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Space Sciences



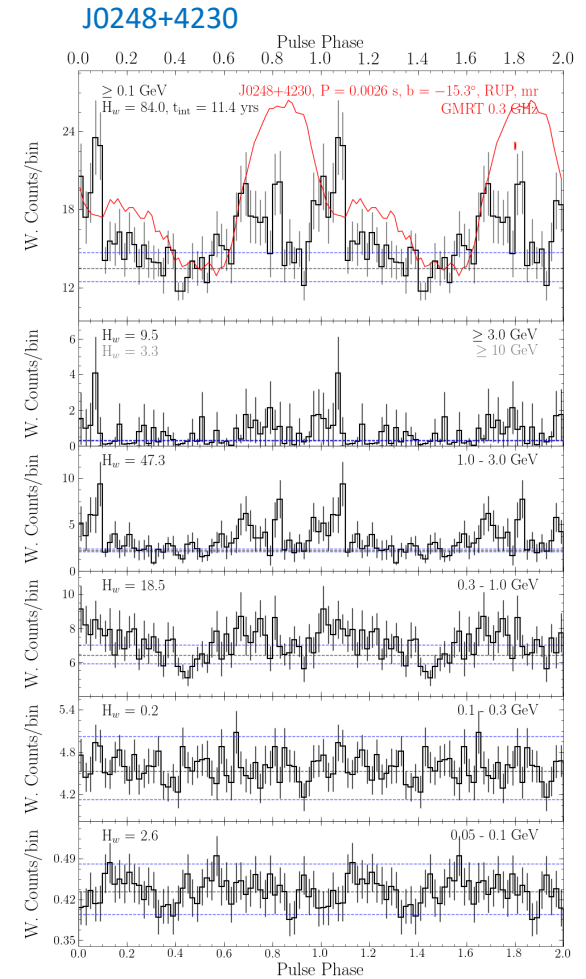
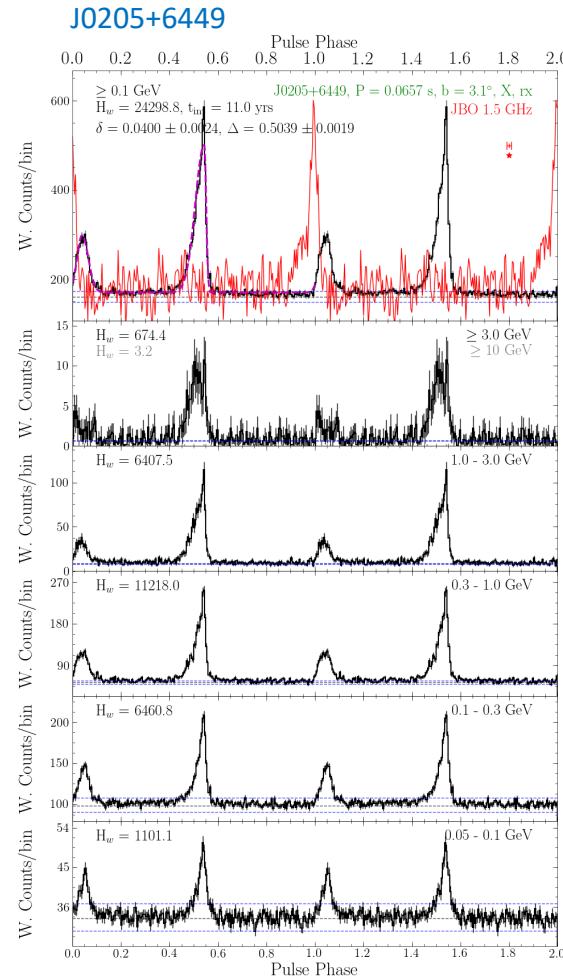
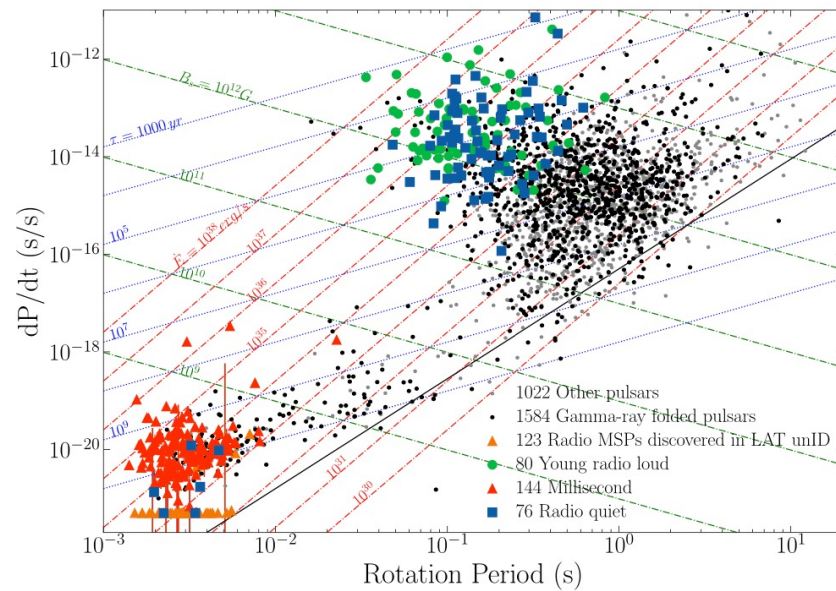
# Quantitative exploration of pulsar light curve similarity

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Research done in collaboration with C. R. García

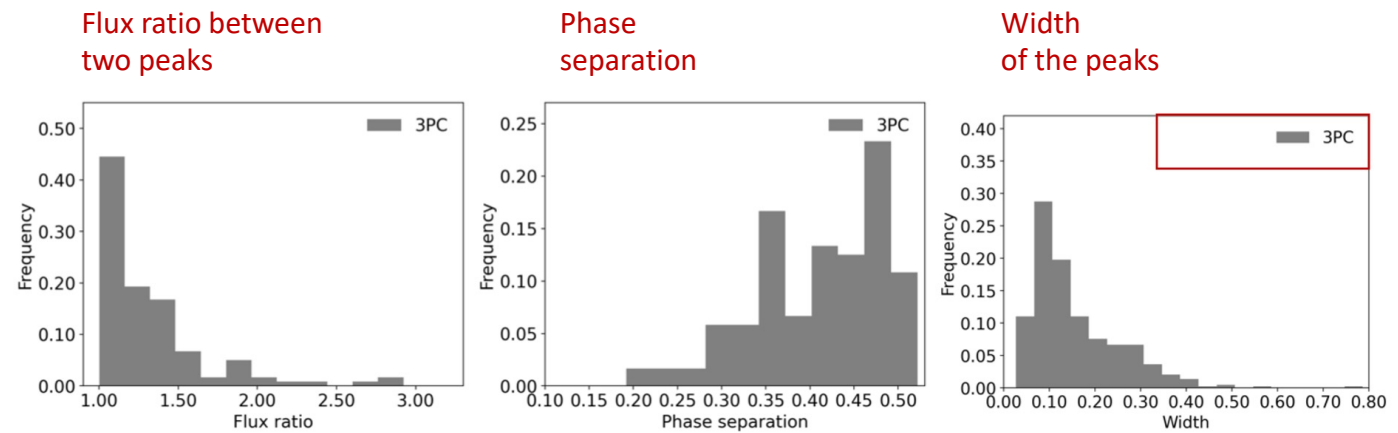
# Variety of Pulsar Light curves in the 3PC

- About 300 pulsars reported
- 3PC light curves sampled with different levels of precision (from 25 to 800)



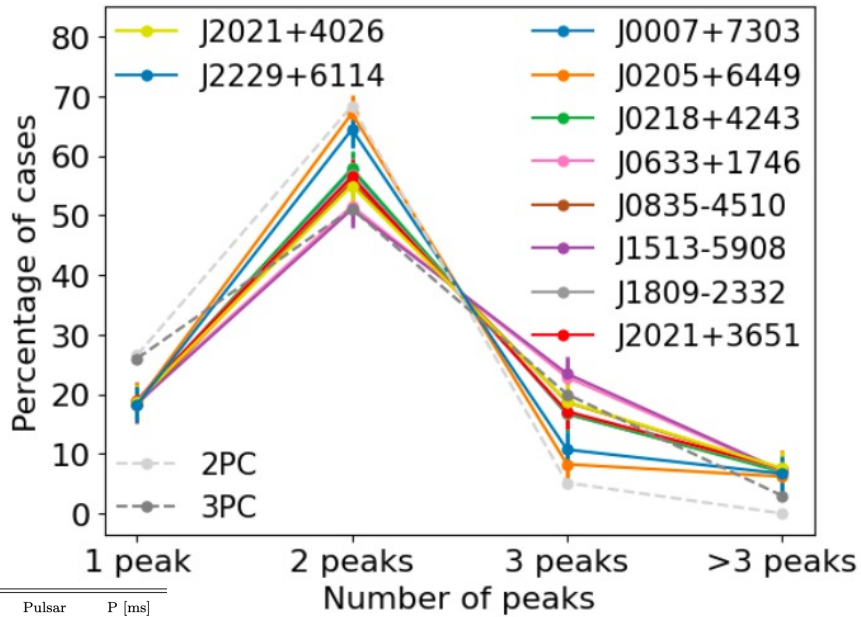
# Variety of Pulsar Light curves in the 3PC

