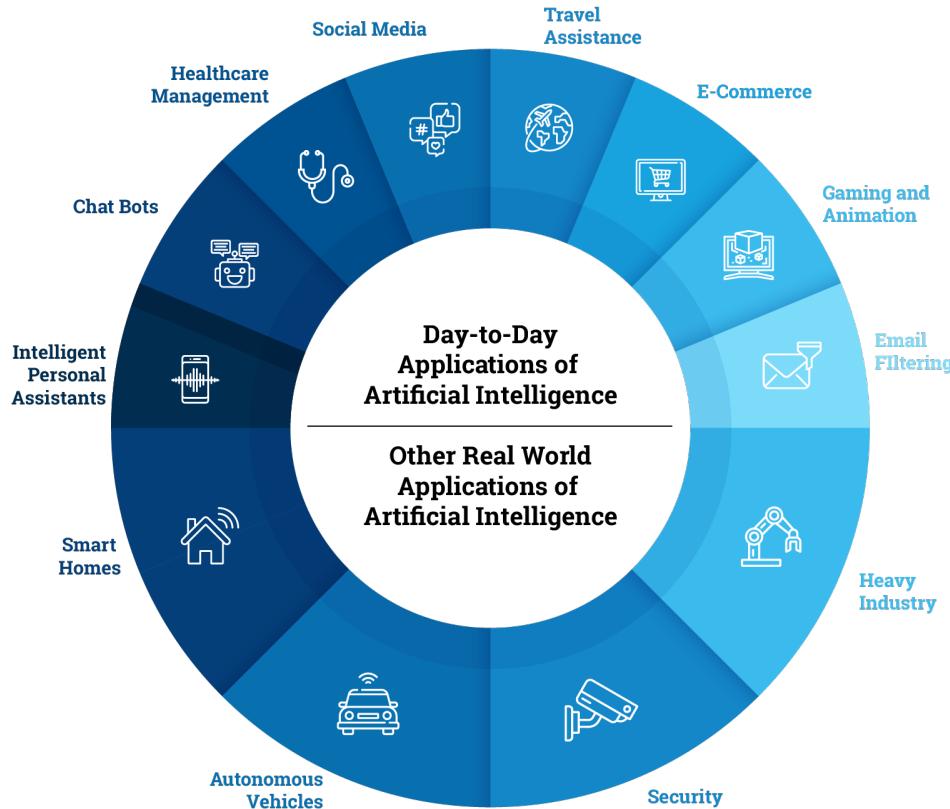
Deep Learning



IA en las últimas décadas

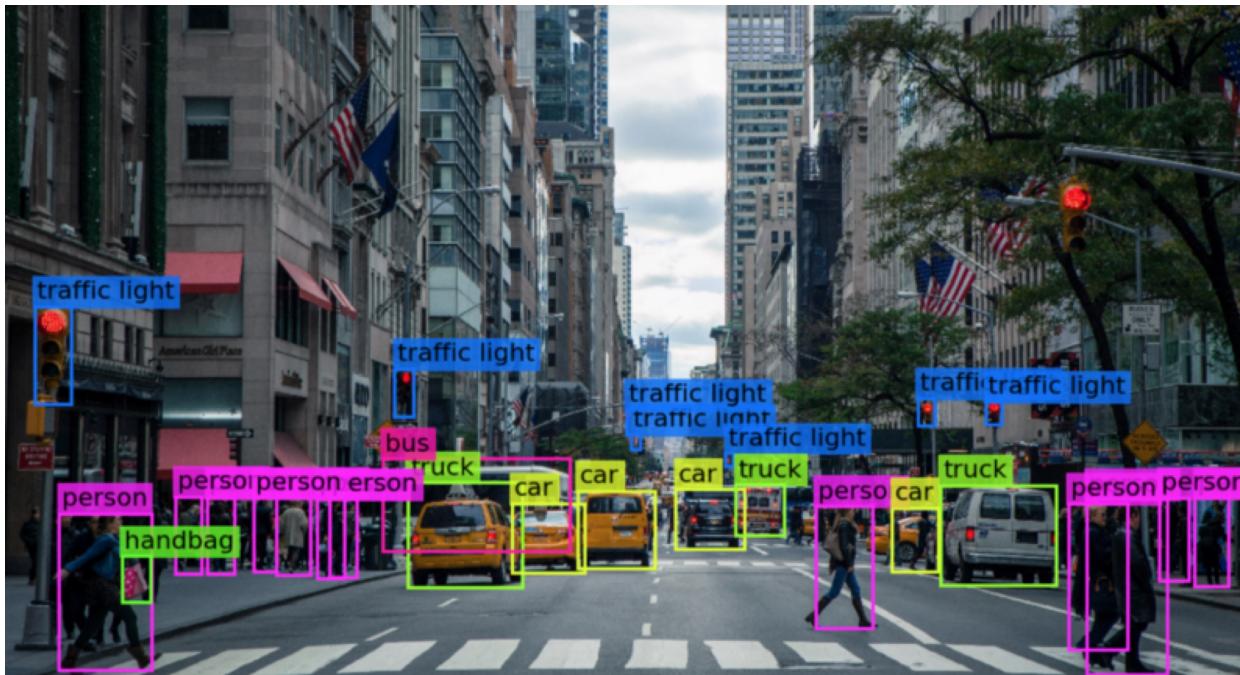


Aplicaciones en la vida real



Aplicaciones en la vida real

Computer Vision



Aplicaciones en la vida real

Traducción de lenguajes

澳洲 是 与 北韩 有 邦交 的 少数 国家 之一



① Phrase Segmentation

澳洲 是 与 北韩 有 邦交 的 少数 国家 之一



② Phrase Translation

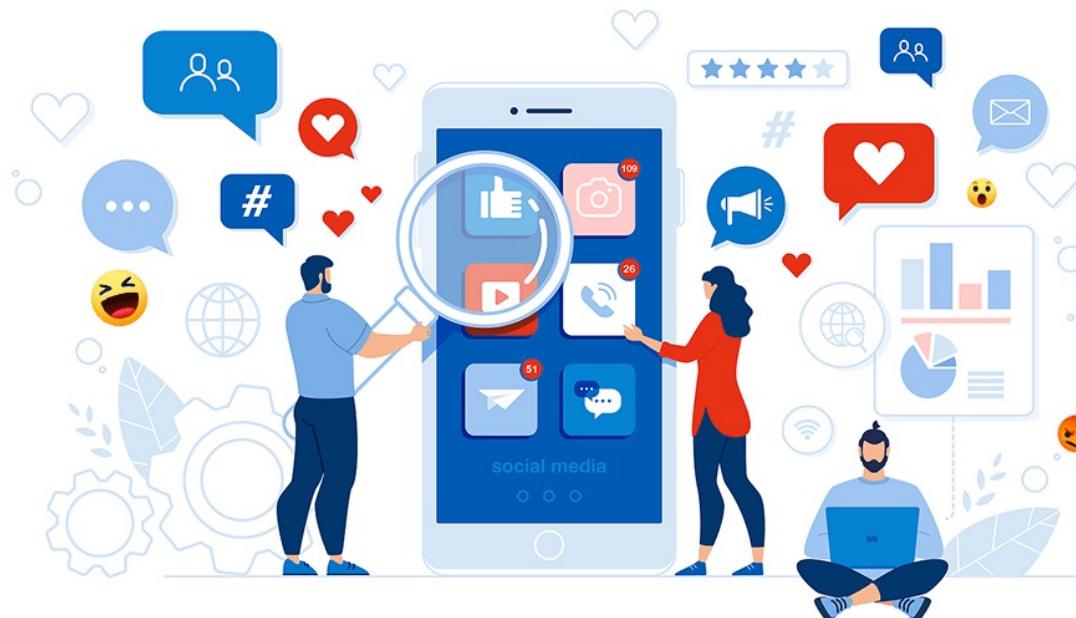
Australia is with North Korea have diplomatic relations one of the few countries that

A diagram showing the reordering of the translated phrases: 'Australia is', 'one of the few countries that', 'have diplomatic relations', and 'with North Korea'.

