



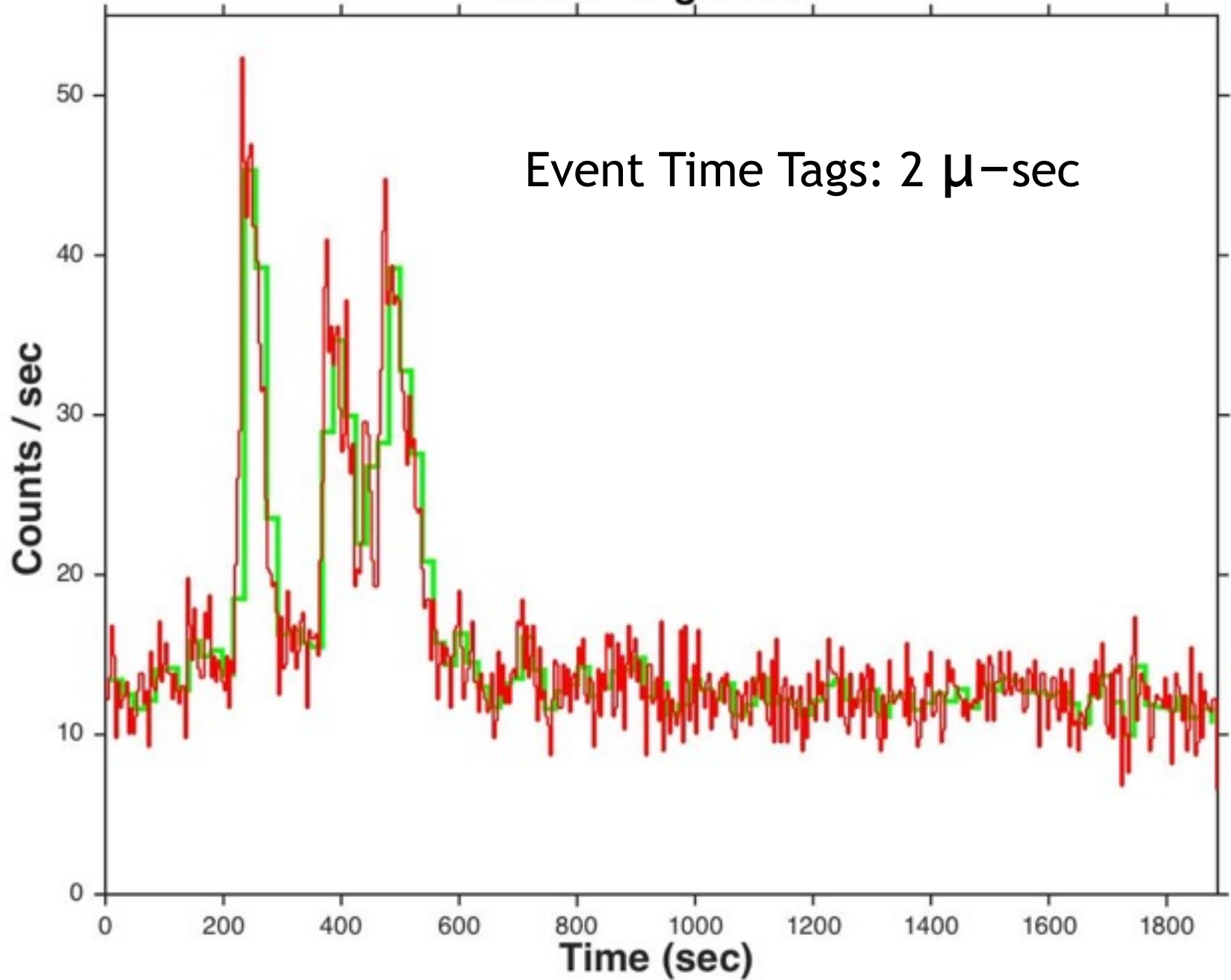
Optimal Segmentation with Pruned Dynamic Programming