- Light curves described via global features



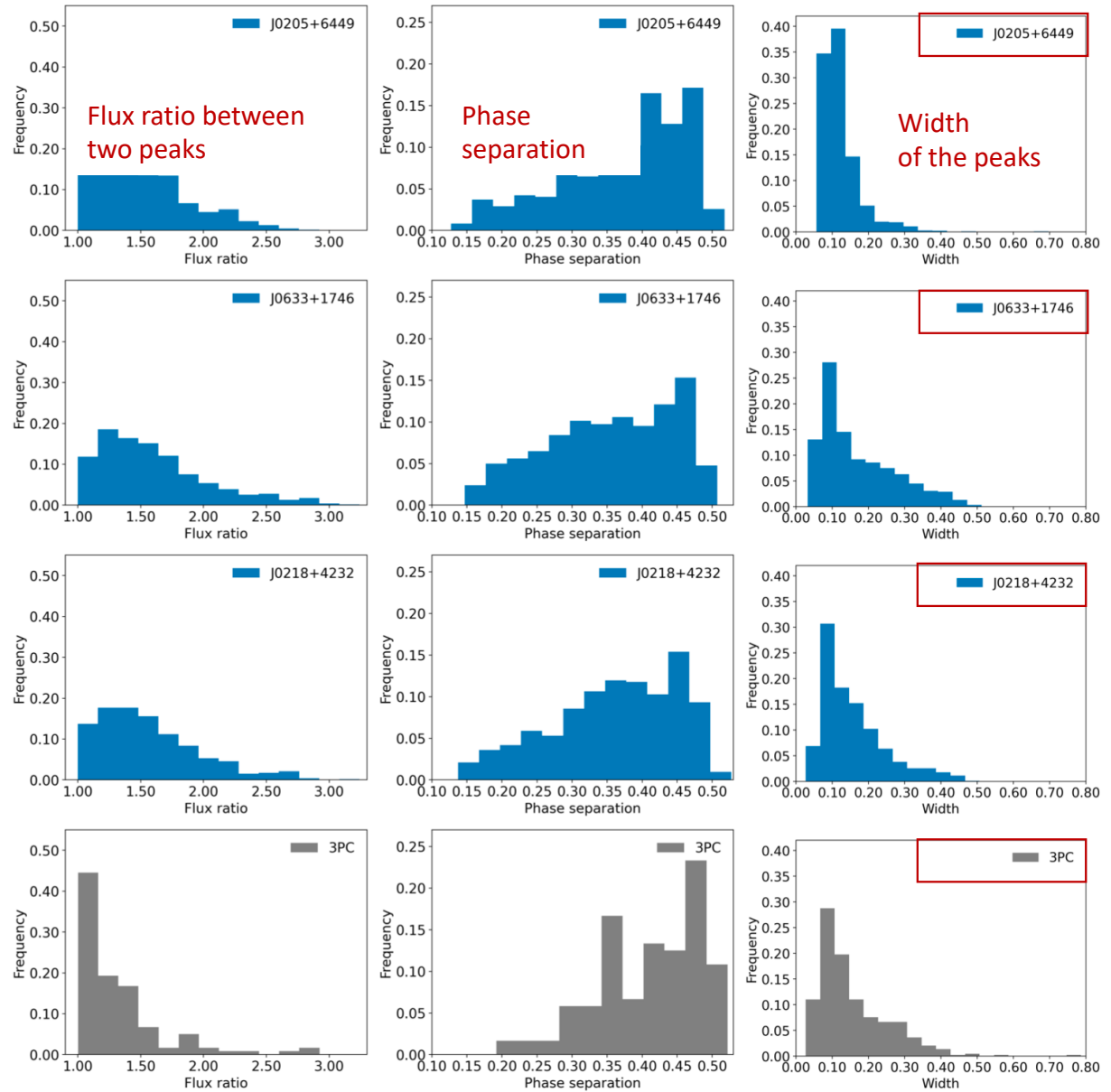
# Light curves global properties

Number (or percentage) of light curves with n peaks

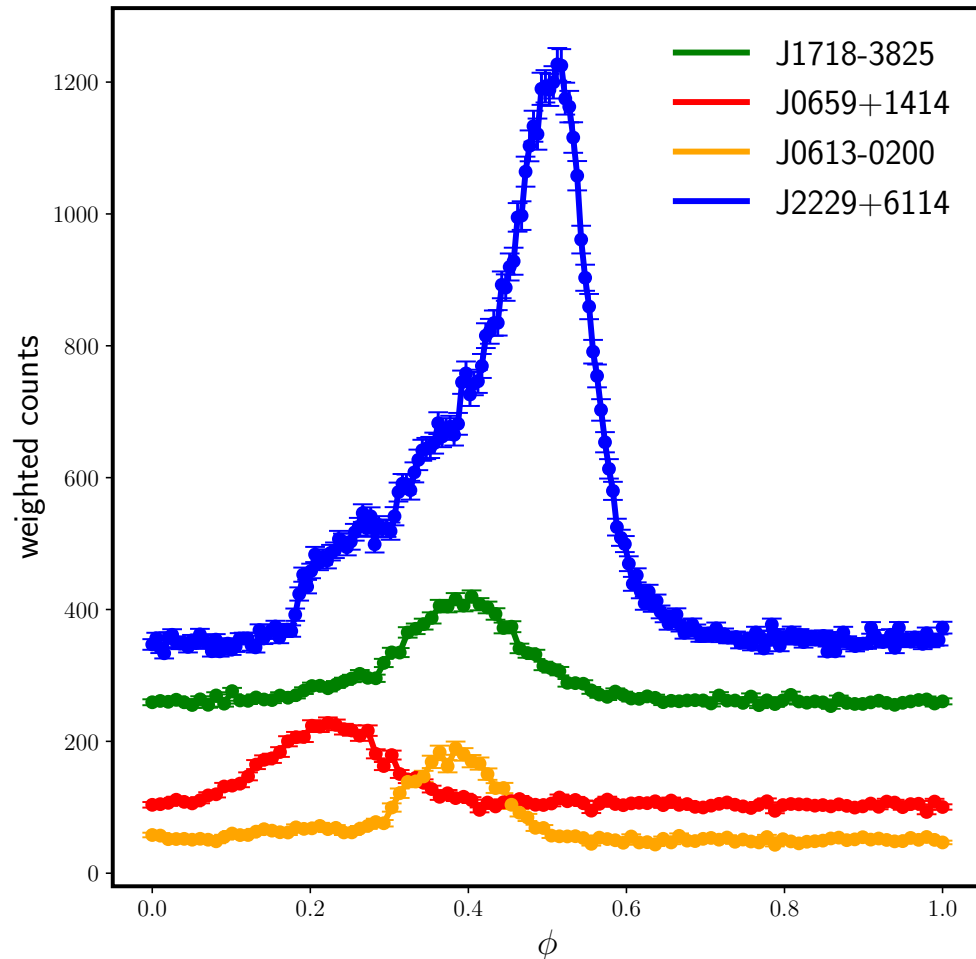


Pulsar	P [ms]
J0007+7303	315.9
J0205+6449	65.7
J0218+4232	2.3
J0633+1746	237.1
J0835-4510	89.4
J1513-5908	151.6
J1809-2332	146.8
J2021+3651	103.7
J2021+4026	265.3
J2229+6114	51.7

Iñiguez-Pascual, Torres, Viganò, MNRAS 2024 (see talk by D. Viganò)