③ Phrase Reordering

Aplicaciones en la vida real

Redes sociales



Aplicaciones en la vida real

Asistentes virtuales

“Hey Siri”



“Hey Cortana”



“Alexa”



“OK Google”



“Hi Bixby”



2011



2014



2014



2016



2017

Aplicaciones en la vida real

Self Driving Vehicles





Aplicaciones de AI en Astronomía



Aplicaciones de AI en Astronomía

Classification of galaxy morphologies

Galaxy Zoo Challenge (2014)

Top solutions used CNNs

Winner: Rotation-invariant convolutional Neural Networks (Dieleman et al, 2015., MNRAS, 450, 2)

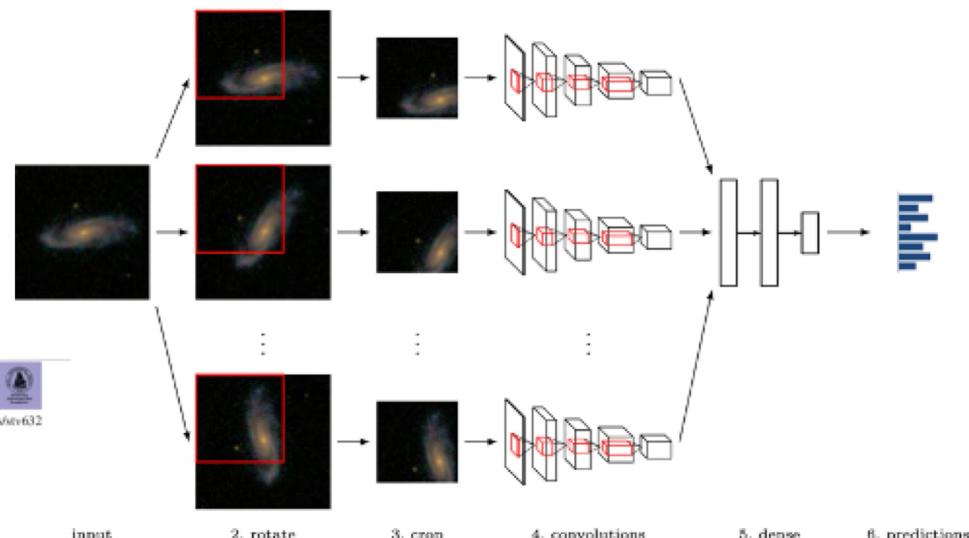
Monthly Notices
of the
ROYAL ASTRONOMICAL SOCIETY
MNRAS 450, 1441–1459 (2015)

Rotation-invariant convolutional neural networks for galaxy morphology prediction

Sander Dieleman,¹★ Kyle W. Willett²★ and Joni Dambre¹

¹Electronics and Information Systems department, Ghent University, Sint-Pietersnieuwstraat 41, B-9000 Ghent, Belgium

²School of Physics and Astronomy, University of Minnesota, 116 Church St SE, Minneapolis, MN 55455, USA

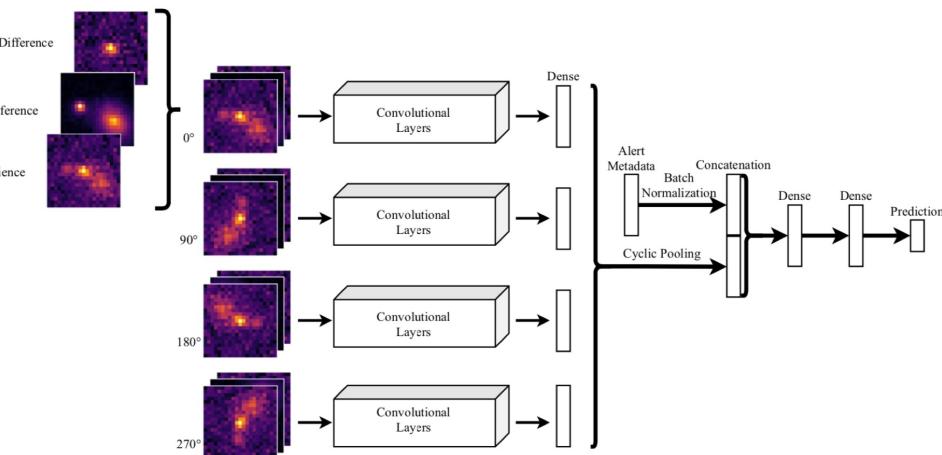
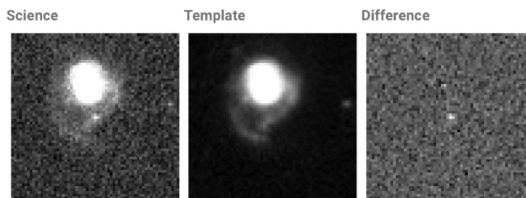


Aplicaciones de AI en Astronomía

Classification of variable and transients from images

Alert Classification for the ALeRCE Broker System: The Real-time Stamp Classifier

R. CARRASCO-DAVIS  ^{1,2,*} E. REYES  ^{1,2,*} C. VALENZUELA  ^{3,1,4,5} F. FÖRSTER  ^{3,1,6} P. A. ESTÉVEZ  ^{1,2}
G. PIGNATA  ^{7,1} F. E. BAUER  ^{8,1,9} I. REYES  ^{1,3,2} P. SÁNCHEZ-SÁEZ  ^{1,10,8,4} G. CABRERA-VIVES  ^{11,1}
S. EYHERAMENDY  ^{4,1} M. CATELAN  ^{8,1} J. ÁRREDONDO, ¹ E. CASTILLO-NAVARRETE, ^{3,1} D. RODRÍGUEZ-MANCINI, ^{1,11}
D. RUZ-MIERES  ^{3,1,12} A. MOYA  ^{1,3} L. SABATINI-GACITÚA, ^{1,3} C. SEPÚLVEDA-COBÓ, ^{1,3} A. A. MAHABAL, ^{13,14}
J. SILVA-FARFÁN, ⁶ E. CAMACHO-IÑIGUEZ, ⁸ AND L. GALBANY¹⁵



Aplicaciones de AI en Astronomía

Anomaly detection of SNe light curves

THE ASTROPHYSICAL JOURNAL SUPPLEMENT SERIES, 255:24 (11pp), 2021 August
 © 2021. The American Astronomical Society. All rights reserved.

<https://doi.org/10.3847/1538-4365/ac0893>



A Deep-learning Approach for Live Anomaly Detection of Extragalactic Transients

V. Ashley Villar^{1,2,3,4,10}, Miles Cranmer⁵, Edo Berger^{6,7}, Gabriella Contardo⁸, Shirley Ho⁸, Griffin Hosseinzadeh^{6,7}, and Joshua Yao-Yu Lin⁹

¹ Department of Astronomy, Columbia University, New York, NY 10027, USA; vav2110@columbia.edu

² Department of Astronomy & Astrophysics, The Pennsylvania State University, University Park, PA 16802, USA

³ Institute for Computational & Data Sciences, The Pennsylvania State University, University Park, PA, USA

⁴ Institute for Gravitation and the Cosmos, The Pennsylvania State University, University Park, PA 16802, USA

⁵ Princeton University, Princeton, NJ 08544, USA

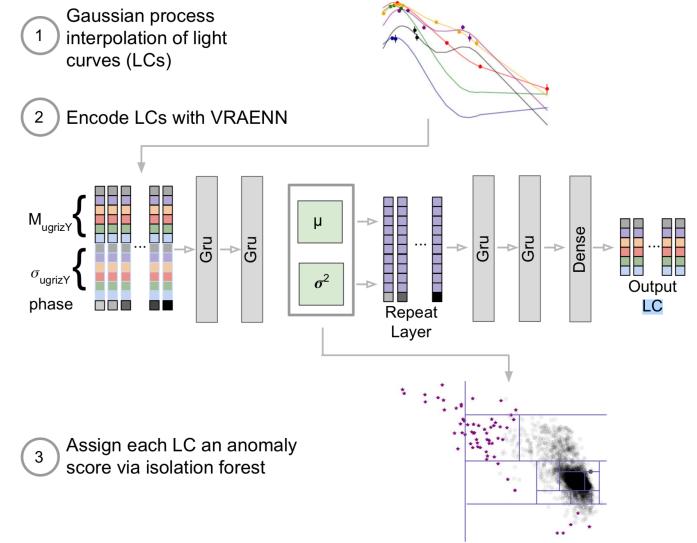
⁶ Center for Astrophysics | Harvard & Smithsonian, 60 Garden Street, Cambridge, MA 02138-1516, USA

⁷ The NSF AI Institute for Artificial Intelligence and Fundamental Interactions, 77 Massachusetts Avenue, Cambridge, MA 02139, USA

⁸ Flatiron Institute, New York City, NY 10010, USA

⁹ University of Illinois at Urbana-Champaign, Urbana, IL 61820, USA

Received 2021 March 17; revised 2021 June 2; accepted 2021 June 3; published 2021 August 10



Aplicaciones de AI en Astronomía

Modeling and forecasting of AGN light curves

THE ASTROPHYSICAL JOURNAL, 903:54 (17pp), 2020 November 1

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<https://doi.org/10.3847/1538-4357/abb9a9>



Deep Modeling of Quasar Variability

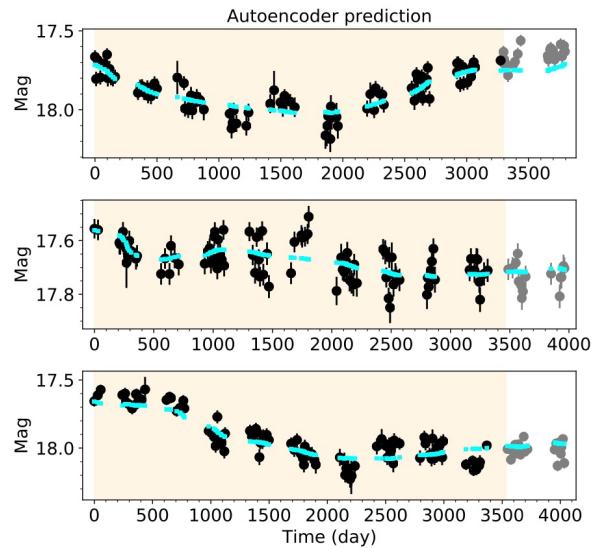
Yutaro Tachibana (優太朗橘)^{1,2,4}, Matthew J. Graham^{2,4}, Nobuyuki Kawai¹, S. G. Djorgovski², Andrew J. Drake², Ashish A. Mahabal², and Daniel Stern³

