

Foundation Models From Outside Astronomy, and Their Utility for Blazar Variability

Sasha (Alexander) Plavin

Black Hole Initiative Fellow, Harvard → Jansky Fellow, NRAO



Are **foundational time-series models** here already?

Sasha (Alexander) Plavin

Black Hole Initiative Fellow, Harvard → Jansky Fellow, NRAO



Inputs

Text (question)



Do we see a
full ring morphology
here?

Submit input
+ question
to VLM/LLM
(ChatGPT)

VLM/LLM



Generate
answer
(structured
output)

Answer (JSON)

```
{  
  "full_ring": true,  
  "confidence": 0.92,  
  "reason": "Bright ring  
surrounding central  
dark region."  
}
```

Images



Natural image



Astronomical image

Submit input
to VLM/LLM

VLM/LLM



Get
embedding



Regress /
Classify

Prediction



e.g.,
score = 0.93

or



e.g.,
full ring

how about time series?

Inputs

Text (question)



Do we see a
full ring morphology
here?

Submit input
+ question
to VLM/LLM
(ChatGPT)

VLM/LLM



Generate
answer
(structured
output)

Answer (JSON)

```
{  
  "full_ring": true,  
  "confidence": 0.92,  
  "reason": "Bright ring  
surrounding central  
dark region."  
}
```