# Variety of Pulsar Light curves in the 3PC



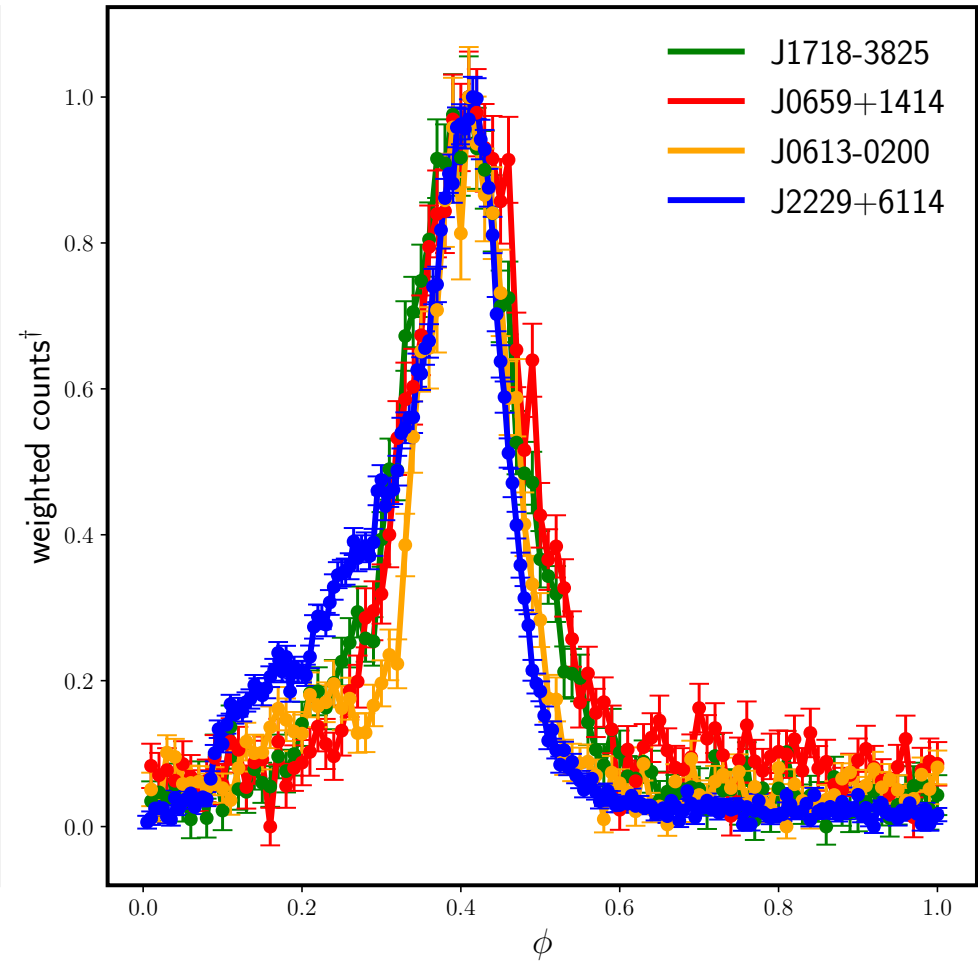
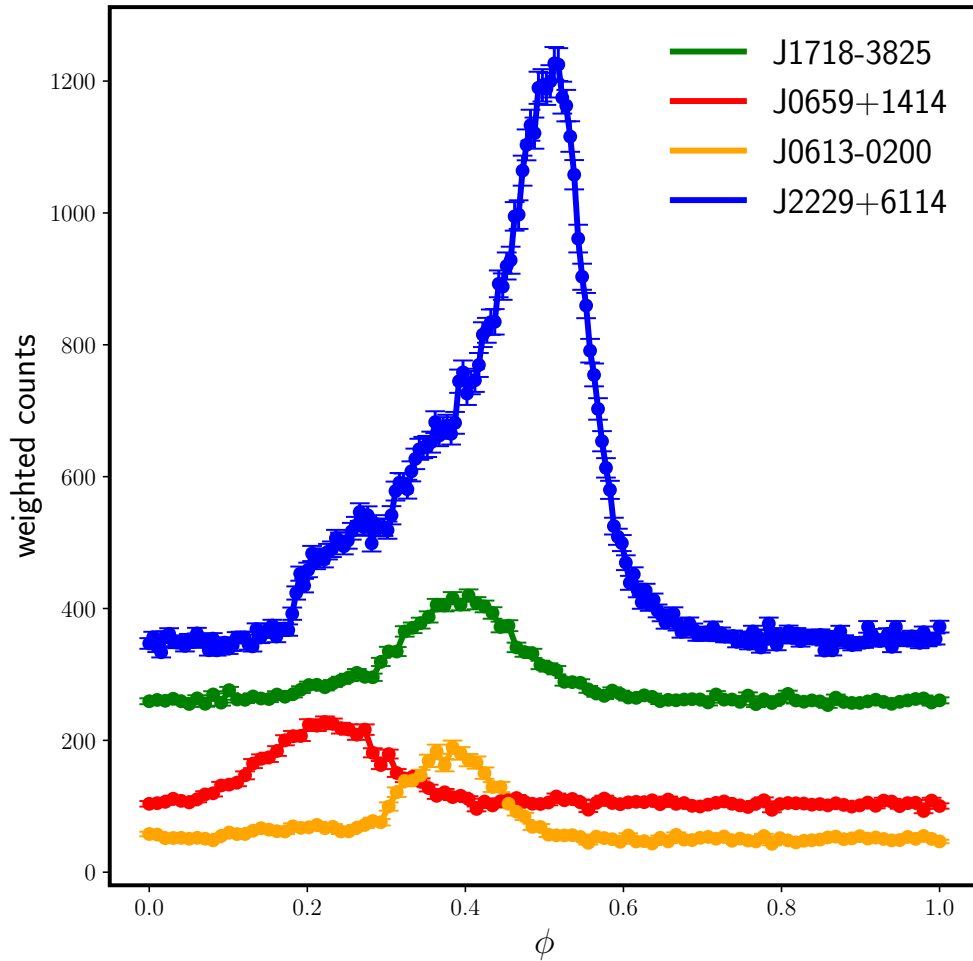
## 3PC light curves examples, obtained directly from the catalog

- Very different in flux level and shape, are they?
- Consider the transformation