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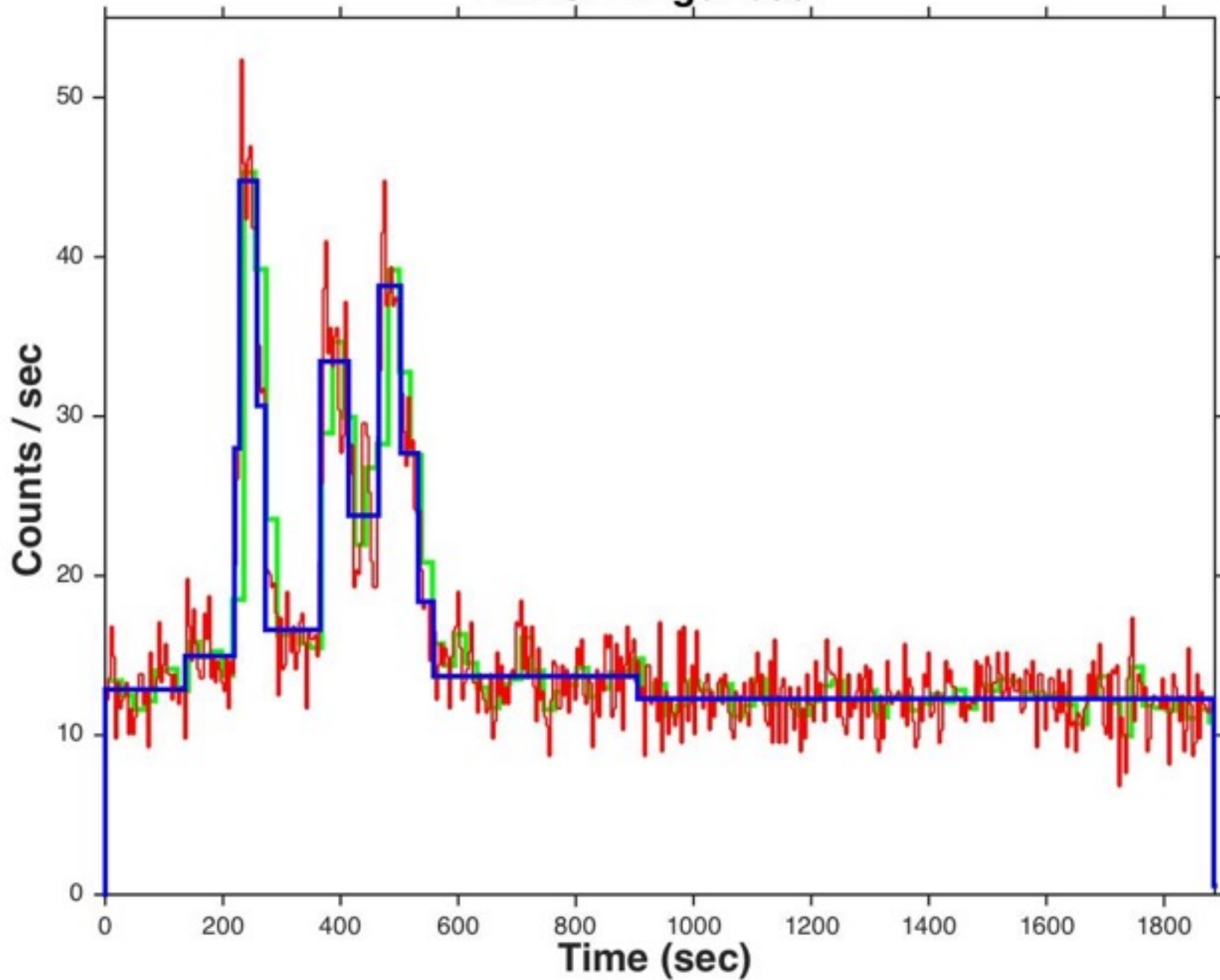
Thanks: Brad Jackson + Students, Jay Norris, Jim Chiang

AI @ SLAC September 26, 2017

BATSE Triger 0551



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What is Optimal Segmentation?

DATA:

- *any signal measurements*
- *sequential in time, space, energy, ...*

MODEL:

- *partition data interval into blocks*
- *block = set of consecutive data points*
- *model of signal in each block*
- *block cost (aka fitness, risk, etc.)*