Images



Natural image



Astronomical image

Submit input
to VLM/LLM

VLM/LLM



Get
embedding



Regress /
Classify

Prediction

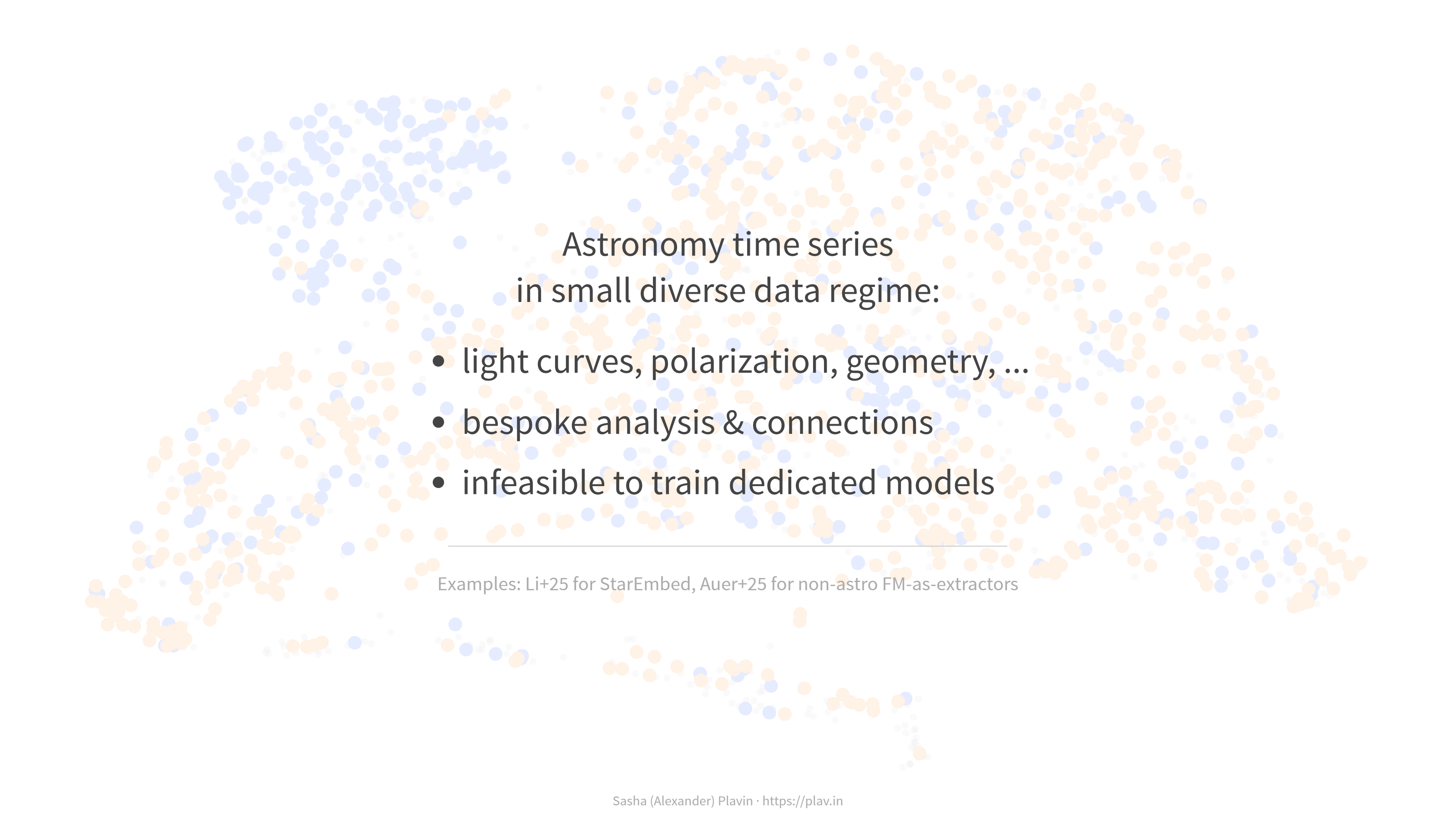


e.g.,
score = 0.93

or



e.g.,
full ring

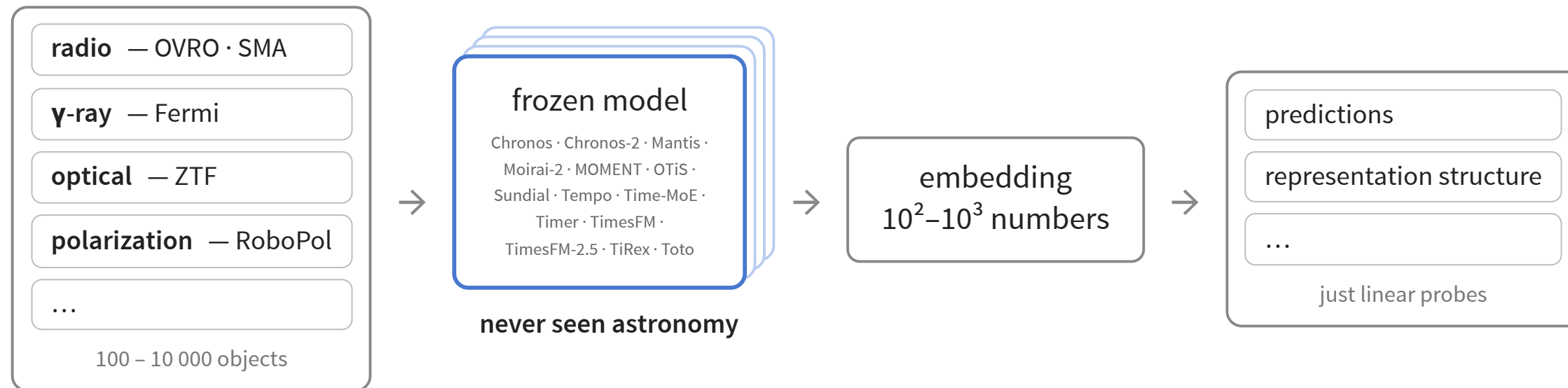
The background of the slide is a scatter plot with numerous small dots. The dots are colored in two main groups: blue and orange. They are scattered across the entire slide area, with some clusters and many empty spaces, creating a sparse, star-like pattern.

Astronomy time series in small diverse data regime:

- light curves, polarization, geometry, ...
- bespoke analysis & connections
- infeasible to train dedicated models

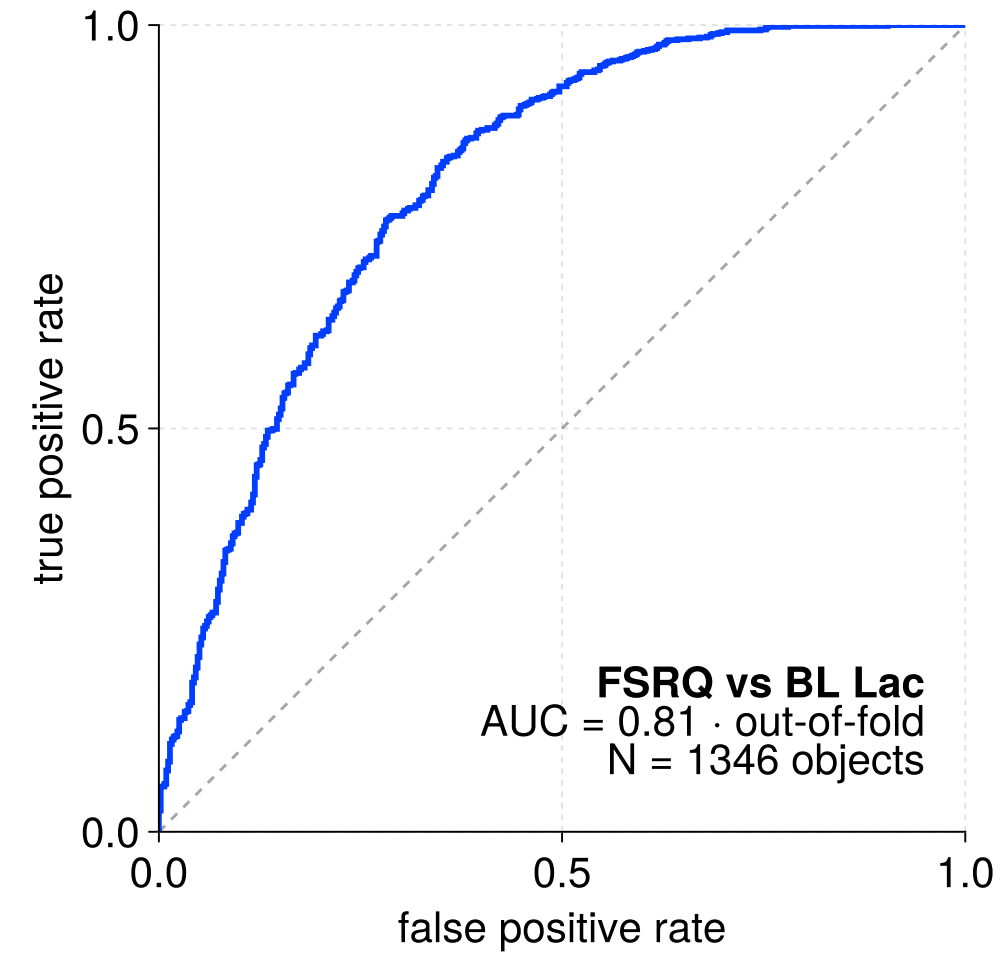
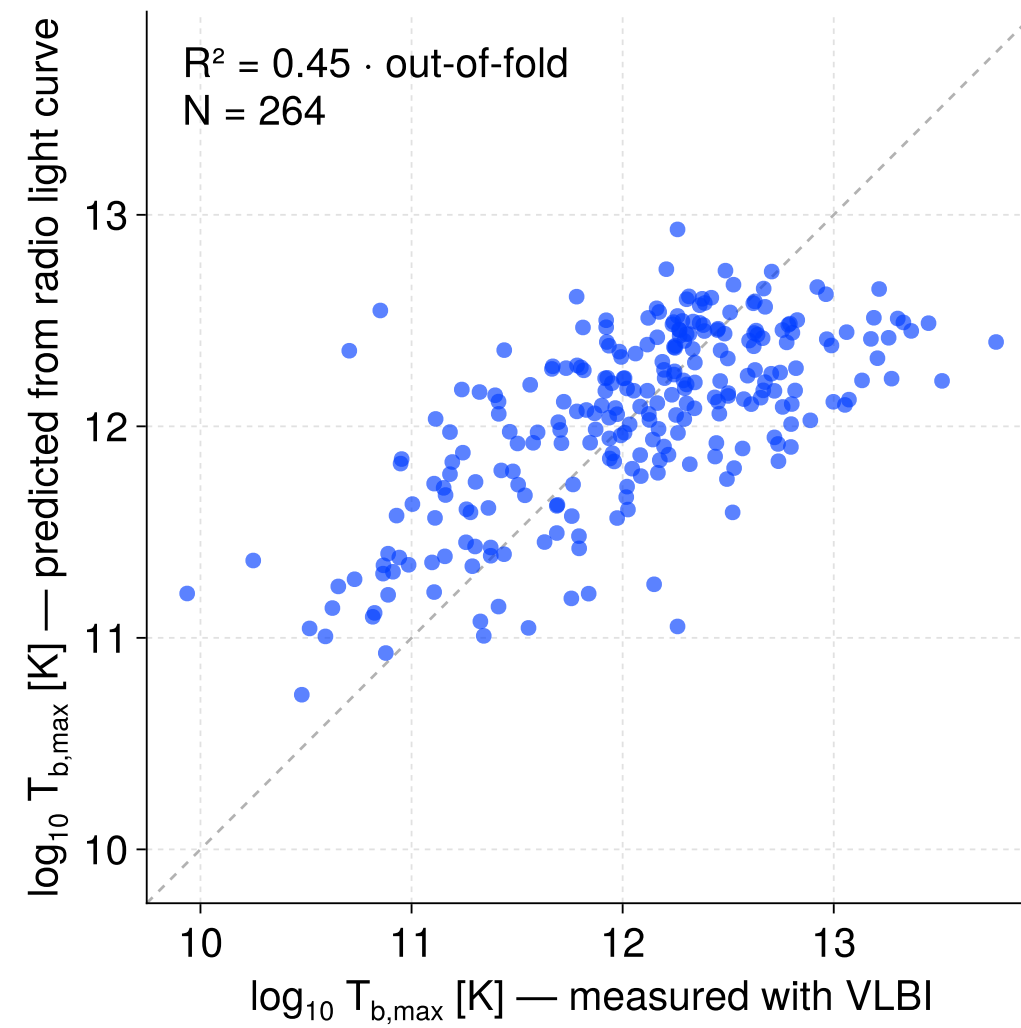
Examples: Li+25 for StarEmbed, Auer+25 for non-astro FM-as-extractors