¹ Department of Physics, Tokyo Institute of Technology, 2-12-1 Ookayama, Meguro-ku, Tokyo 152-8551, Japan

² Department of Physics, Math, and Astronomy, California Institute of Technology, Pasadena, CA, 91125, USA; mjg@caltech.edu

³ Jet Propulsion Laboratory, California Institute of Technology, 4800 Oak Grove Drive, Pasadena, CA 91109, USA

Received 2020 March 2; revised 2020 September 2; accepted 2020 September 16; published 2020 November 2



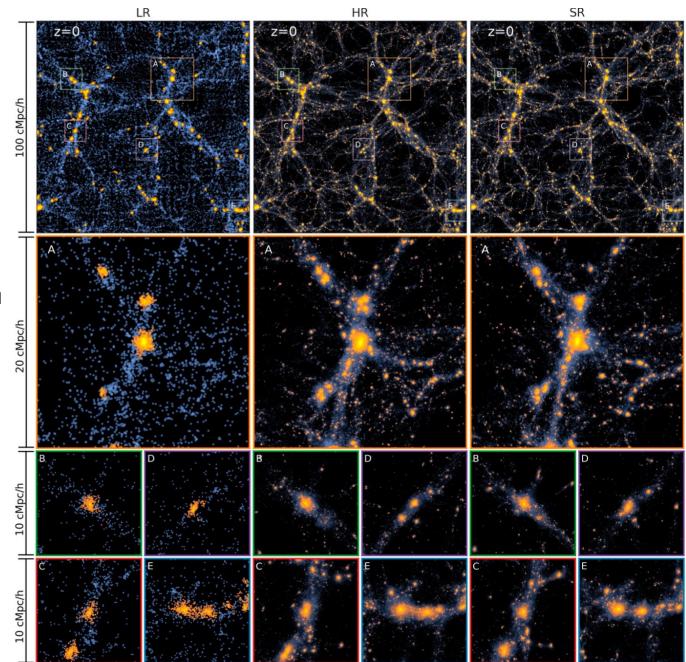
Aplicaciones de AI en Astronomía

Acceleration of Cosmological Simulations

AI-assisted superresolution cosmological simulations

Yin Li^{a,b,1}, Yueying Ni^{c,d,1}, Rupert A. C. Croft^{c,d}, Tiziana Di Matteo^{c,d}, Simeon Bird^e, and Yu Feng^{f,g}

^aCenter for Computational Astrophysics, Flatiron Institute, Simons Foundation, New York, NY 10010; ^bCenter for Computational Mathematics, Flatiron Institute, Simons Foundation, New York, NY 10010; ^cMcWilliams Center for Cosmology, Department of Physics, Carnegie Mellon University, Pittsburgh, PA 15213; ^dNSF AI Planning Institute for Physics of the Future, Carnegie Mellon University, Pittsburgh, PA 15213; ^eDepartment of Physics and Astronomy, University of California Riverside, Riverside, CA 92521; ^fBerkeley Center for Cosmological Physics, University of California Berkeley, Berkeley, CA 94720; and ^gDepartment of Physics, University of California Berkeley, Berkeley, CA 94720



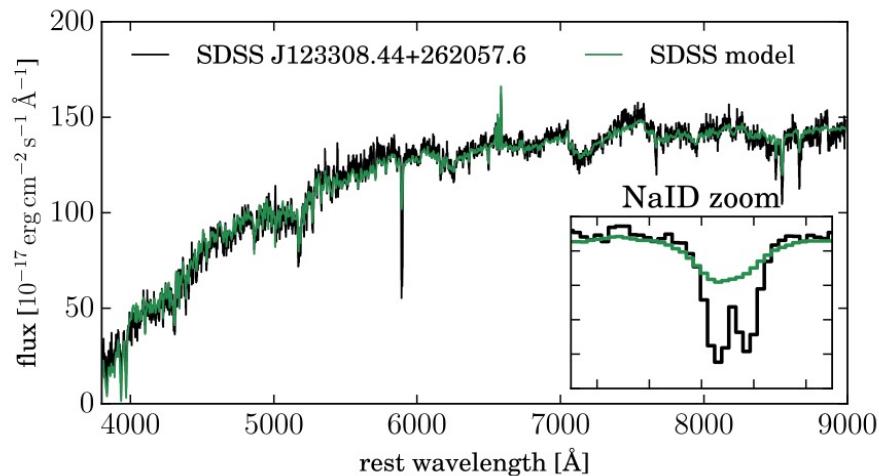
Aplicaciones de AI en Astronomía

Anomaly detection of SDSS spectra

The weirdest SDSS galaxies: results from an outlier detection algorithm

Dalya Baron[★] and Dovi Poznanski

School of Physics and Astronomy, Tel-Aviv University, Tel Aviv 69978, Israel



Aplicaciones de AI en Astronomía

A neural network to identify good astrometric solutions in Gaia

A classifier for spurious astrometric solutions in *Gaia* eDR3

Jan Rybizki ^{○,1*}, Gregory M. Green, ^{1,†} Hans-Walter Rix,¹ Kareem El-Badry,¹ Markus Demleitner,² Eleonora Zari,¹ Andrzej Udalski,³ Richard L. Smart ^{○,4} and Andrew Gould^{1,5}

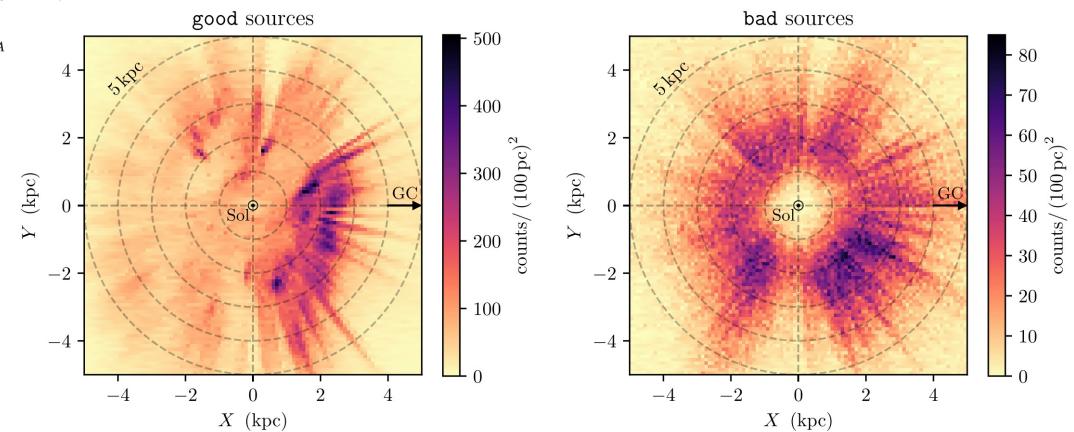
¹Max Planck Institute for Astronomy, Königstuhl 17, D-69117 Heidelberg, Germany

²Astronomisches Rechen-Institut, Zentrum für Astronomie der Universität Heidelberg, Mönchhofstrasse 12-14, D-69120 Heidelberg, Germany

³Astronomical Observatory, University of Warsaw, Al. Ujazdowskie 4, PL-00478 Warsaw, Poland

⁴INAF – Osservatorio Astrofisico di Torino, via Osservatorio 20, I-10025 Pino Torinese, TO, Italy

⁵Department of Astronomy, Ohio State University, 4055 McPherson Laboratory, 140 West 18th Avenue, Columbus, OH 43210, USA



Aplicaciones de AI en Astronomía

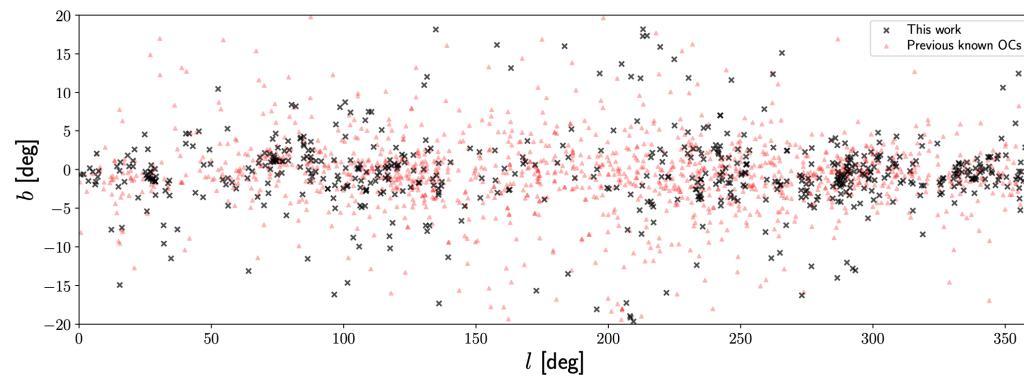
Clustering algorithms + neural network to find new open clusters