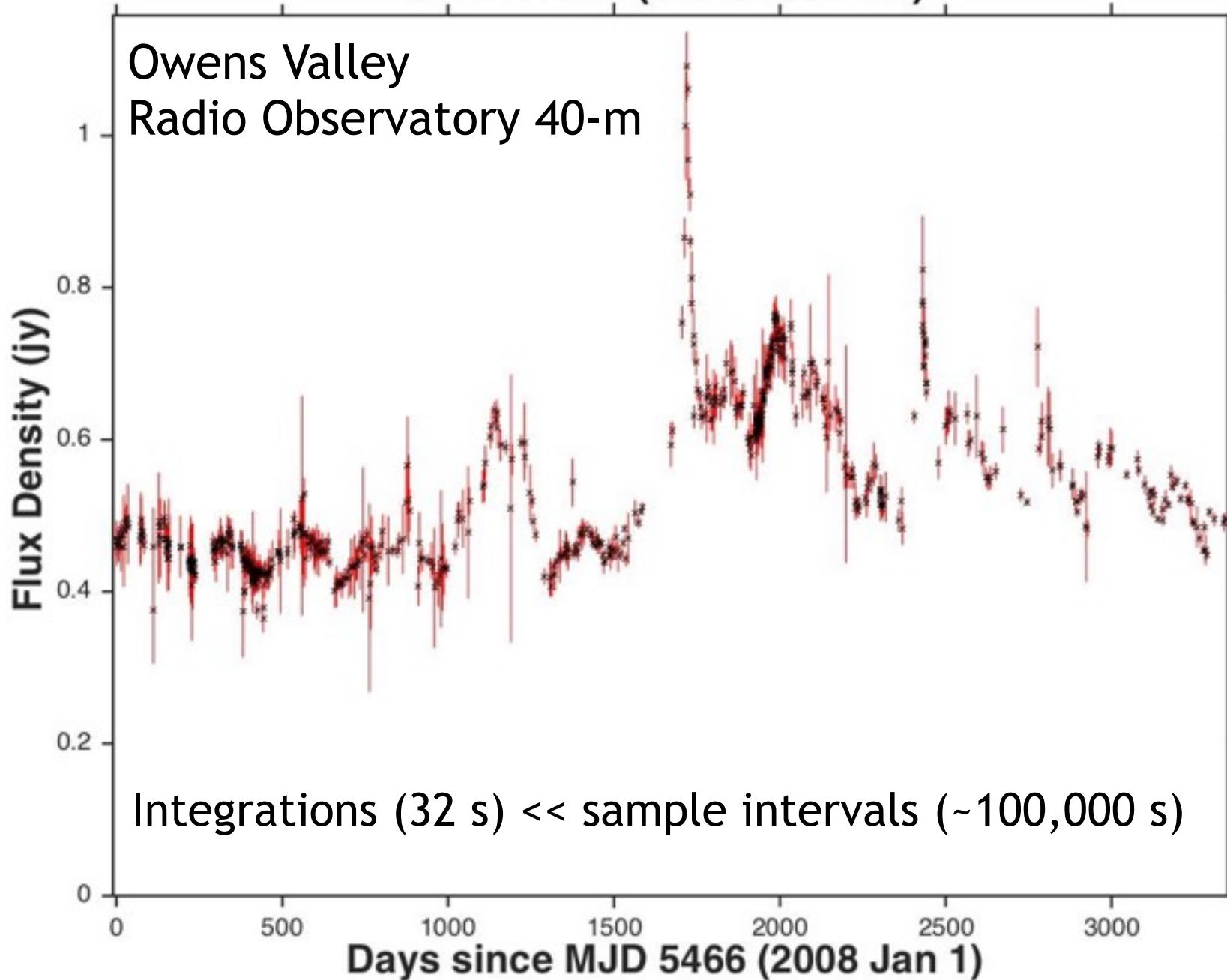
OPTIMIZATION:

- *total model cost = sum of block costs*
- *optimize over all possible partitions ($\sim 2^N$)*