what do we do?

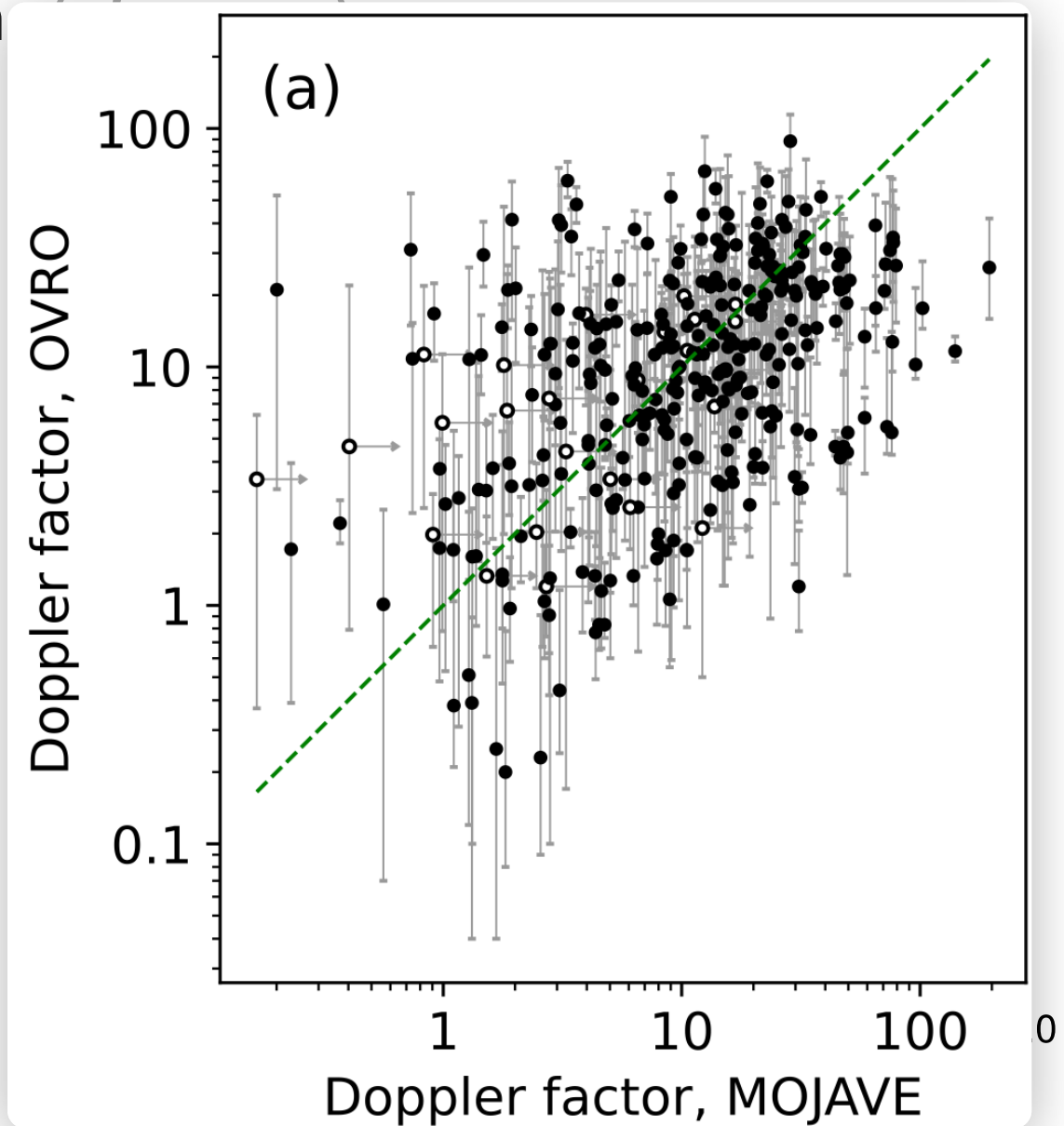
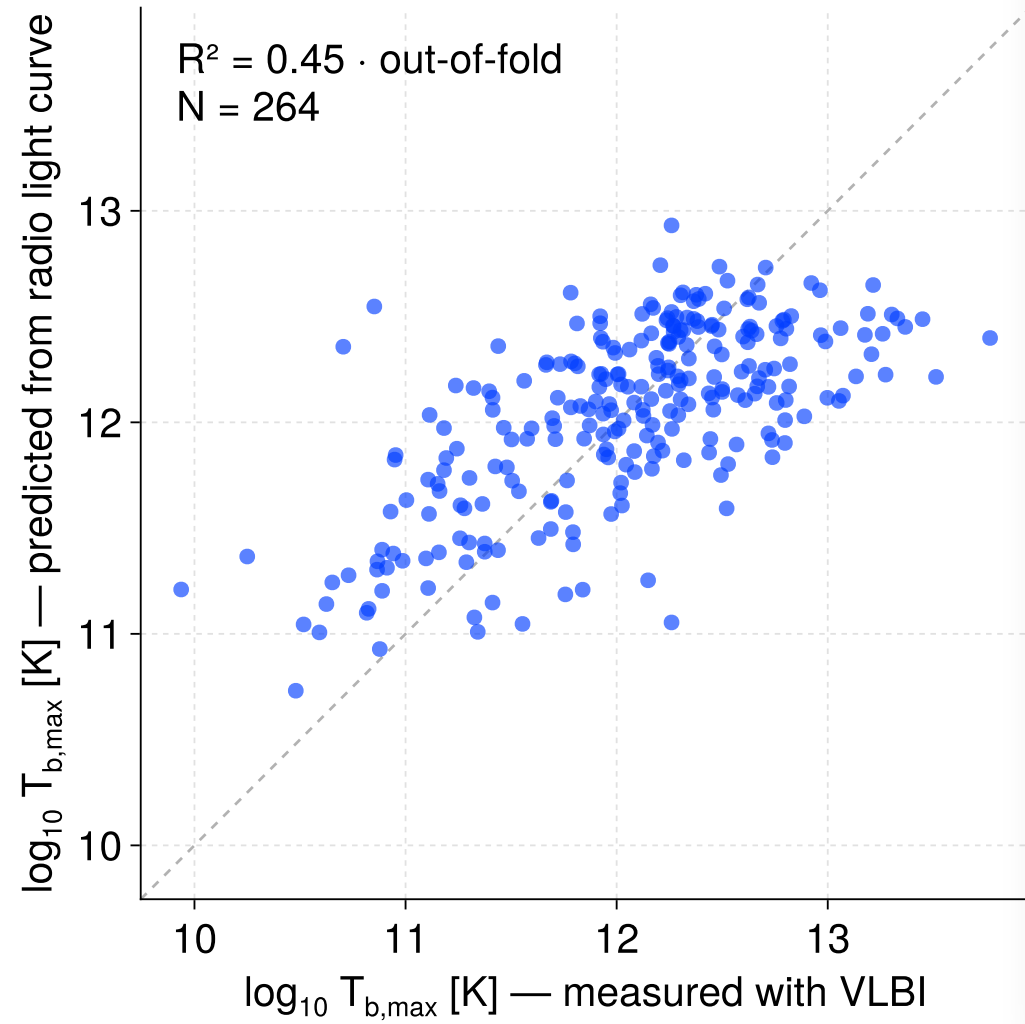