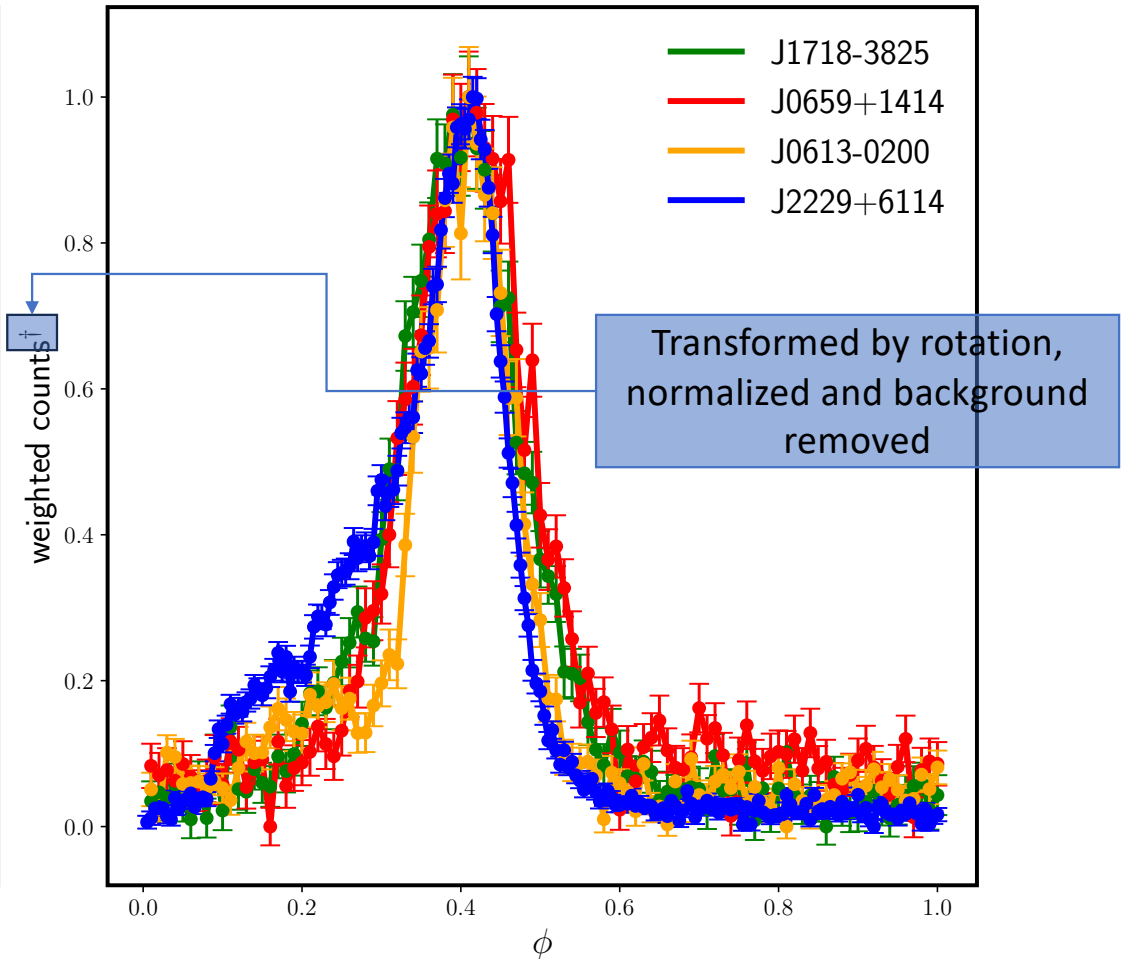
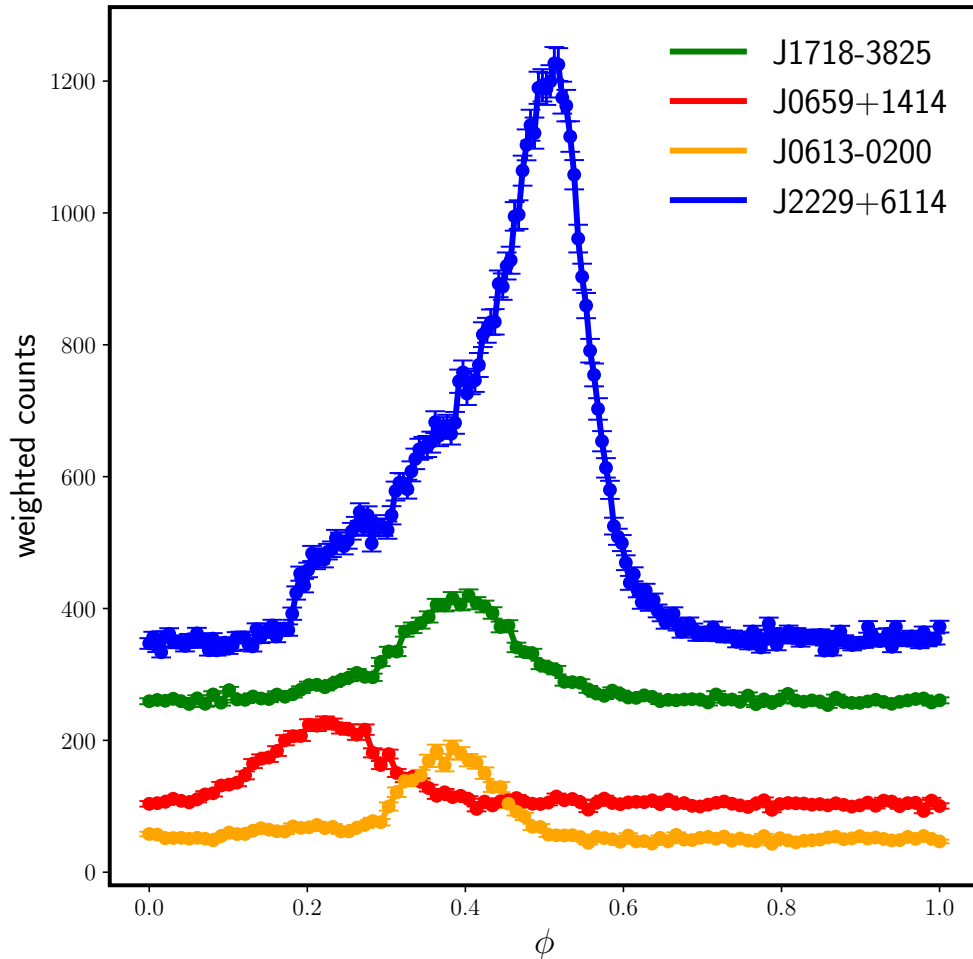
$$\text{Counts} \rightarrow (\text{Counts} - \text{Background}) / (\text{Max-Background})$$

- And all light curves rotations in phase
- Based on:
  - Phase is arbitrary so rotate them until 'alignment'
  - maps all light curves in a range from 0 to 1 in 'counts'
  - emphasizes morphological differences

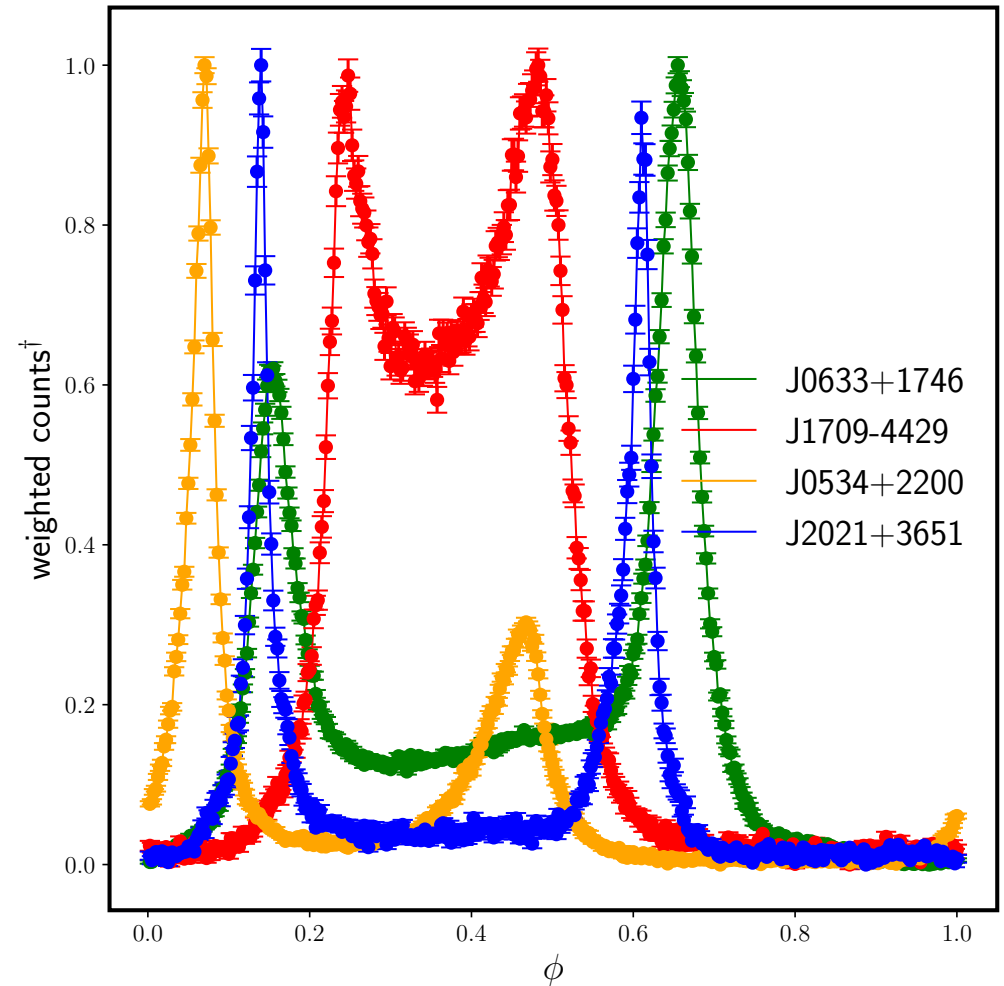
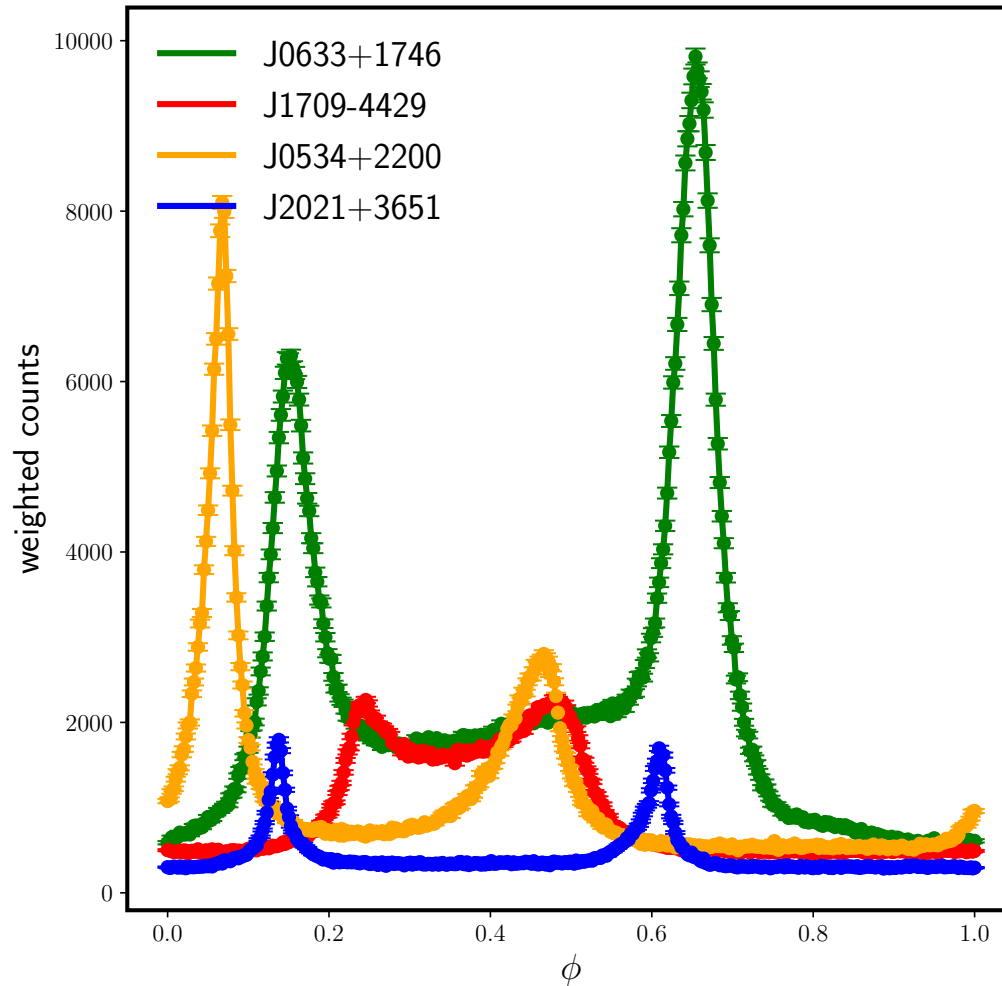
# A realization: morphological similarity underlies