Hunting for open clusters in Gaia DR2: 582 new open clusters in the Galactic disc*

A. Castro-Ginard¹, C. Jordi¹, X. Luri¹, J. Álvarez Cid-Fuentes², L. Casamiquela³, F. Anders¹, T. Cantat-Gaudin¹,
M. Monguió¹, L. Balaguer-Núñez¹, S. Solà², and R. M. Badia²

¹ Dept. Física Quàntica i Astrofísica, Institut de Ciències del Cosmos (ICCUB), Universitat de Barcelona (IEEC-UB),
Martí i Franquès 1, 08028 Barcelona, Spain
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² Barcelona Supercomputing Center (BSC), Barcelona, Spain
³ Laboratoire d'Astrophysique de Bordeaux, Univ. Bordeaux, CNRS, B18N, allée Geoffroy, Saint-Hilaire 33615, Pessac, France



Aplicaciones de AI en Astronomía

Transfer learning to find stars accreted onto the Milky Way

Cataloguing accreted stars within Gaia DR2 using deep learning

B. Ostdiek^{1,*}, L. Necib², T. Cohen¹, M. Freytsis^{3,4}, M. Lisanti⁵, S. Garrison-Kimmel⁶, A. Wetzel⁷,
R. E. Sanderson^{8,9}, and P. F. Hopkins⁶

¹ Institute of Theoretical Science, Department of Physics, University of Oregon, Eugene, OR 97403, USA
e-mail: bostdiek@g.harvard.edu

² Walter Burke Institute for Theoretical Physics, California Institute of Technology, Pasadena, CA 91125, USA
e-mail: lneceb@caltech.edu

³ Raymond and Beverly Sackler School of Physics and Astronomy, Tel Aviv University, Tel Aviv 69978, Israel
⁴ School of Natural Sciences, Institute for Advanced Study, Princeton, NJ 08540, USA

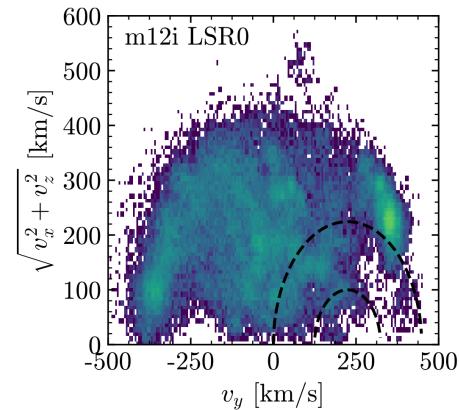
⁵ Department of Physics, Princeton University, Princeton, NJ 08544, USA

⁶ TAPIR, California Institute of Technology, Pasadena, CA 91125, USA

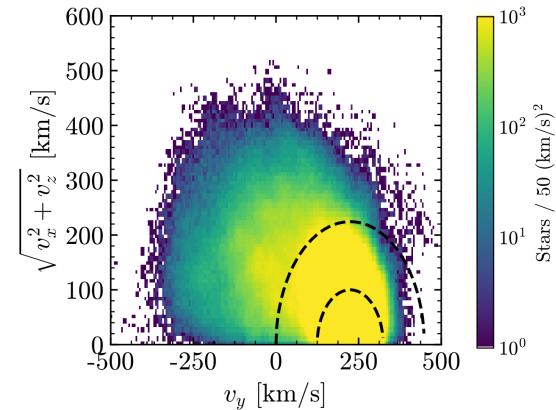
⁷ Department of Physics, University of California, Davis, CA 95616, USA

⁸ Department of Physics and Astronomy, University of Pennsylvania, Philadelphia, PA 19104, USA

⁹ Center for Computational Astrophysics, Flatiron Institute, New York, NY 10010, USA



(a) Accreted stars.



(b) in situ stars.

Aplicaciones de AI en Astronomía

Finding strong gravitational lenses in the Kilo Degree Survey with Convolutional Neural Networks

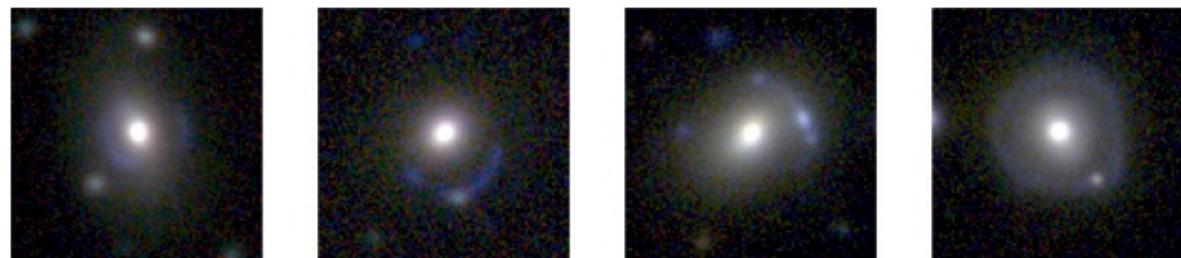
C. E. Petrillo,¹ C. Tortora,¹ S. Chatterjee,¹ G. Vernardos,¹ L. V. E. Koopmans,¹ G. Verdoes Kleijn,¹ N. R. Napolitano,² G. Covone,³ P. Schneider,⁴ A. Grado² and J. McFarland¹

¹*Kapteyn Astronomical Institute, University of Groningen, Postbus 800, NL-9700 AV Groningen, the Netherlands*

²*INAF – Osservatorio Astronomico di Capodimonte, Salita Moiariello, 16, I-80131 Napoli, Italy*

³*Dipartimento di Scienze Fisiche, Università di Napoli Federico II, Compl. Univ. Monte S. Angelo, I-80126 Napoli, Italy*

⁴*Argelander-Institut für Astronomie, Auf dem Hügel 71, D-53121 Bonn, Germany*





**ALeRCE es una iniciativa chilena para crear un
“community Broker” para ZTF, LSST y otros
grandes telescopios**

Futuros ecosistemas para “time domain astronomy”

Survey telescopes



Alert brokers/TOMs

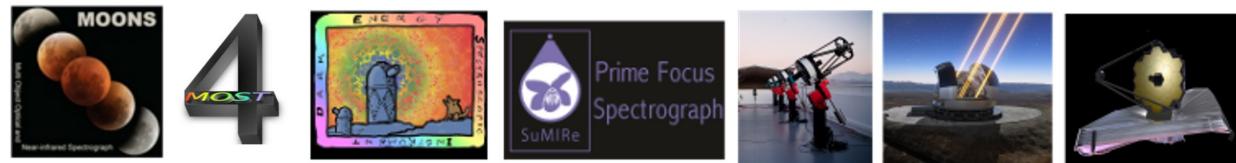


ALeRCE
Automatic Learning for the
Rapid Classification of Events

+ other brokers/TOMs

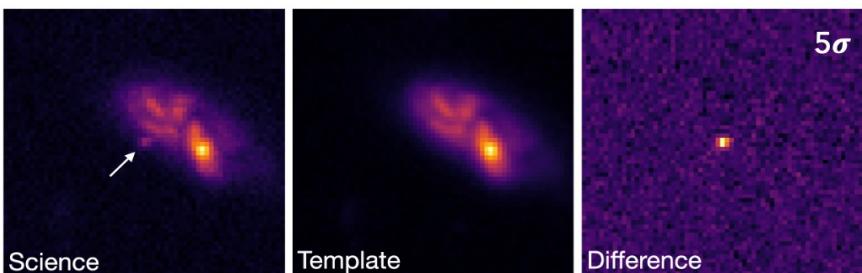


Follow up telescopes



Brokers

Los brokers son sistemas que procesan alertas astronómicas.