what do we get?

predictions (of course)

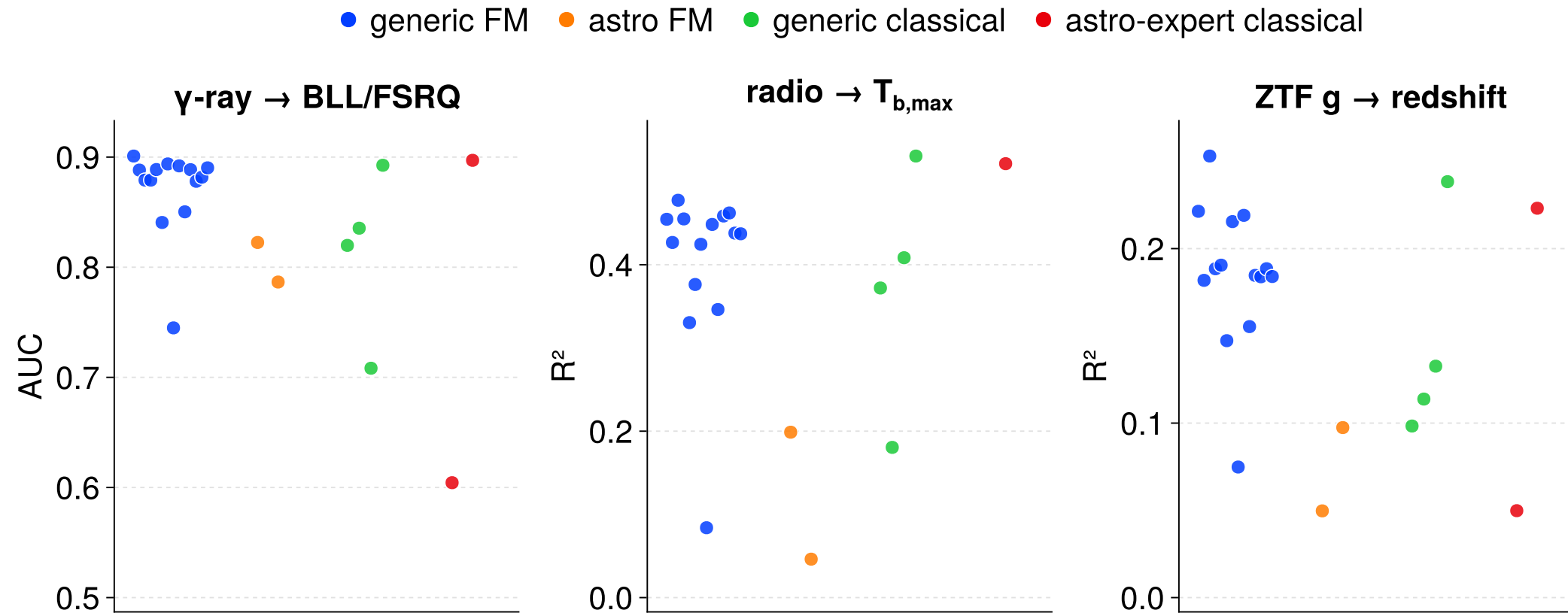


prediction



predictions: quality

the audit: 31 physical targets × 6 bands × 20+ extractors — cross-validated, out-of-fold