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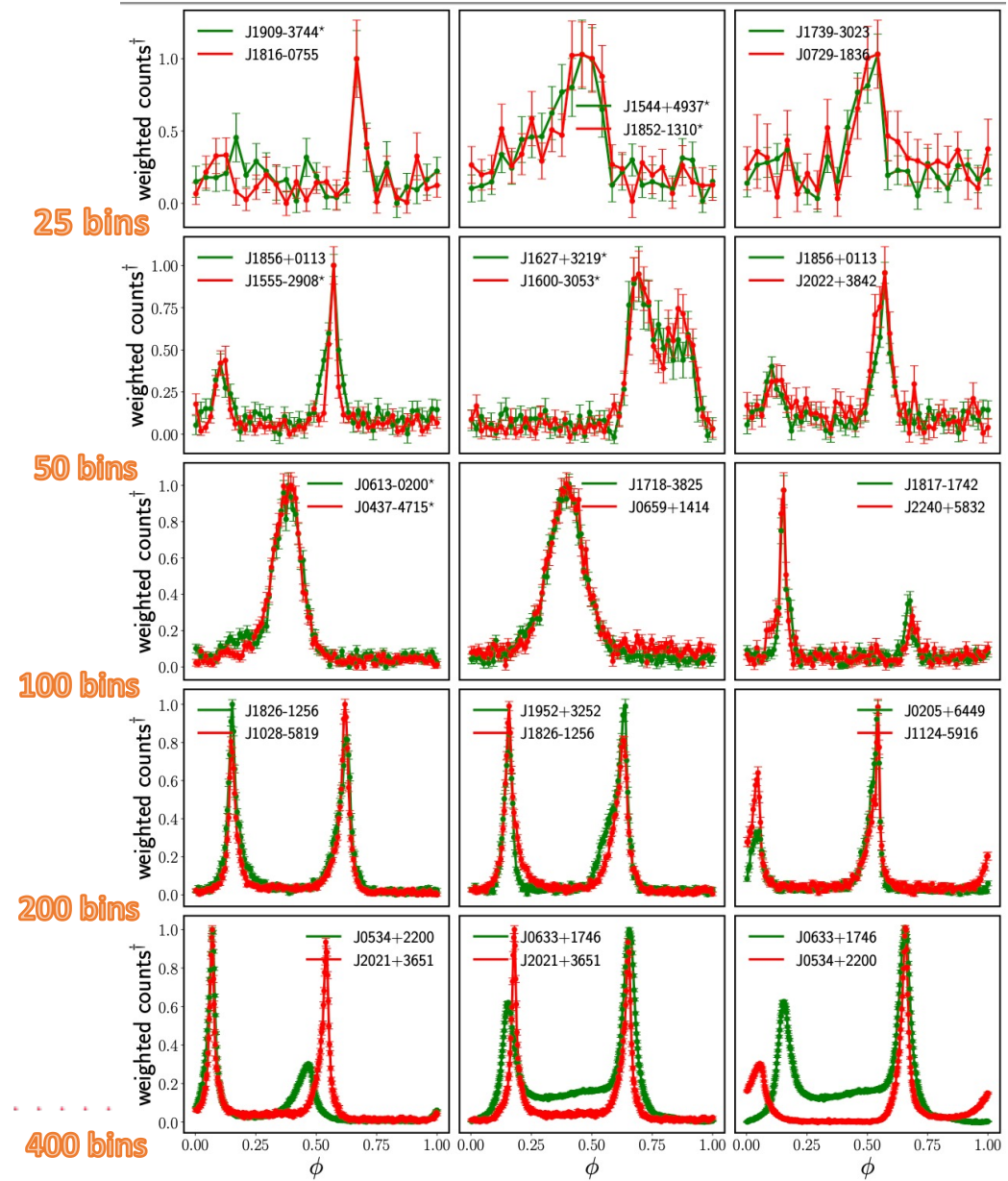
# Not always of course, there is also dissimilarity





Consider Euclidean distance from each pulsar to all others, including all rotations, and rank them

- Arising similarities all over the sample
- Significant matching in many cases
- **But limited**
  - to comparing pulsars having the same number of bins in the light curve, or subject to a rebinning process to make them so, with the subsequent loss of information
  - Very similar pulsars, not precisely aligned are not singled out as being similar



# Dynamic time warping: concept

- DTW is an optimization method used to compare time series.
- It works by **dynamically aligning** these time series even with **different sizes**.
- The Euclidean distance (ED) is used once such dynamic alignments (or paths) are established.
- The goal is the optimization of the, so the path with the minimum cost is labelled as *the optimal warping path* and its cost will be *DTW value*.

# Dynamic time warping: math

- For this process to be effective we set the next conditions to consider a path:
  - **Boundary condition.**
    - The initial and final elements of the time series must face each other respectively.
  - **Monotonicity condition.**
    - No elements of the series that break the temporal order of the series can be matched against each other.
  - **Continuity condition.**
    - The elements of the series being aligned must be adjacent points, not allowing temporal jumps.
- The resolution for this optimal problem can be seen via:

$$DTW(X, Y) = \min_{\pi \in A(X, Y)} \left( \sum_{(i, j) \in \pi} d(x_i, y_j)^2 \right)^{\frac{1}{2}}$$
$$\left\{ \begin{array}{l} X, Y: \text{time series} \\ A: \text{set of paths} \\ \pi: \text{path} \\ d: ED \end{array} \right.$$