ZTF

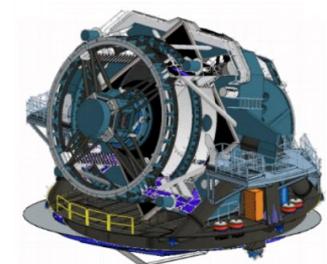
2018-2023

1.4 TB per night

~1 billion objects

~1 trillion measurements

~1 million alerts per night



LSST

2022-2032

10x

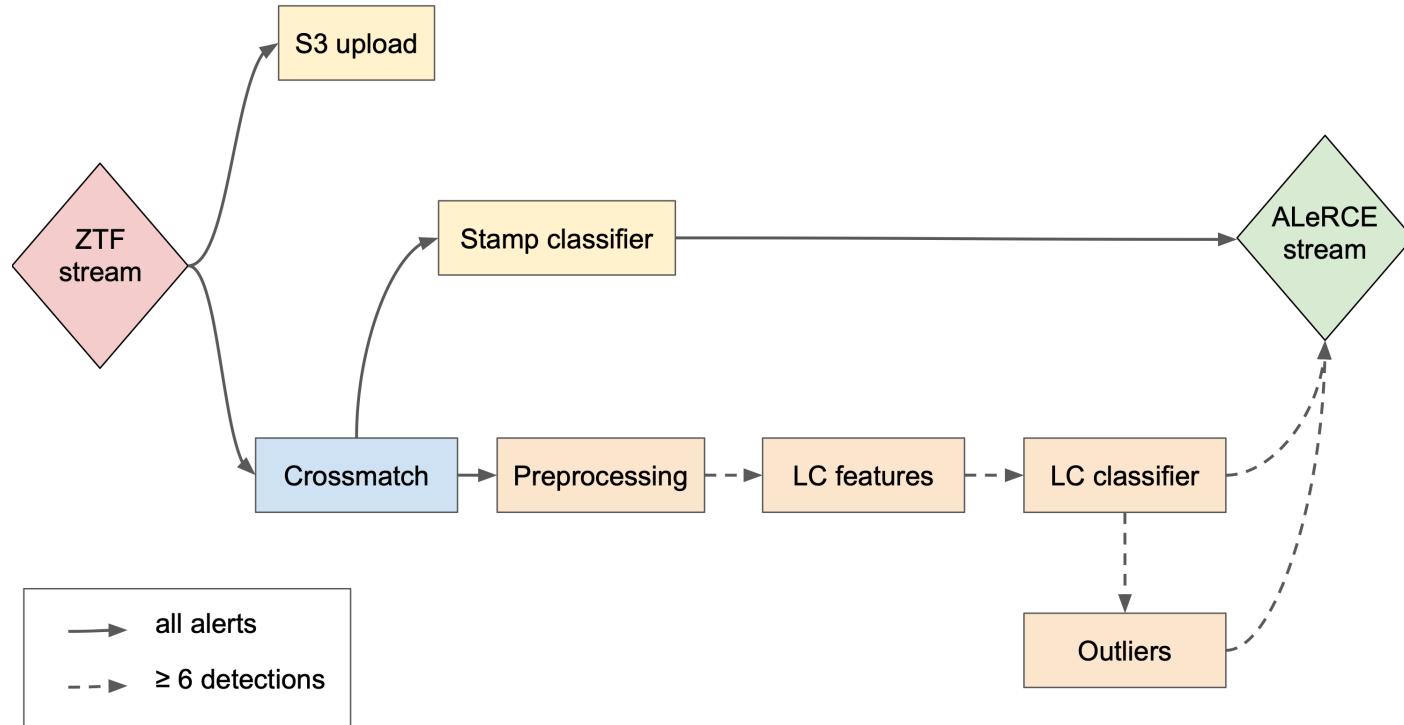
15 TB per night

~37 billion objects

~7 trillion measurements

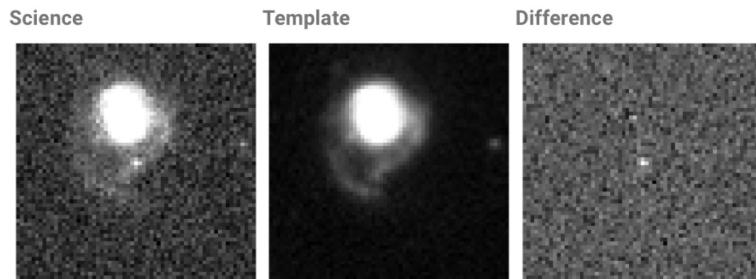
~10 million alerts per night

Pipeline de ALeRCE

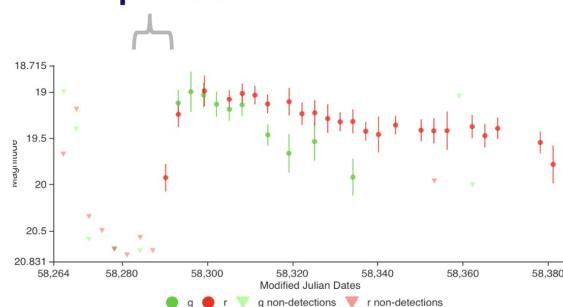


Clasificadores de ALeRCE

Convolutional Neural Network (using 1st detection stamp)

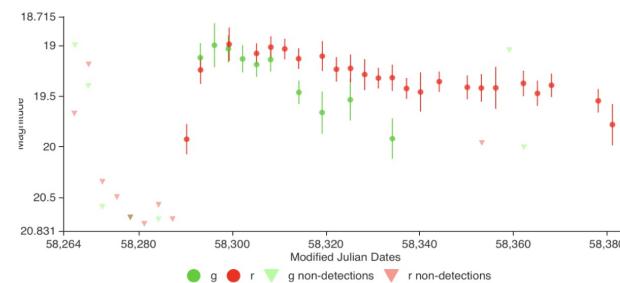


Stamp classifier



Balanced Random Forest (for objects with 6 or more alerts in a given band)

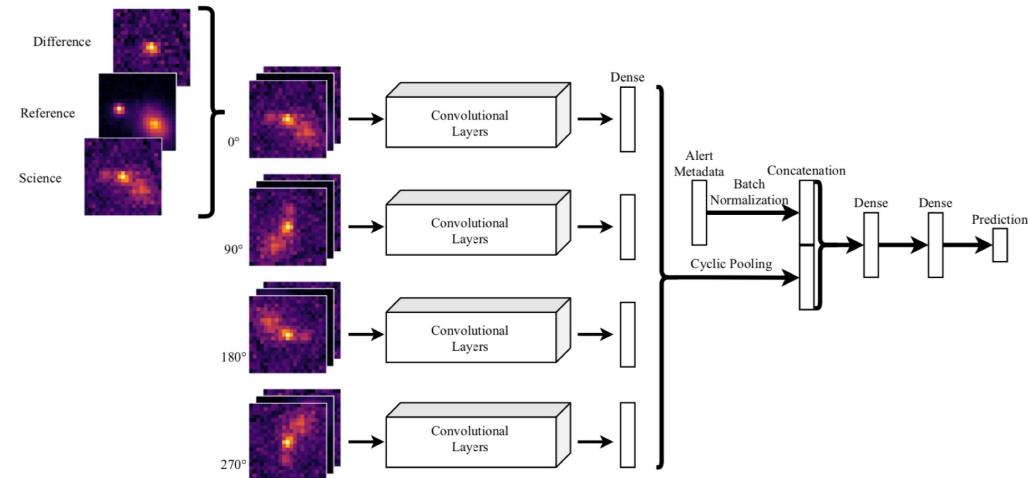
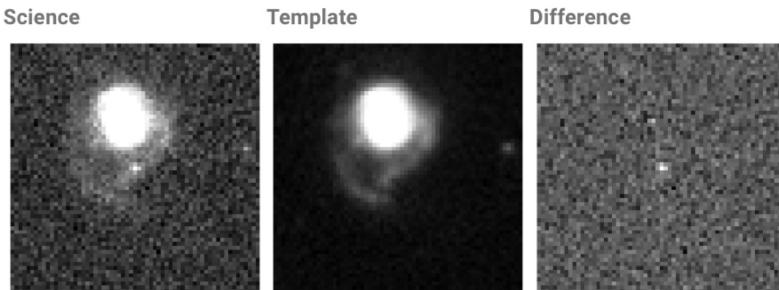
Light curve classifier



Sánchez-Sáez et al. 2021, AJ, 161,141

Clasificador de imágenes de ALeRCE

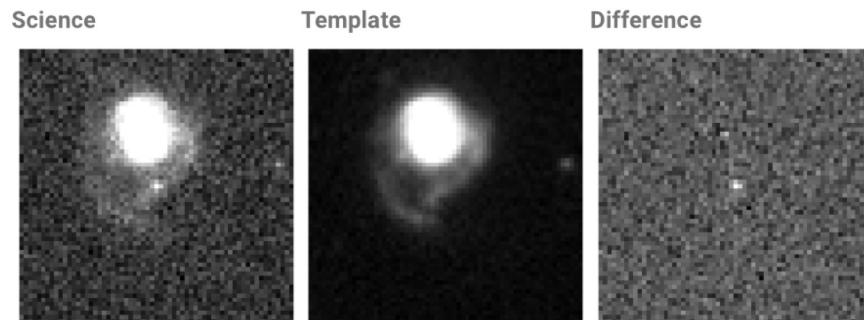
Convolutional Neural Network
(using the first stamp)