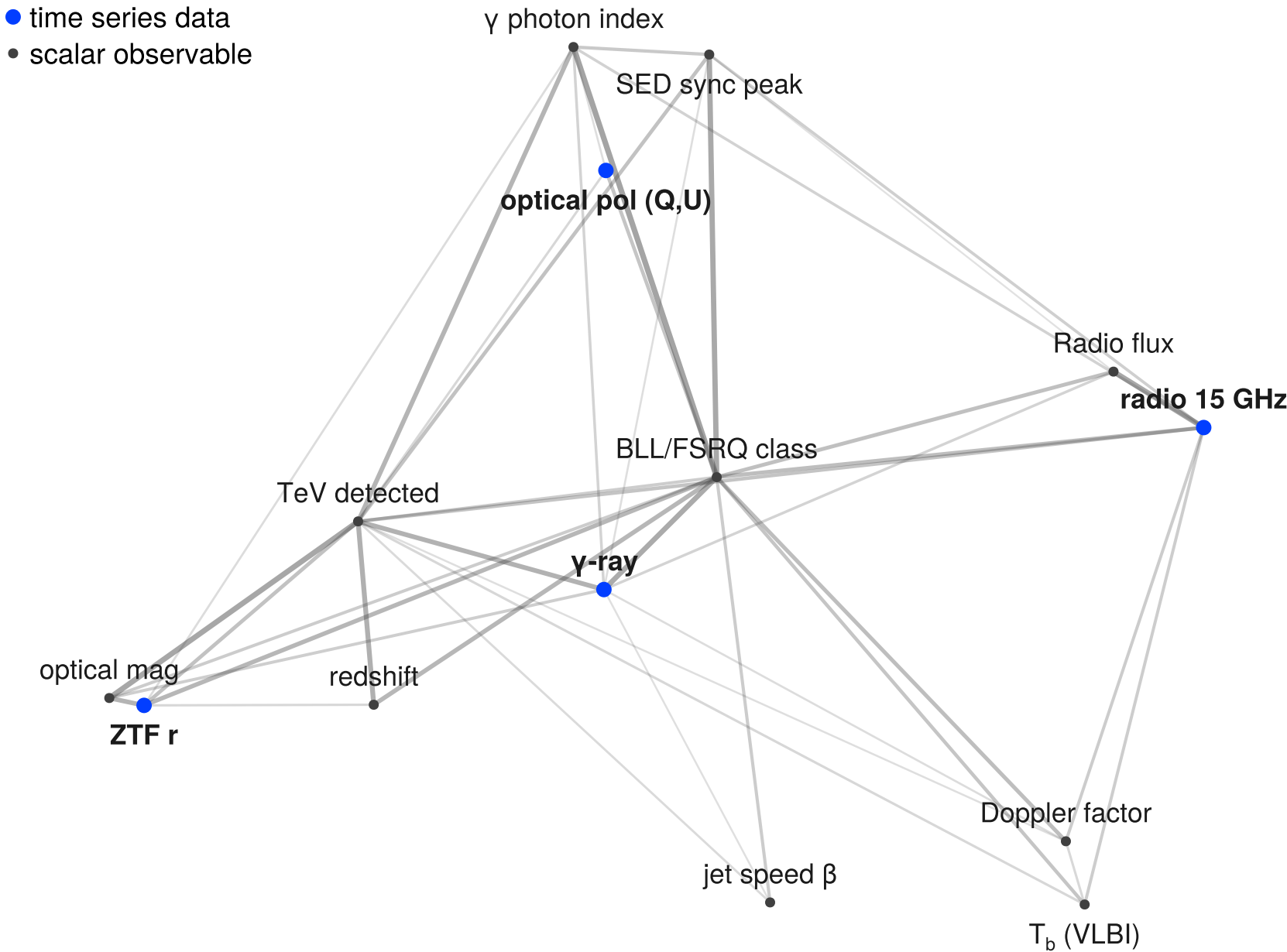


my personal pick: chronos from amazon

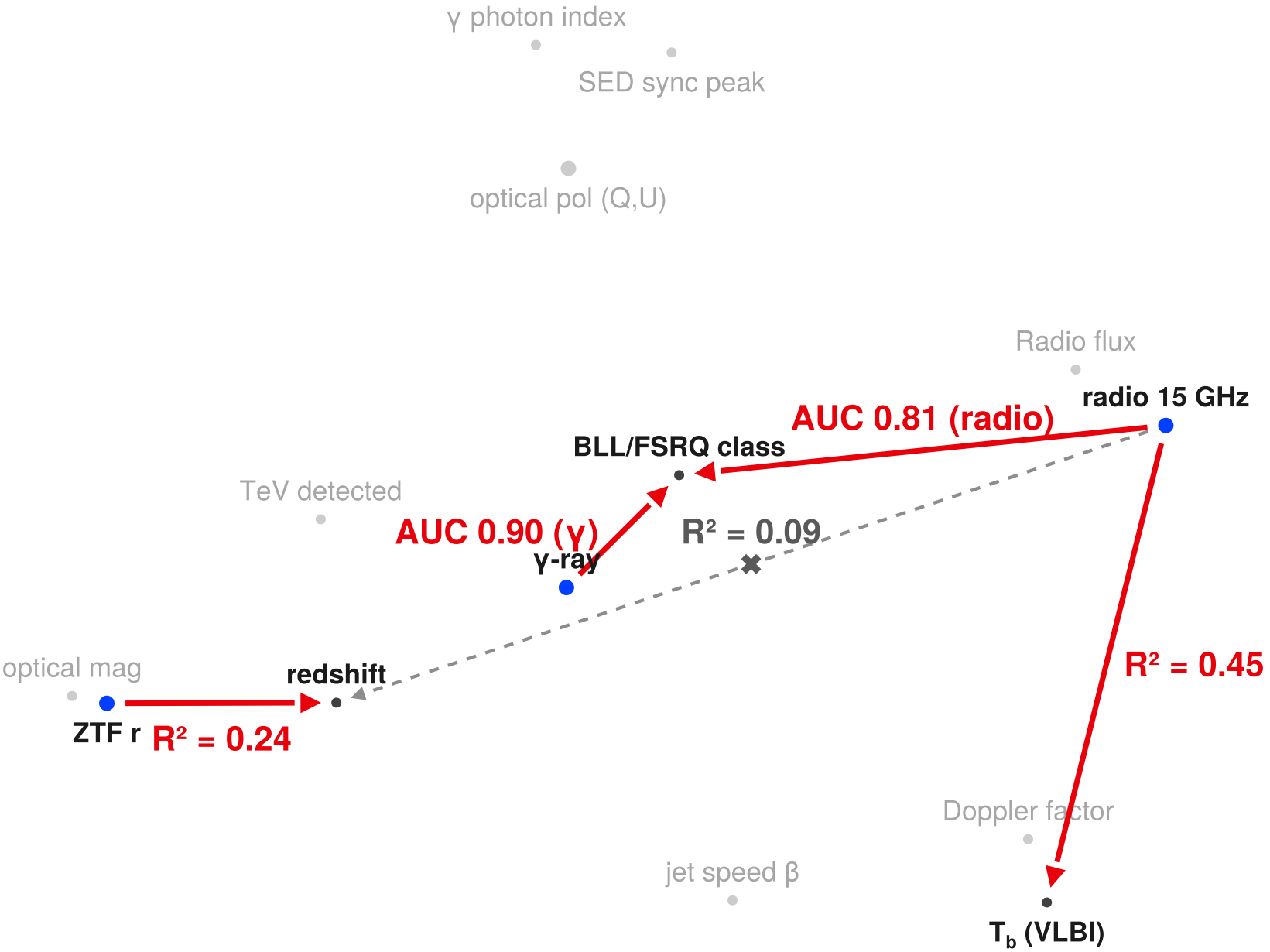
even better than astro-FM (Astromer, StarCLR) – at zero cost for us