# Dynamic time warping: practice & simple example

- We define the time series as:

$$ts_1 = 3, 1, 2, 2, 1$$

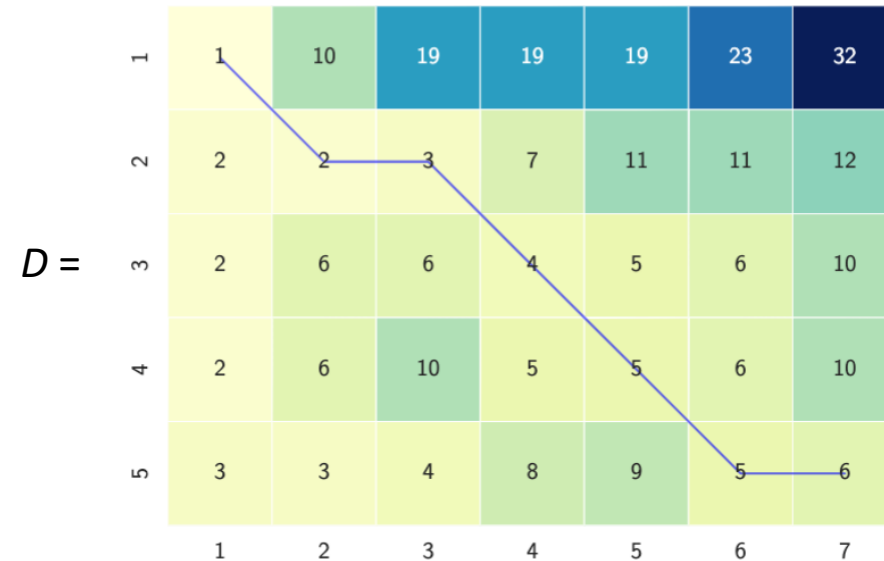
$$ts_2 = 2, 0, 0, 3, 3, 1, 0.$$

- The ED matrix ( $E$ ) between the  $ts_1$  and  $ts_2$ :

$$E(d(i, j)^2) = \begin{bmatrix} 1 & 9 & 9 & 0 & 0 & 4 & 9 \\ 1 & 1 & 1 & 4 & 4 & 0 & 1 \\ 0 & 4 & 4 & 1 & 1 & 1 & 4 \\ 0 & 4 & 4 & 1 & 1 & 1 & 4 \\ 1 & 1 & 1 & 4 & 4 & 0 & 1 \end{bmatrix}$$

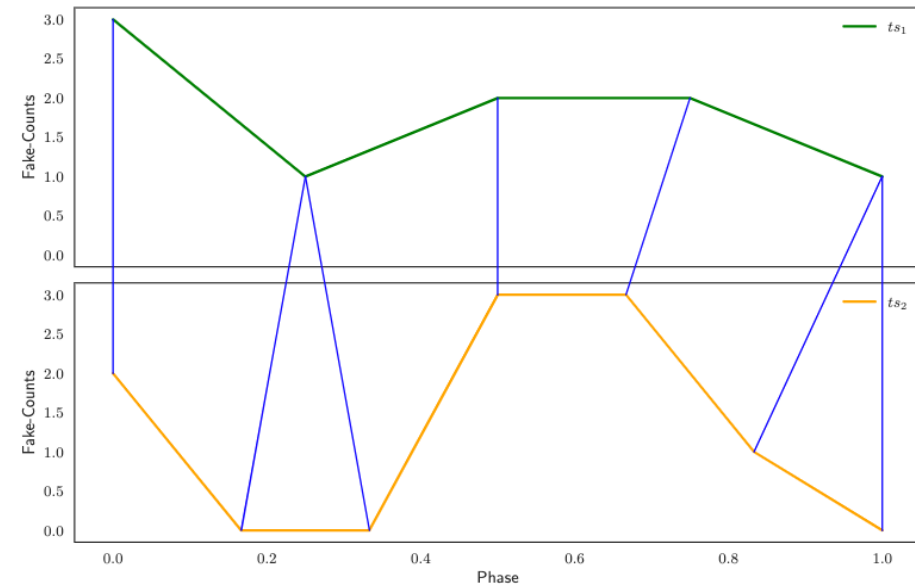
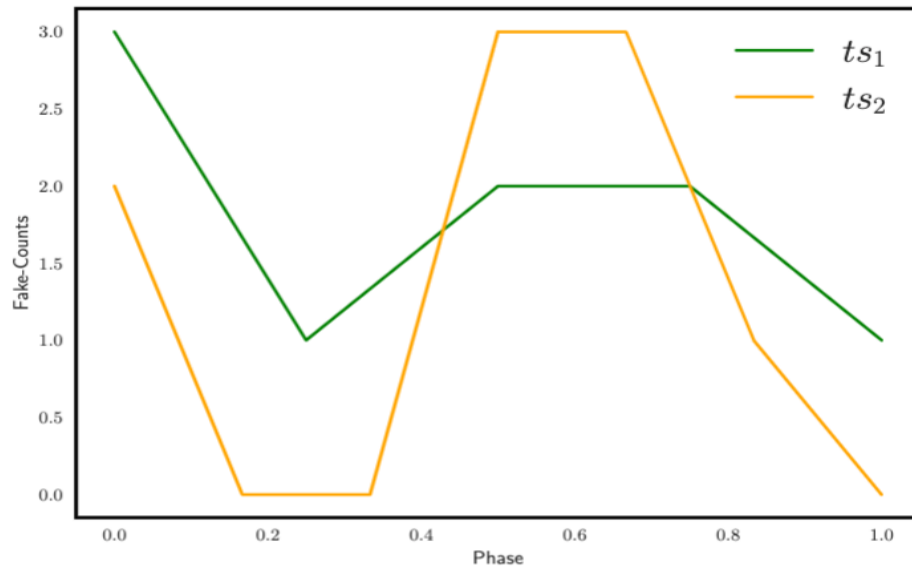
- The cumulative cost matrix through:

$$D(i, j) = E(i, j) + \min(D(i-1, j), D(i, j-1), D(i-1, j-1)).$$



- For instance, the value seen in row 2 and column 2,  $D(2,2) = E(2,2) + \min(D(1,2), D(2,1), D(1,1)) = 1 + \min(10, 2, 1) = 2$ .
- The DTW value is the total cost of the blue line, according to the  $E$ -matrix:  $\sqrt{1 + 1 + 1 + 1 + 1 + 0 + 1} = 2.45$

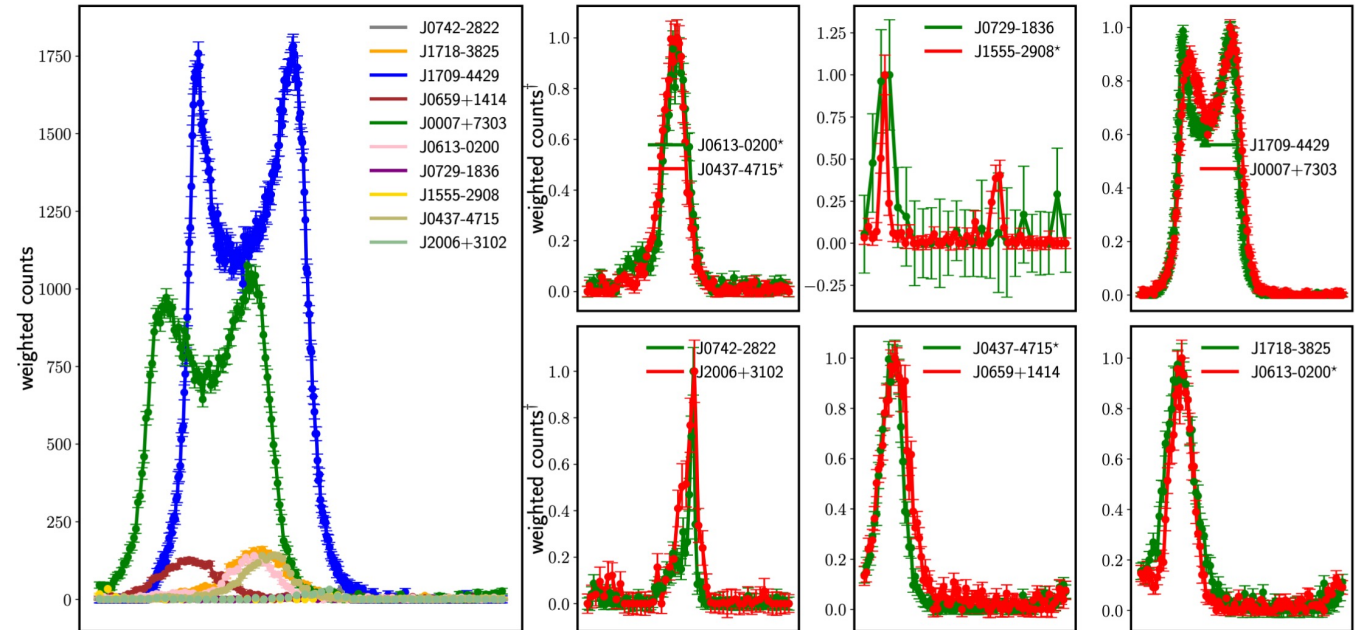
# Dynamic time warping: example in a figure



The *optimal warping path* is represented by the blue lines where each one defines a distance.