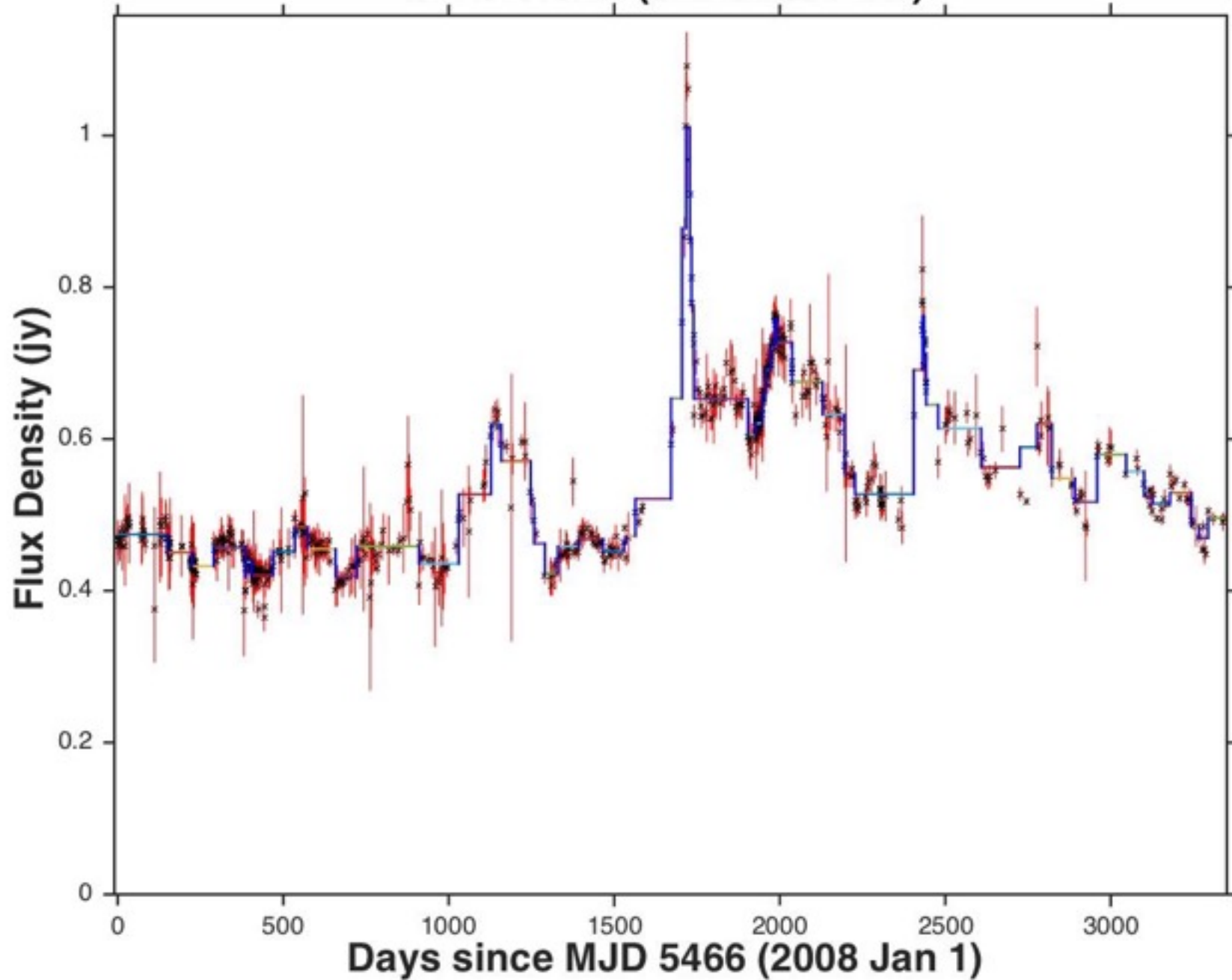
***What good are
Segmented Time Series
Representations? (1)***

***Detect and characterize any and all
statistically significant variability
supported by the data.***

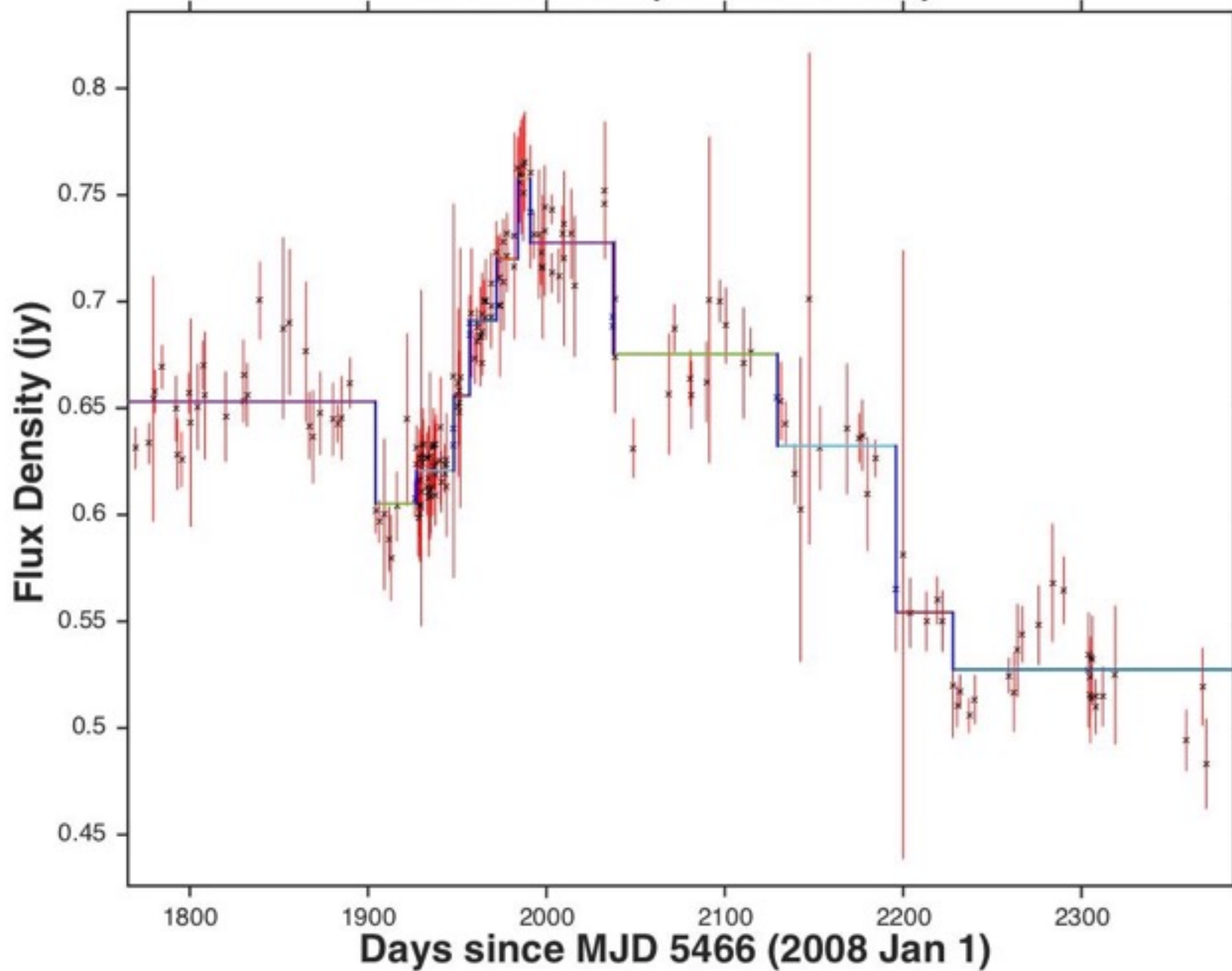
J1104+3812 (Markarian 421)