Carrasco-Davis et al. 2021, AJ, 162, 231

Clasificador de imágenes de ALeRCE

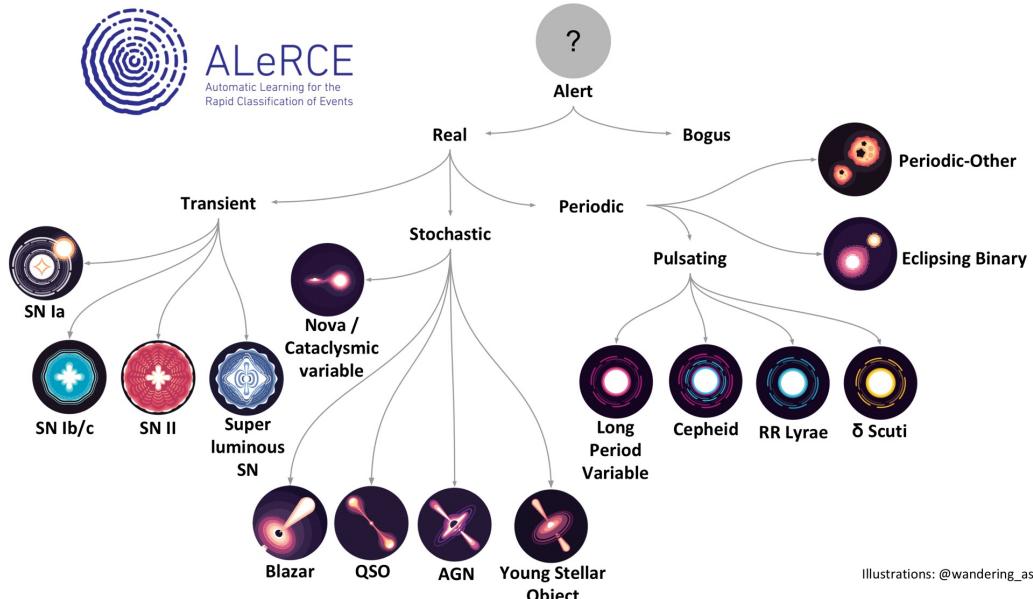
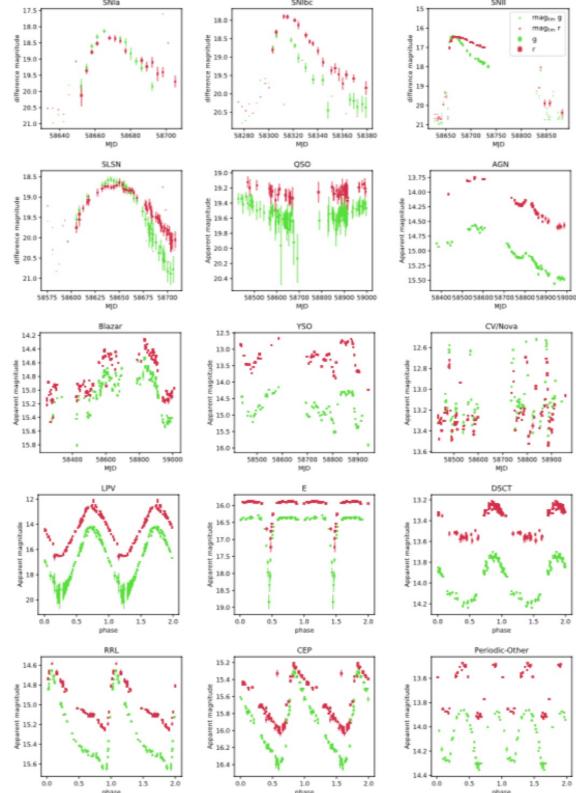
Convolutional Neural Network
(using the first stamp)



		Predicted label				
		AGN	SN	VS	asteroid	bogus
True label	AGN	0.93±0.02	0.01±0.01	0.05±0.01	0±0.0	0±0.0
	SN	0.01±0.0	0.88±0.03	0.01±0.0	0.07±0.01	0.03±0.01
VS	AGN	0.03±0.01	0±0.0	0.97±0.01	0±0.0	0±0.0
	SN	0.01±0.0	0.01±0.0	0.01±0.0	0.98±0.01	0±0.0
asteroid	AGN	0±0.0	0.02±0.01	0.01±0.0	0.01±0.0	0.96±0.01
	SN	0.01±0.0	0.01±0.0	0.01±0.0	0.01±0.0	0±0.0

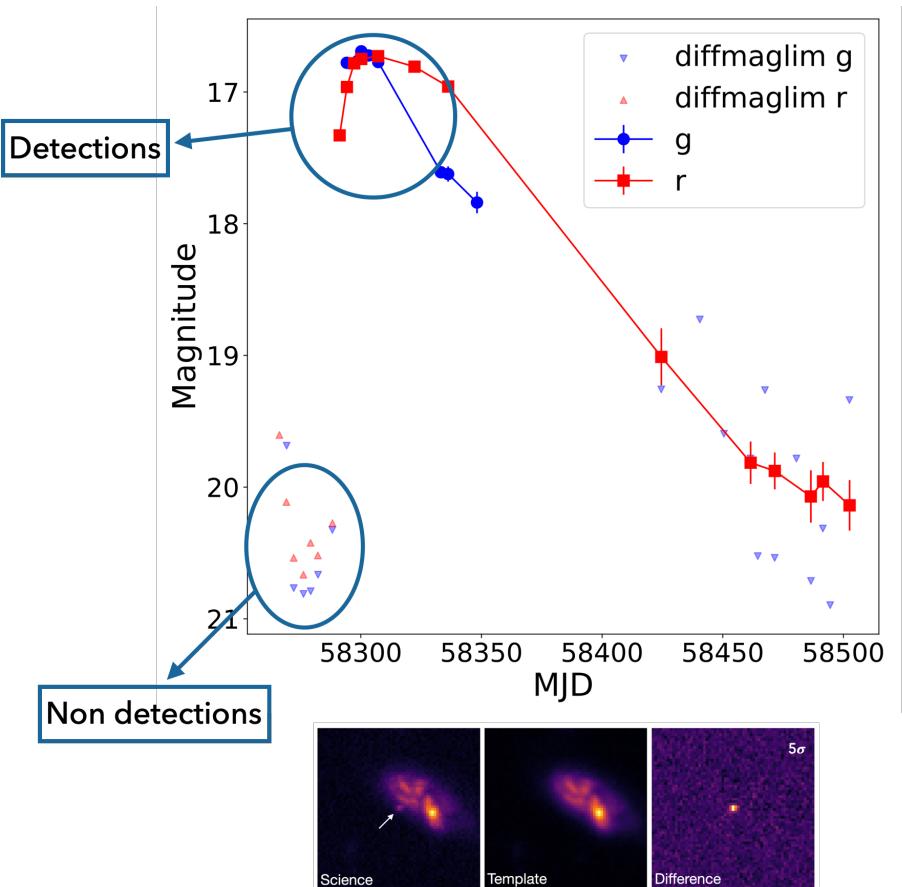
Carrasco-Davis et al. 2021, AJ, 162, 231

Clasificador de curvas de luz de ALeRCE



Illustrations: @wandering_astro

Clasificador de curvas de luz de ALeRCE



Sánchez-Sáez et al. 2021, AJ, 161, 141

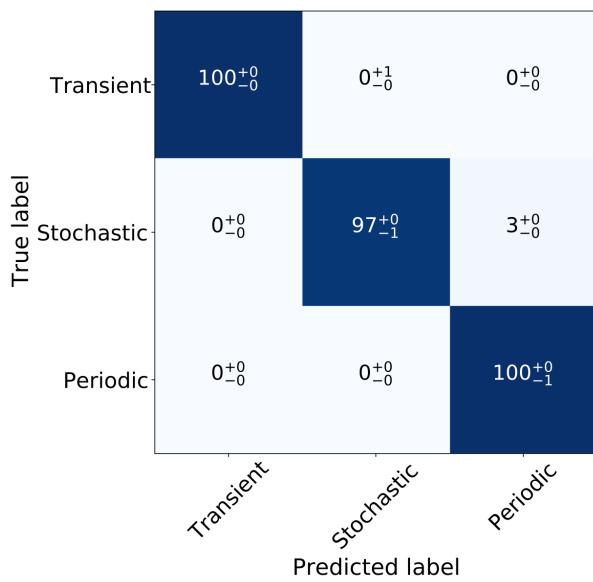
152 features in total:

- 1) Colors from AllWISE and ZTF (8 in total)
- 2) Detection features (for g and r ZTF bands, 124 in total):
 - Supernova parametric model (SPM; adapted from Villar et al. 2019b)
 - Multiband period (adapted from Mondrik et al. 2015)
 - Irregular autoregressive model (IAR; Eyheramendy et al. 2018)
 - Mexican Hat Power Spectrum (MHPS; adapted from Arévalo et al. 2012)
- 3) Non-detection features (for g and r ZTF bands, 18 in total)
- 4) Features from ZTF metadata (galactic coordinates, sgscore1 from PanSTARRS1, and real-bogus)