tie with classical – networks learn expert-crafted features

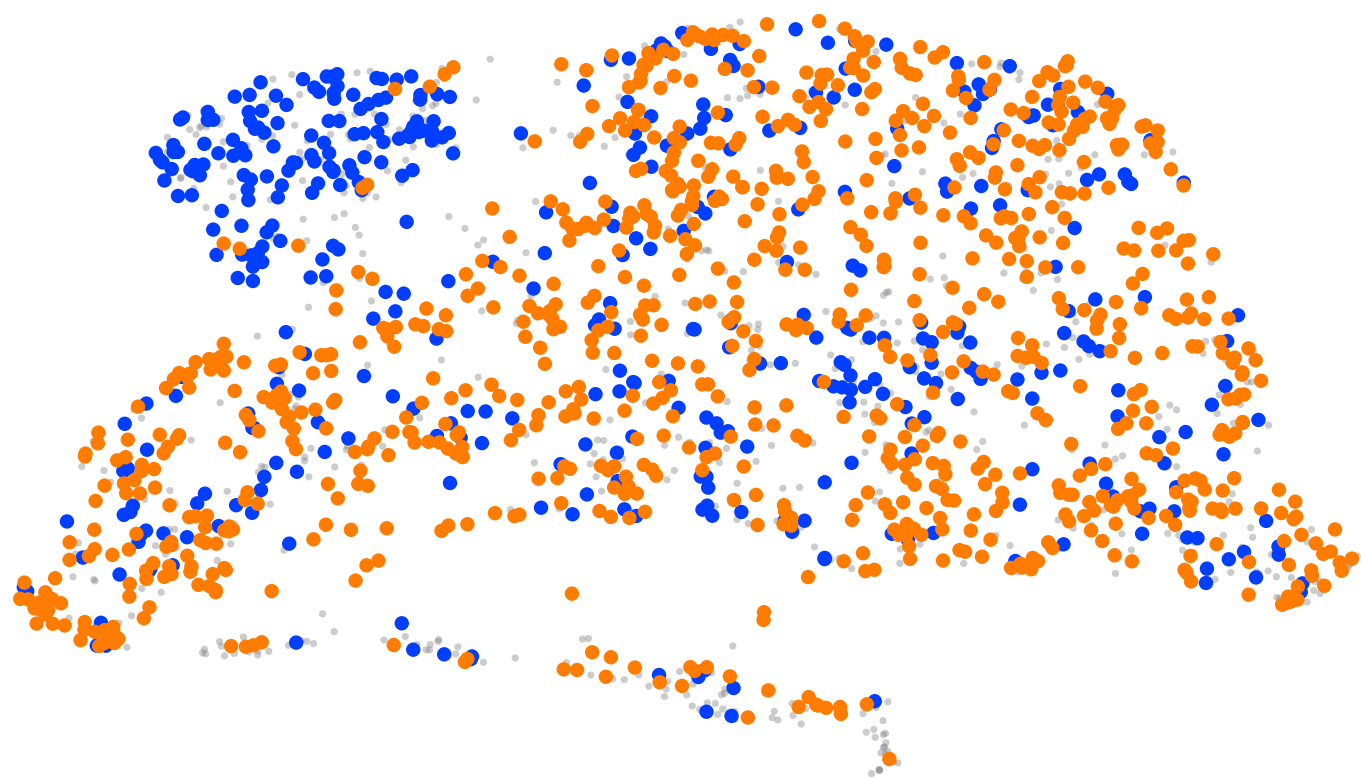
predictability navigation chart



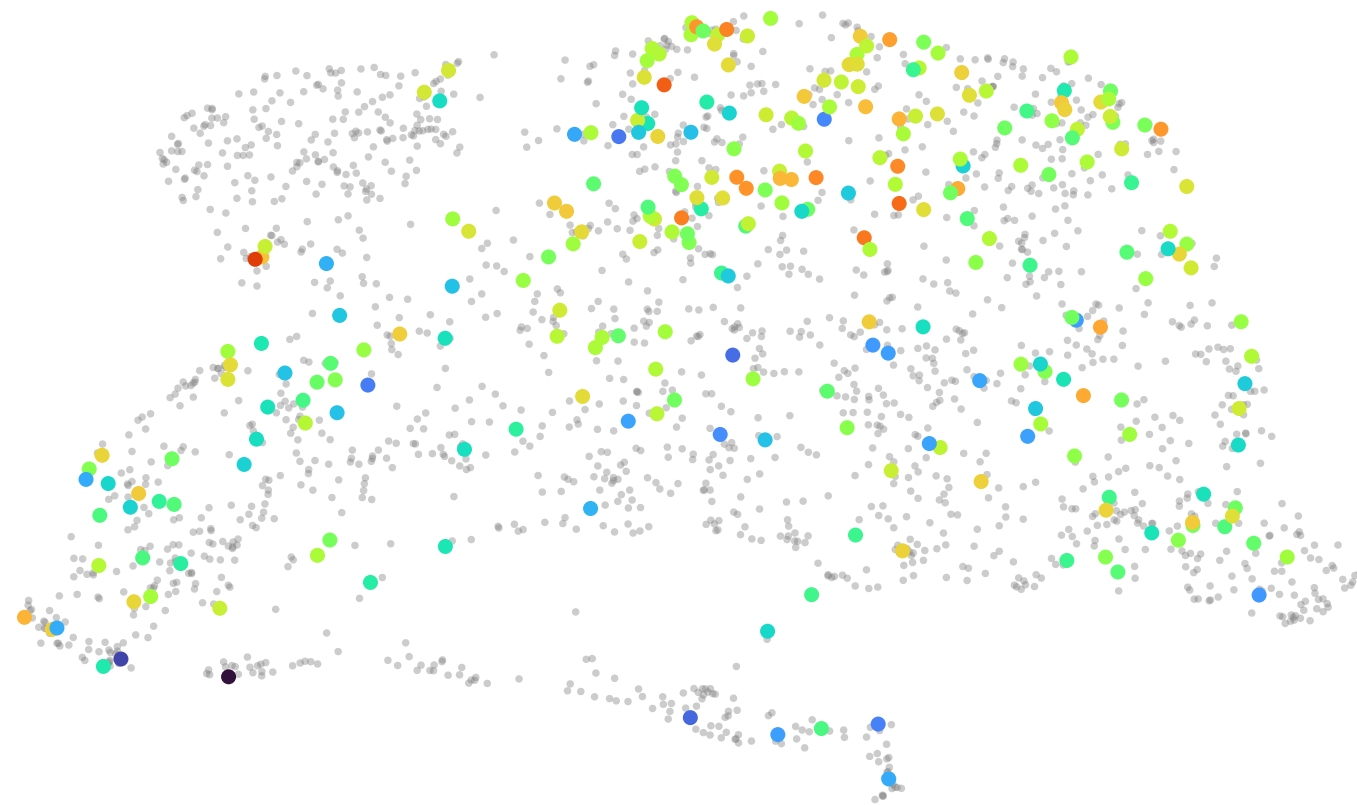