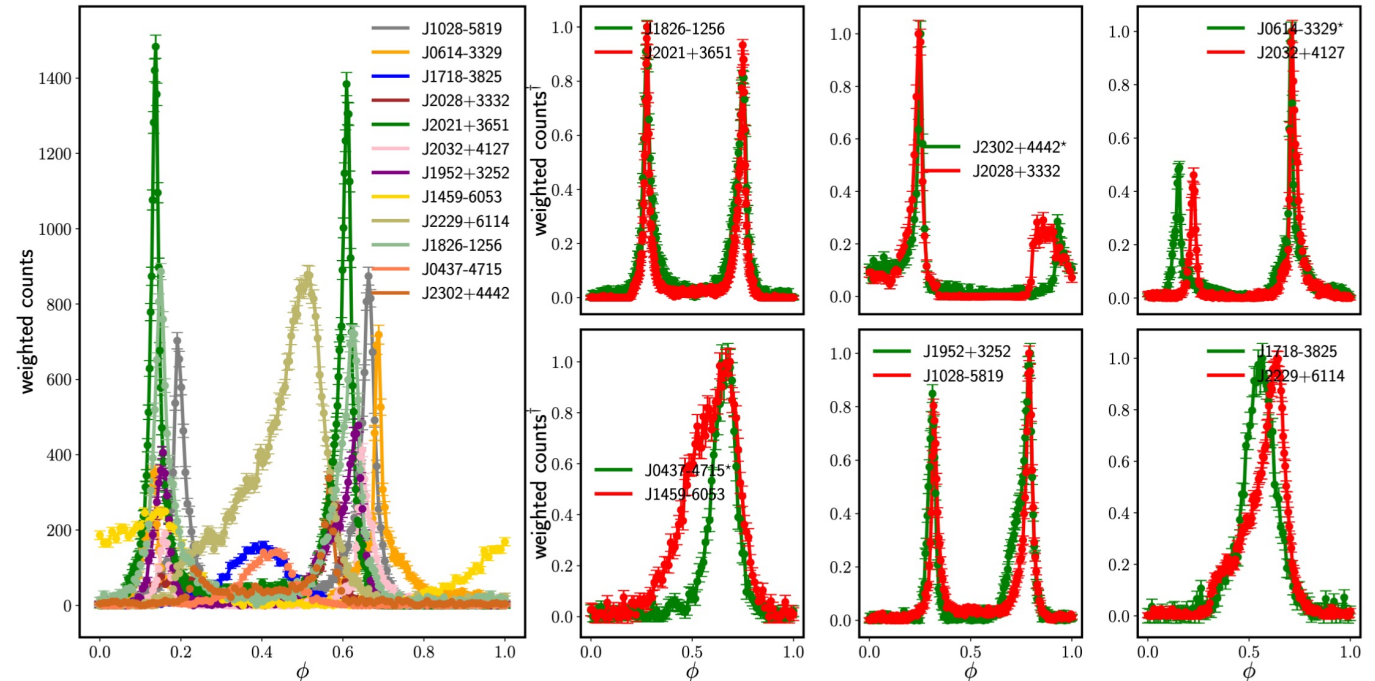
# A comparison devoid of limitations

- A technique useful to find similarities beyond the simple alignment (which is what the Euclidean distance quantifies)
- It associates morphological structures, despite they do not happen at exactly the same phase
- Can be applied to light curves of any number of bins, without rebinning
- Provides a number, that can be used to quantify similarity



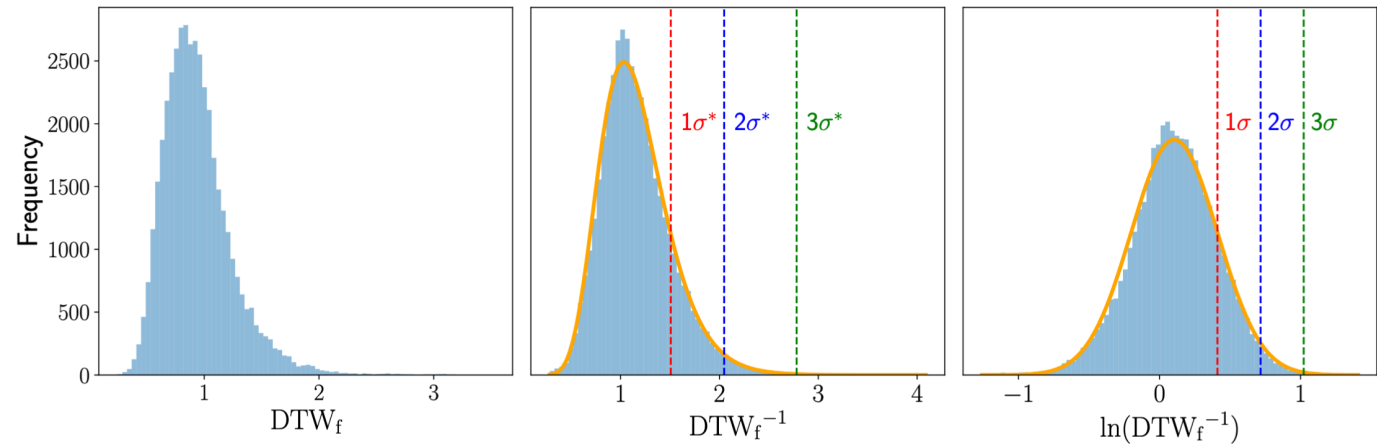
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# A comparison devoid of limitations

- Computing the dynamic distance between all pulsars in the sample, and all its rotations, one obtain its distribution: a global view of how similar/dissimilar pulsar light curves are
- A well behaving quantitative similarity estimator
- Permits to define intervals of confidence for the similarity of light curves



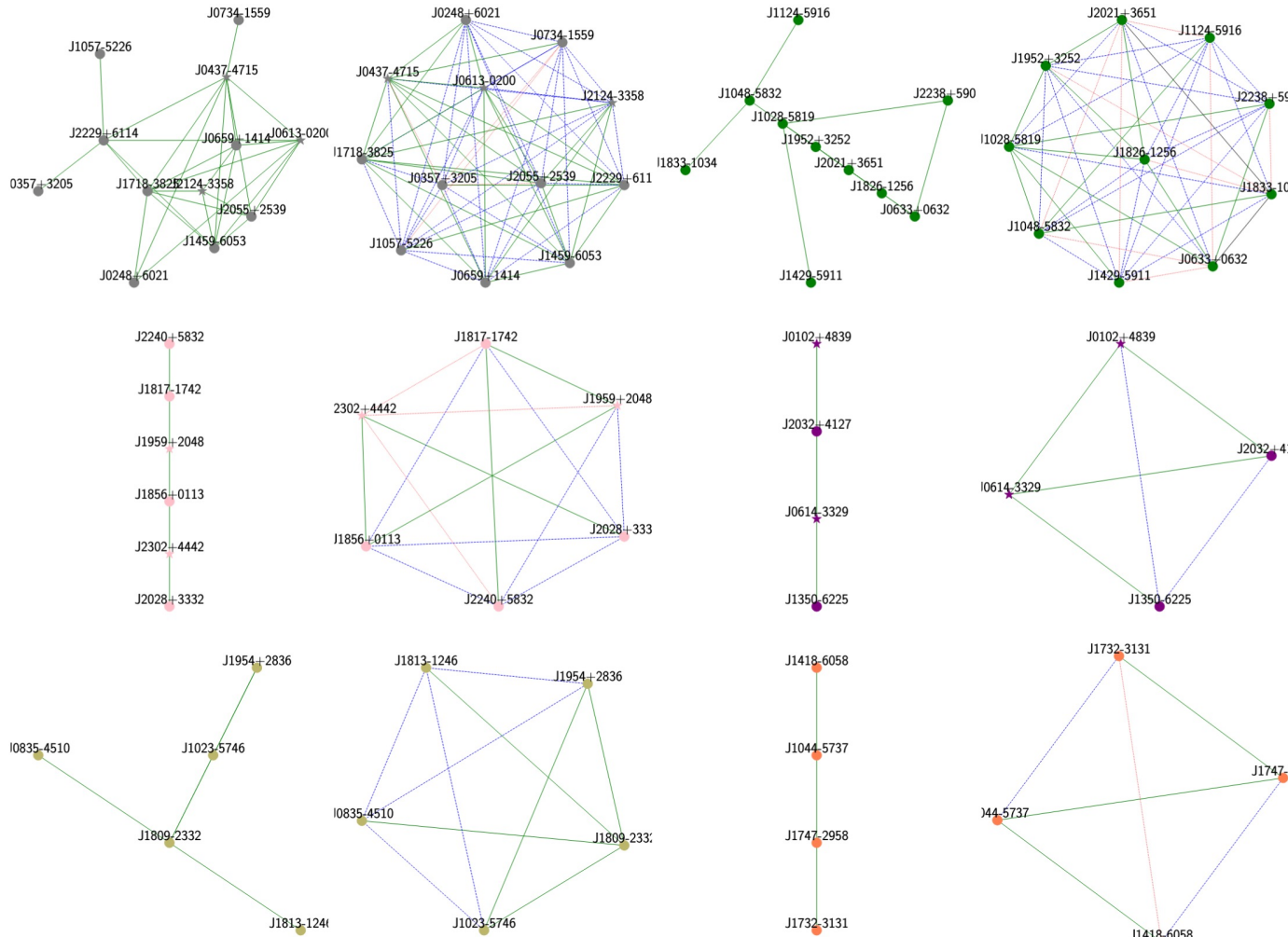
**Left panel:** Distribution of the  $DTW_f$  obtained after comparing the 294 light curves with each other giving rise to 43071 values.