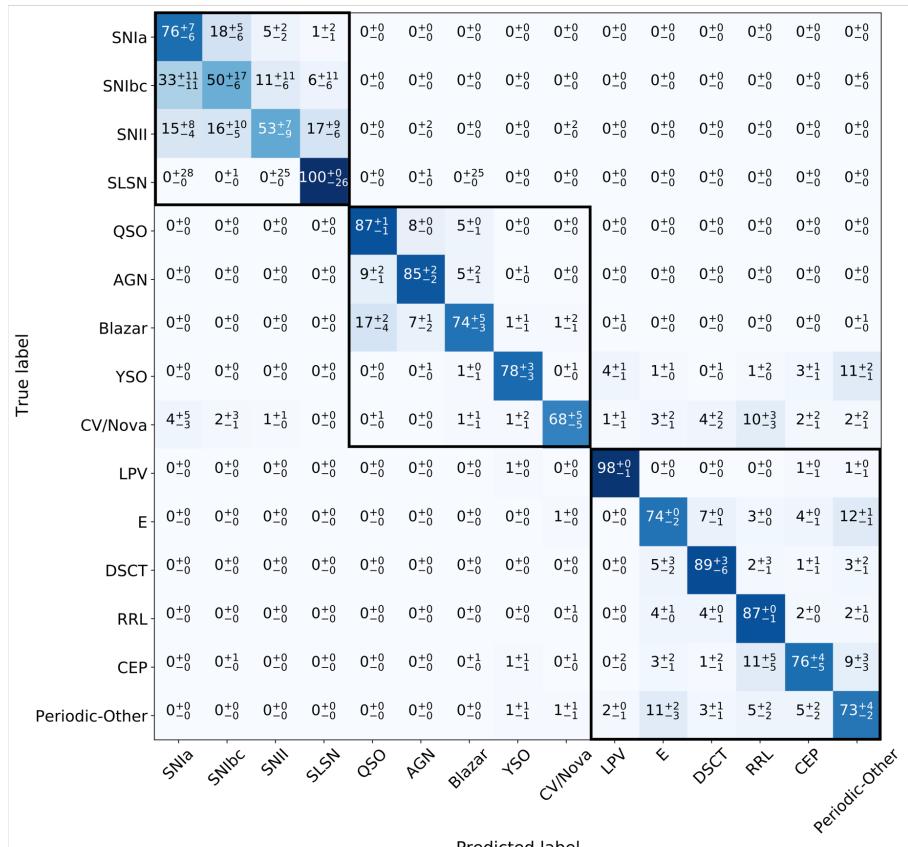


Clasificador de curvas de luz de ALeRCE

Sánchez-Sáez et al. 2021, AJ, 161,141



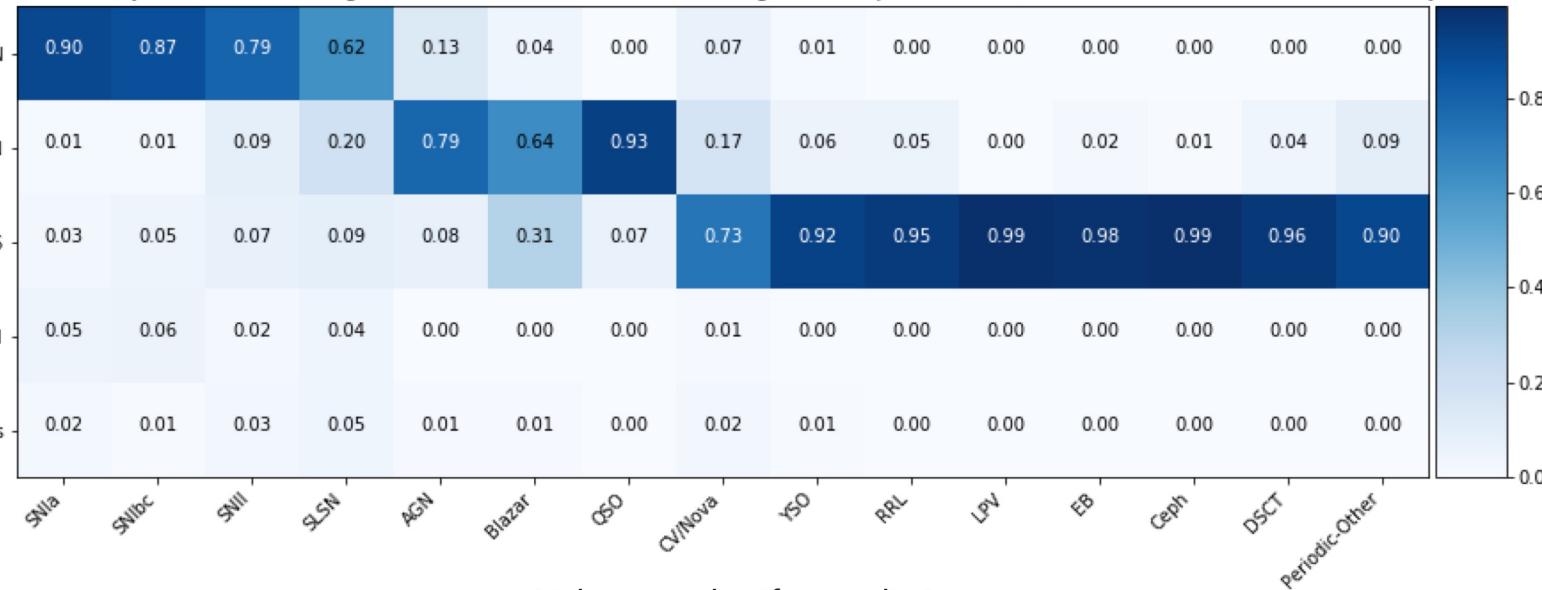
Classifier	Precision	Recall	F1-score
BHRF - top	0.96 ± 0.01	0.99 ± 0.01	0.97 ± 0.01
BHRF - bottom	0.57 ± 0.01	0.76 ± 0.02	0.59 ± 0.01



Comparación de clasificadores de ALeRCE

Fraction of objects classified into given LC class (columns) classified as given stamp class (rows). From a total of 186794 unlabeled objects.

Stamp classifier prediction



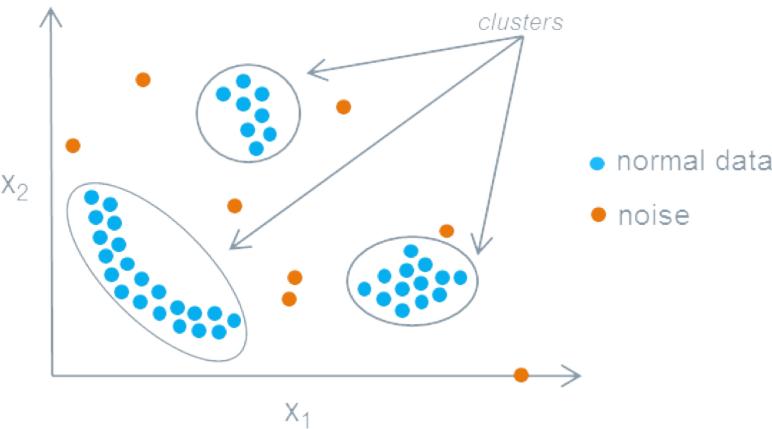
Light curve classifier prediction

Förster et al. 2021. 2021, AJ, 161, 242

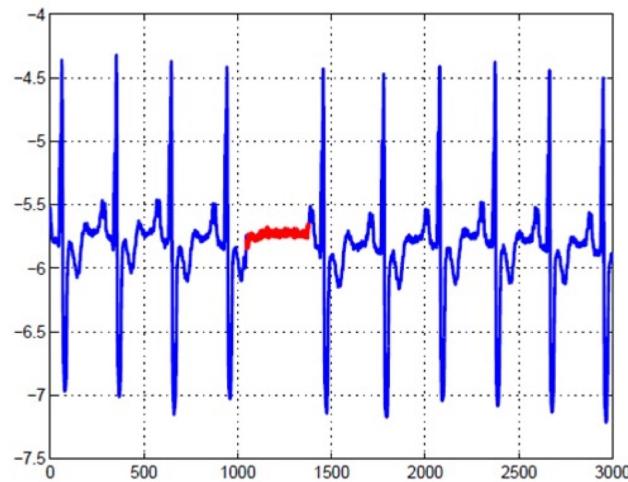
Detección de anomalías

Corresponde a la identificación de eventos raros u observaciones que difieren significativamente de la mayoría de los datos.

Out of Distribution AD: búsqueda de objetos inusuales dentro de los conjuntos de datos.

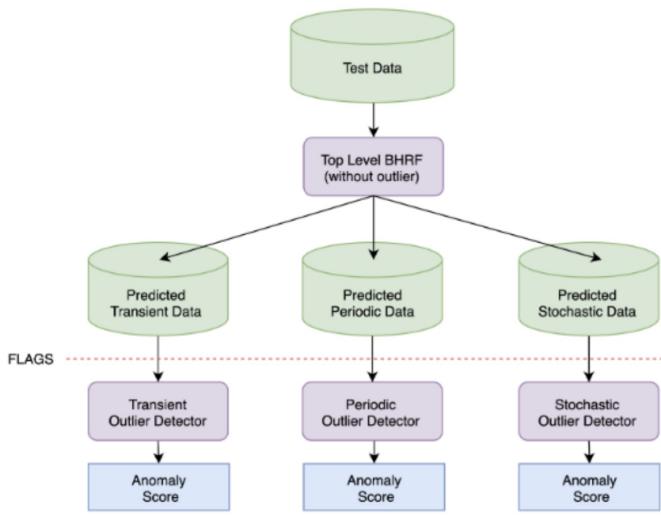


Contextual AD: búsqueda de objetos que de repente empiezan a presentar comportamientos inusuales.