J1104+3812 (Markarian 421)



J1104+3812 (Markarian 421)



Bayesian Block Generalizations

Any Data Mode:

- *point measurements*
- *time-tagged events*
- *live-time intervals*
- *categorical*
- *gaps, uneven sampling, exposure variation, real-time*
- ... *any data mode*

Any Block Shape:

- *constant*
- *linear*
- *exponential*
- *FRED (fast rise/exponential decay) /DERF*
- ... *any block shape*

Improved speed:

$$O(N^2) \rightarrow \sim N$$

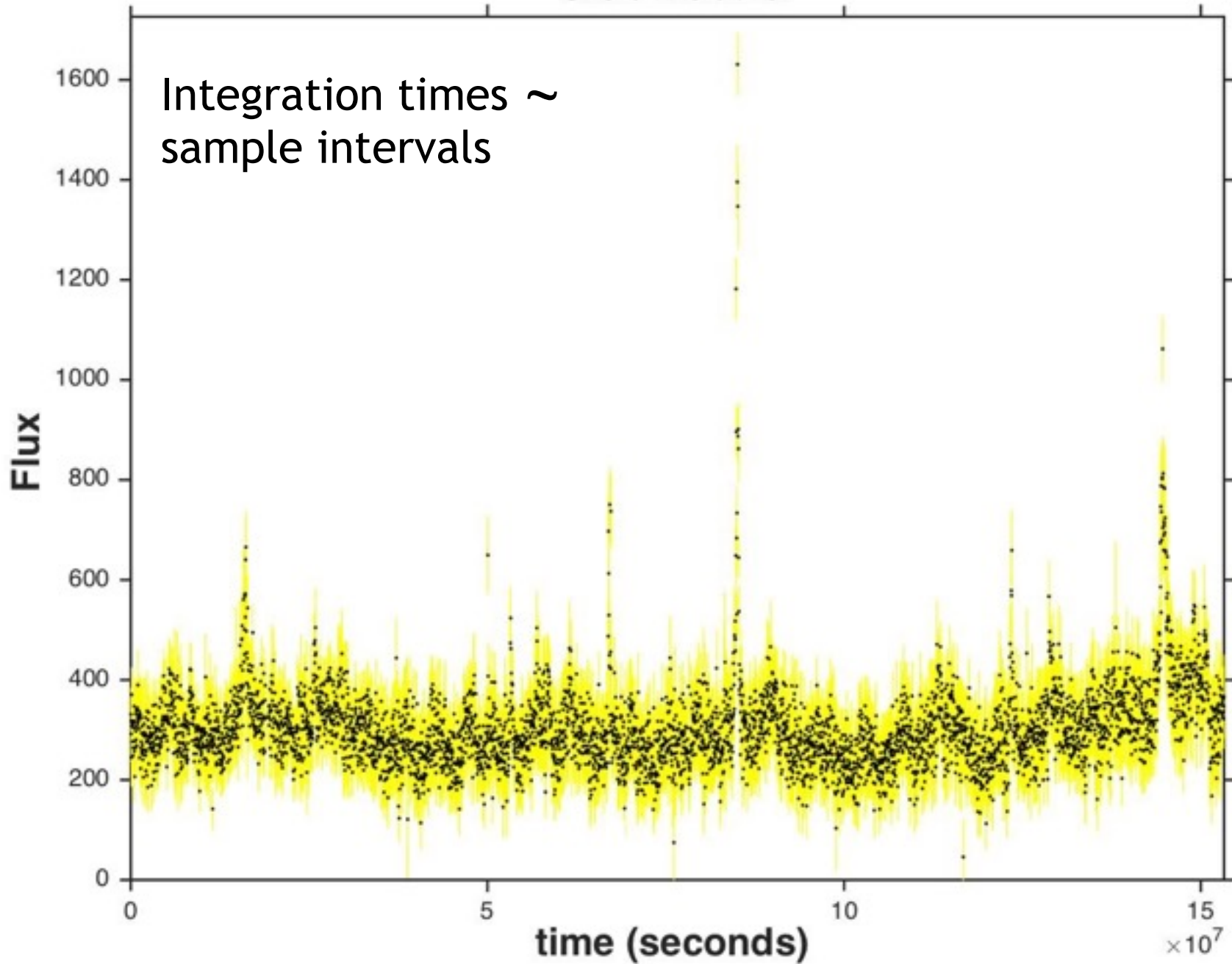
Data Points

... must include whatever information is necessary to compute the *objective function* of a data block
(Typically: time + intensity + error)

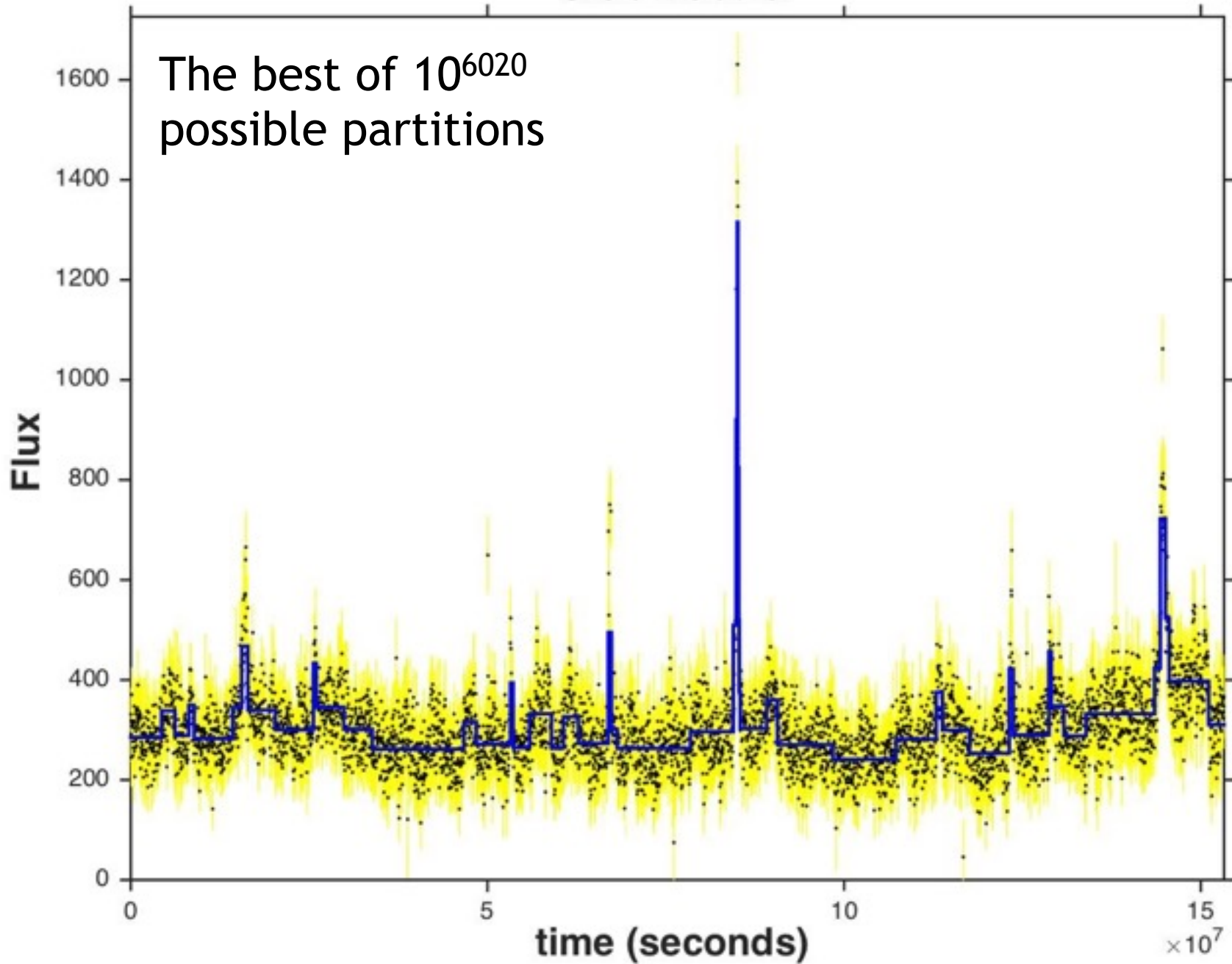
- measurements at a point in time - $X(t_i)$, $\sigma(t_i)$
- measurements averaged over a time interval
- event times (TTE)
- inter-event intervals
- live-time intervals
- auxiliary information (e.g. color)
- categorical (e.g. 0-1, ACGT, ...)