**Center panel:** Distribution of the inverse of  $DTW_f$  (so that the larger DTW are the more similar pairs of light curves). The orange line shows the log-normal distribution, identified as the best fit. The dashed vertical lines denoted with  $1\sigma^*$  (red),  $2\sigma^*$  (blue), and  $3\sigma^*$  (green)

**Right panel:** Distribution of the natural logarithm of  $DTW_f^{-1}$ . The orange line shows the normal distribution. The dashed vertical lines denoted with  $1\sigma$  (red),  $2\sigma$  (blue), and  $3\sigma$  (green) represent the intervals of the distribution according to the known empirical rule.



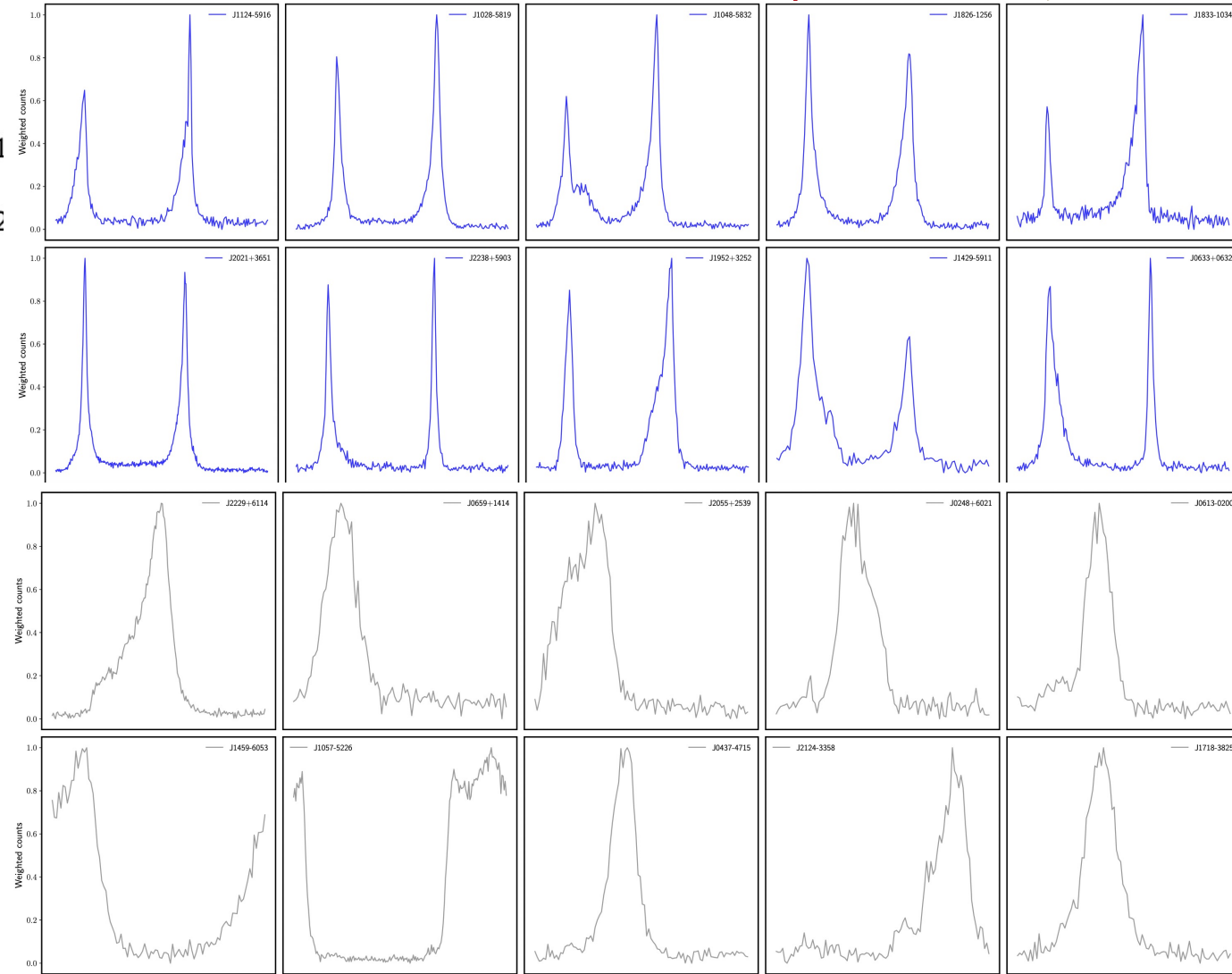
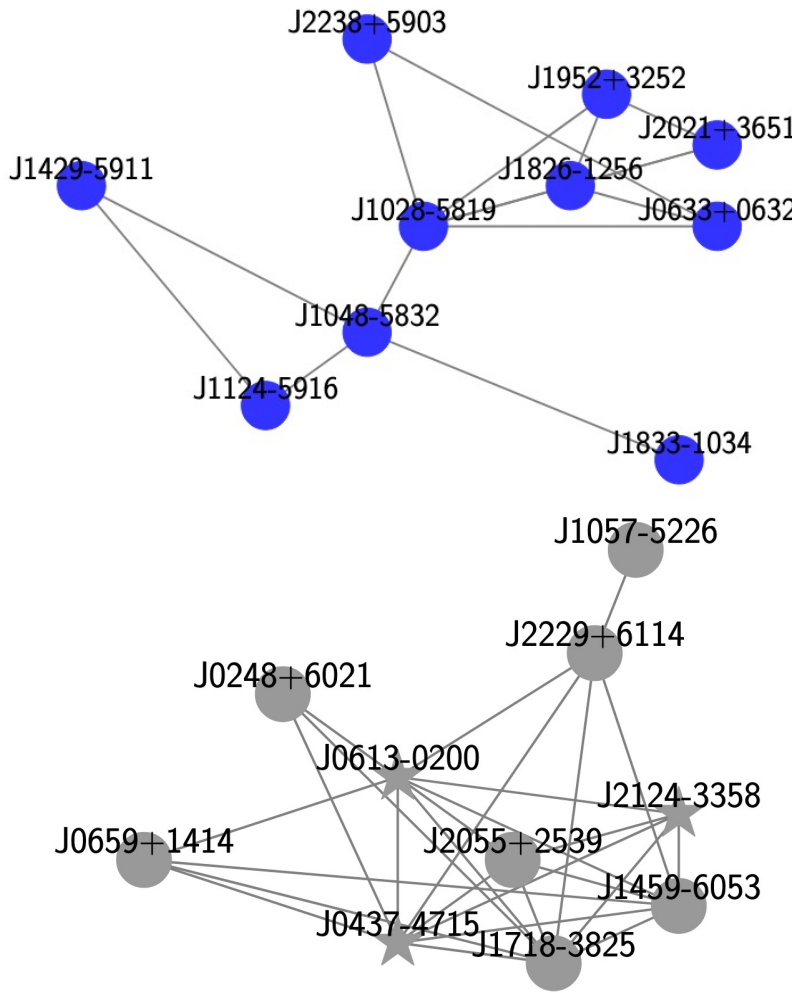
# We can now cluster the light curves using similarity



MSPs are denoted with a star, while pulsars with  $P > 10$  ms are shown with a circle.

Black edges indicate similarity is in the range 1-2 sigma, blue 2-3 sigma and green  $>3$  sigma.

# We can now cluster the light curves using similarity



# Conclusions

# Pulsar similarity can now be phenomenologically quantified

- Light curves look more different than they really are, in many cases.
- Euclidean and DTW provide quantitative estimators of their morphological similarity
- Euclidean distance is sub-efficient:
  - Cannot deal with light curves of different sizes
  - Cannot assign similarity to light curves that really are if they are slightly displaced (e.g., two peaks with slightly different separation)
- Clustering via light curve similarity can be used to compare with physical and spectral properties of pulsars
- Applications of this methodology go much beyond pulsar light curves
  - TPs, magnetar and GRB light curves, etc.



Thank you for your attention



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