predictability navigation chart



the representation space

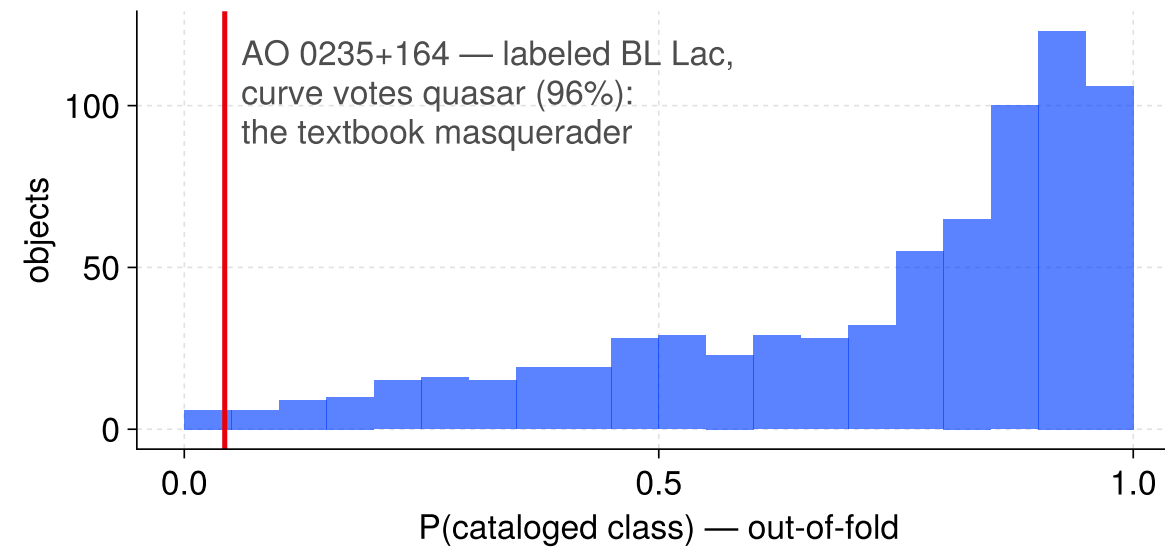
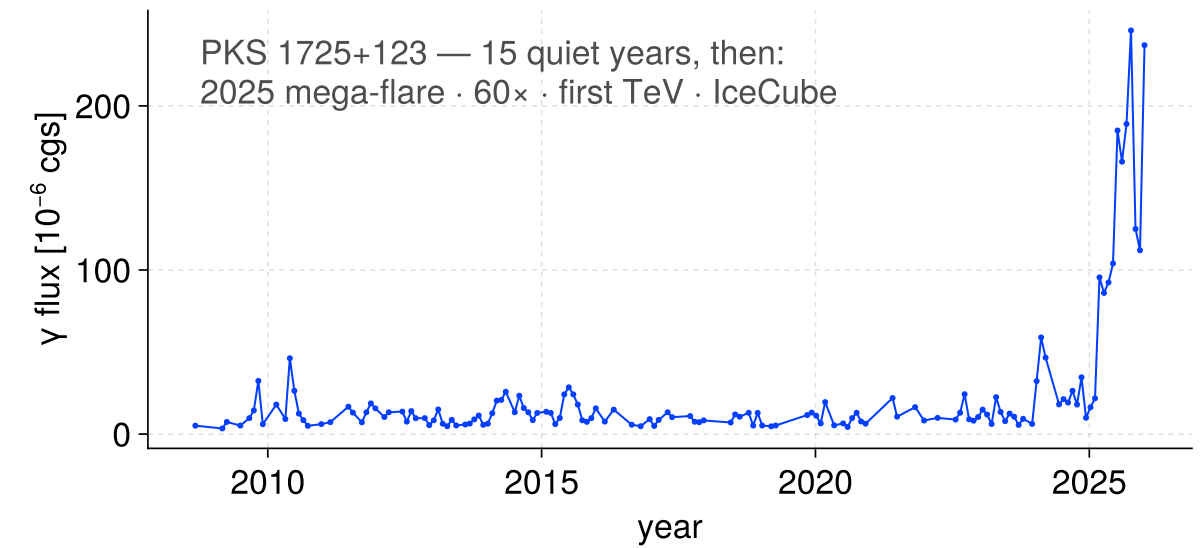
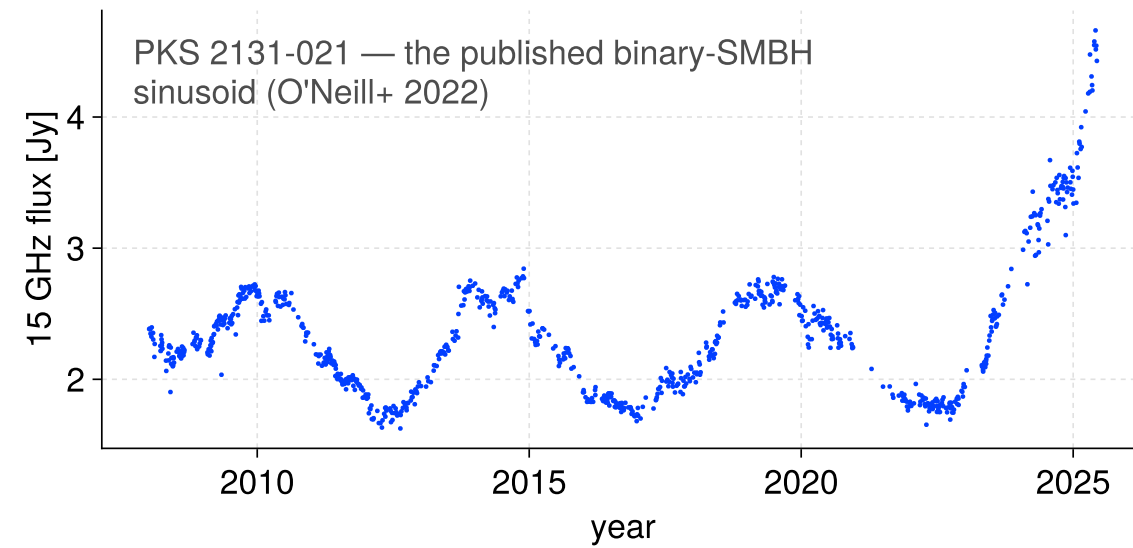