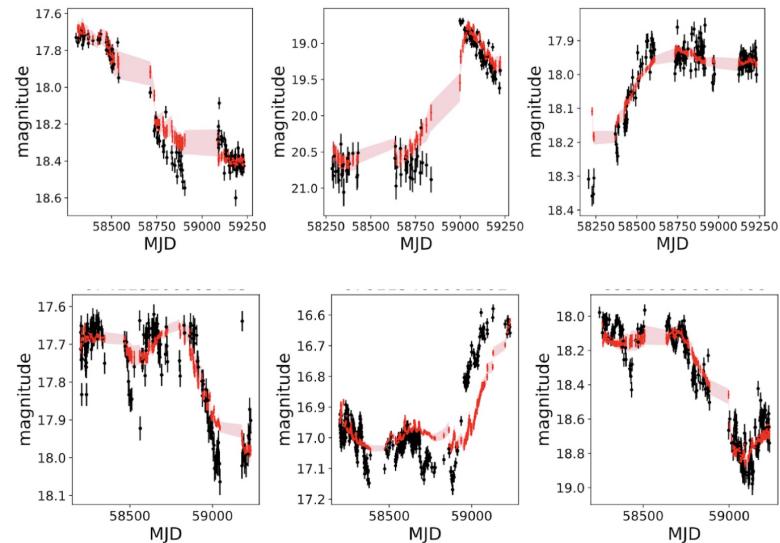


Detección de anomalías en ALeRCE

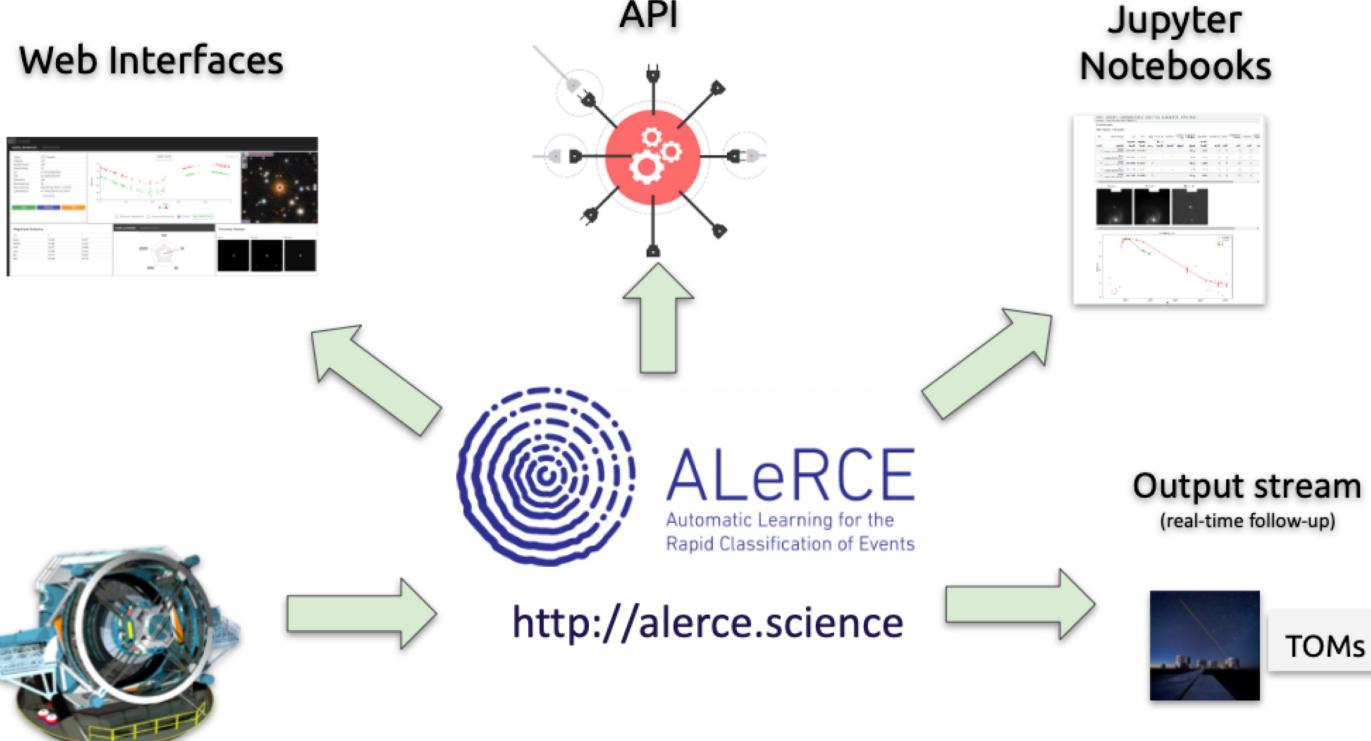
Out of Distribution AD: Detector de anomalías en curvas de alertas de ZTF (Pérez-Carrasco et al. in prep)



Contextual AD: Detector de anomalías en curvas de data release de ZTF (Sánchez-Sáez et al. 2021b, AJ, 162, 206)



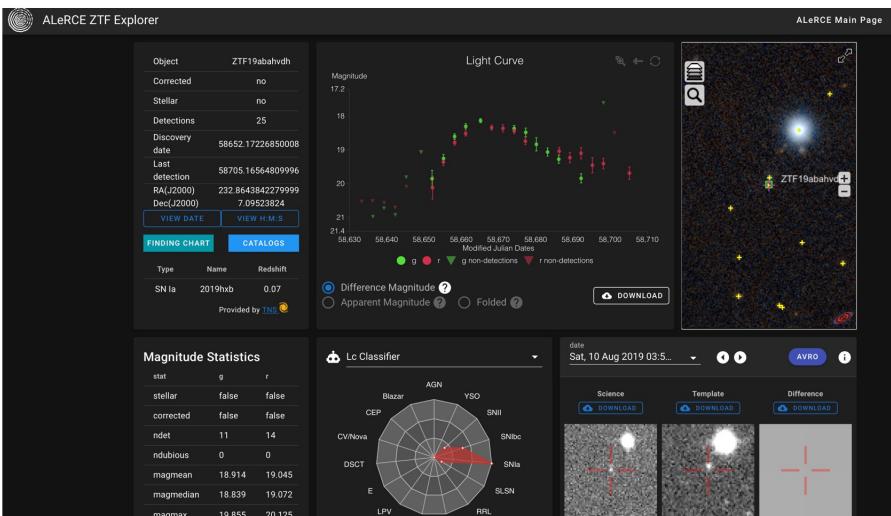
Servicios de ALeRCE



Servicios de ALeRCE

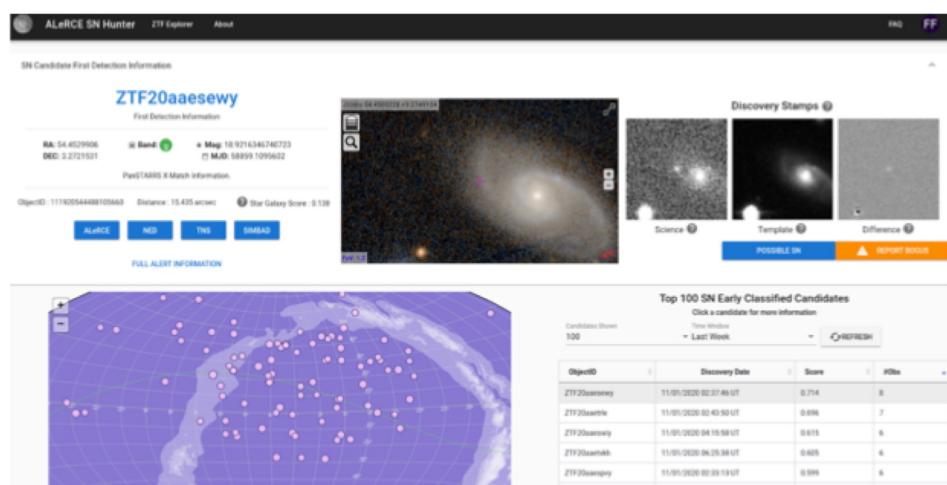
ZTF Explorer

<http://alerce.online>



SN Hunter

<http://snhunter.alerce.online>



Links

Homepage: <http://alerce.science>

Explorer: <https://alerce.online/>

SN Hunter: <https://snhunter.alerce.online/>

ALeRCE tutorial: videos, slides & notebooks.

Muchas gracias!!