by quasar type: strong structure



by VLBI core brightness: subdominant despite successful prediction

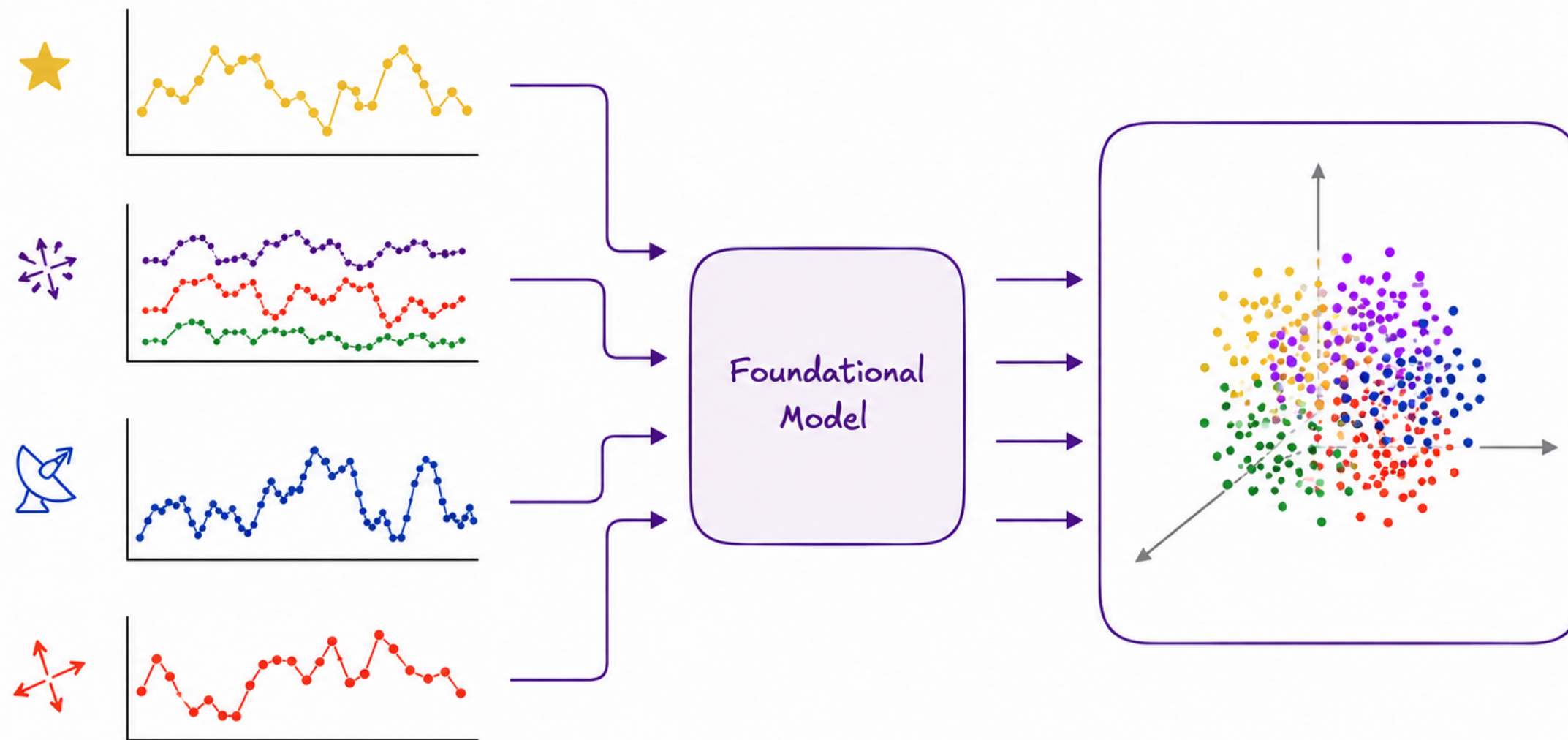
anomaly detection (manually inspected top-12)



... also: an outdated redshift caught,
and more, known and unknown — found automatically

What's still hard

- extremely irregular time coverage
- joint multi-timeseries embeddings: delays, time correlations, ...
- shared embedding spaces: not directly yet; maybe learn simple linear transforms?



Are **foundational time-series models** here already?

tie hand-crafted features and bespoke analysis – for free

- frozen weights + simple linear probe
- from **prediction** to **unfamiliar field exploration**
- ultimately, any data shape: image, timeseries, ...
 - now – the worst these models will ever be

every feature extractor

generic foundation models (14)

Chronos · Chronos-2 · Mantis · Moirai-2 · MOMENT · OTiS · Sundial · Tempo · Time-MoE · Timer · TimesFM · TimesFM-2.5 · TiRex · Toto — from Amazon, Google, Datadog, Salesforce, Tsinghua, ...

astro foundation models (2)

Astromer · StarCLR — pretrained on astronomical light curves, fed their native irregular cadence

generic classical features (4)

catch22 · tsfresh · signatures · stats — hand-crafted time-series feature banks + simple summary statistics

astro-expert classical features (2)

gp-bank · lcf — GP/DRW variability-model fits and astronomy light-curve feature sets

all compared at the same settings: linear probe on top, cross-validated, out-of-fold