- multivariate (multi-wavelength, multi-messenger)
- mixtures of any of the above
- data on the circle (e.g. angles)
- higher dimension (e.g. 3D galaxy positions)
- ...

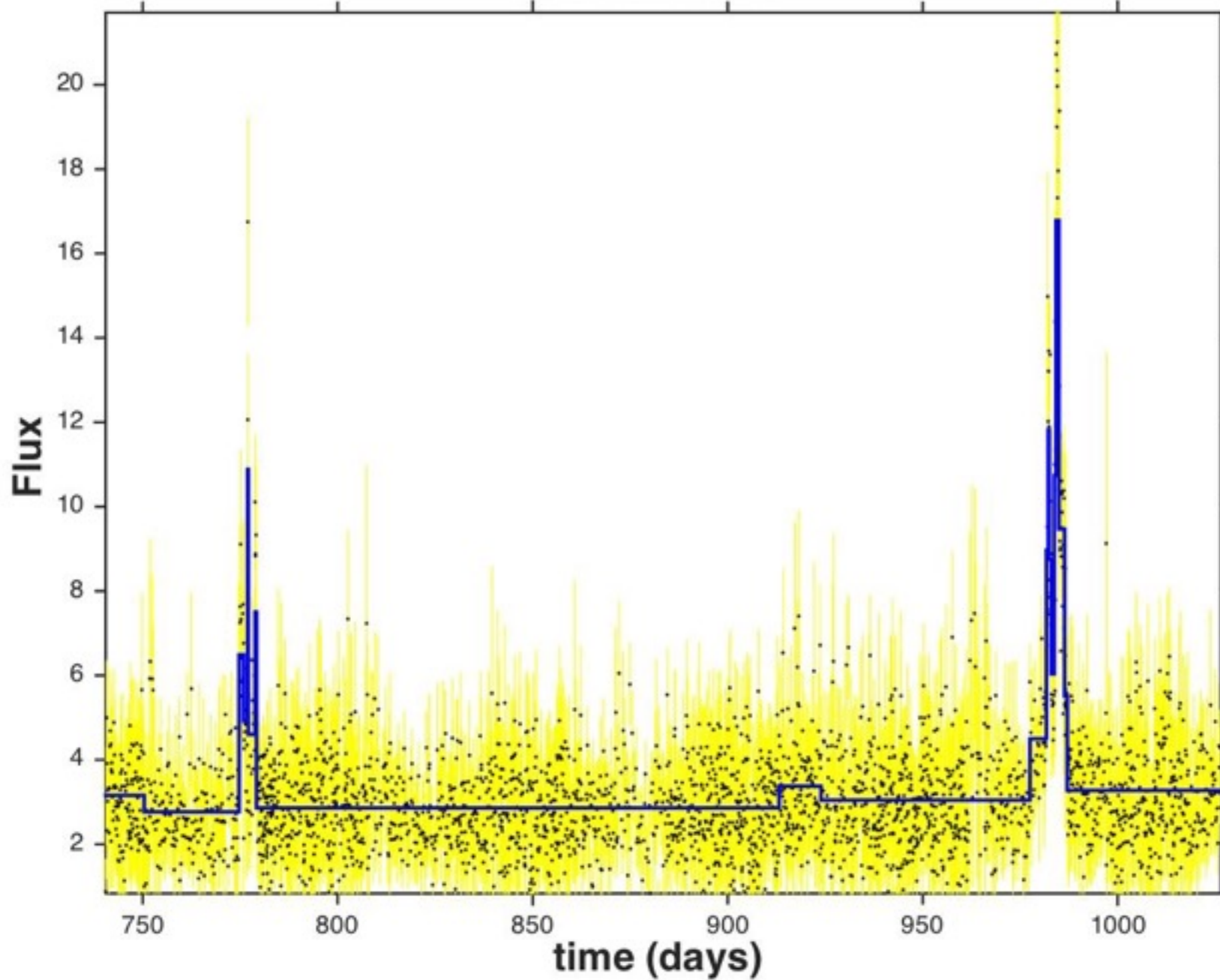
Crab Nebula

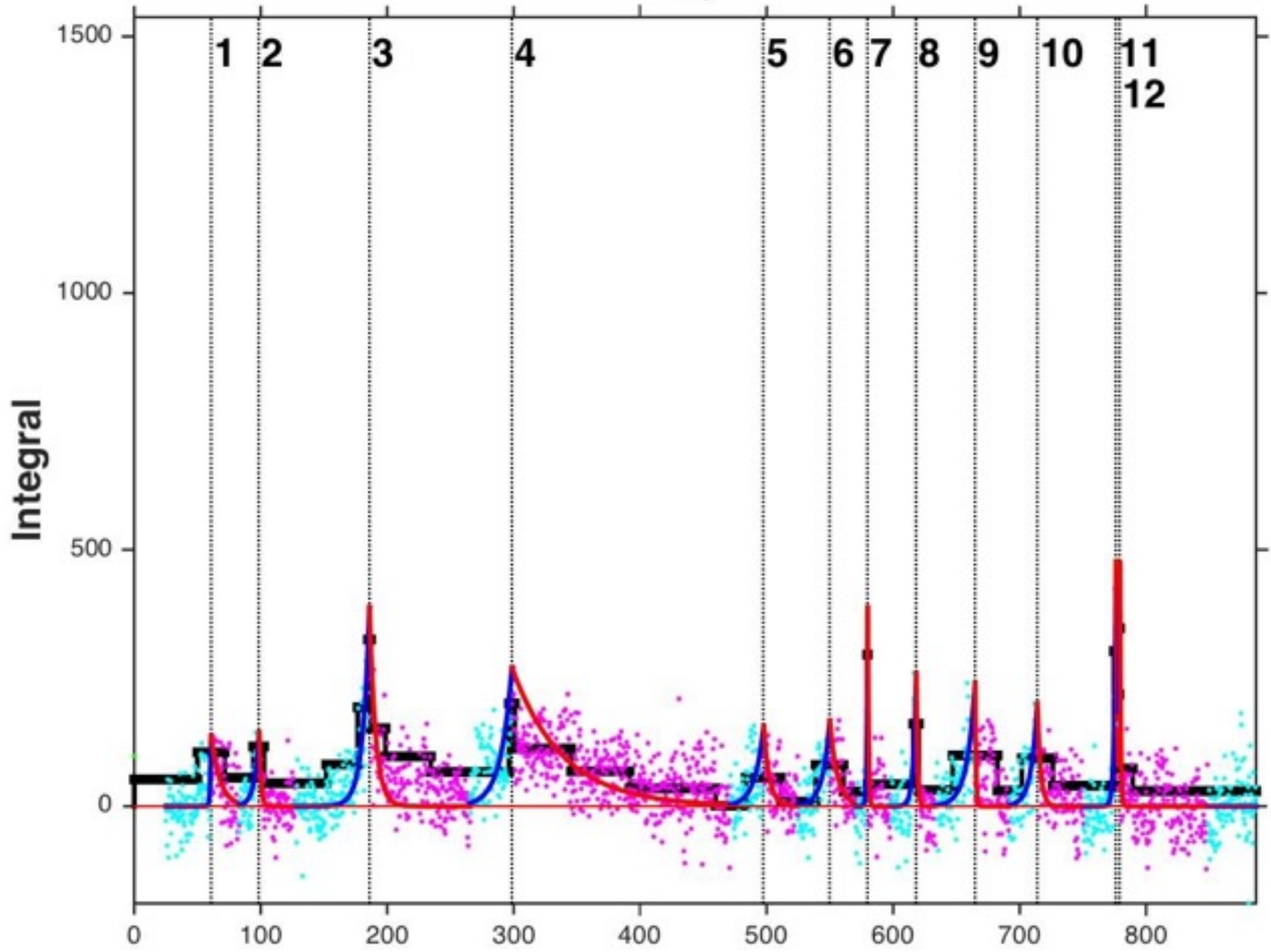


Crab Nebula

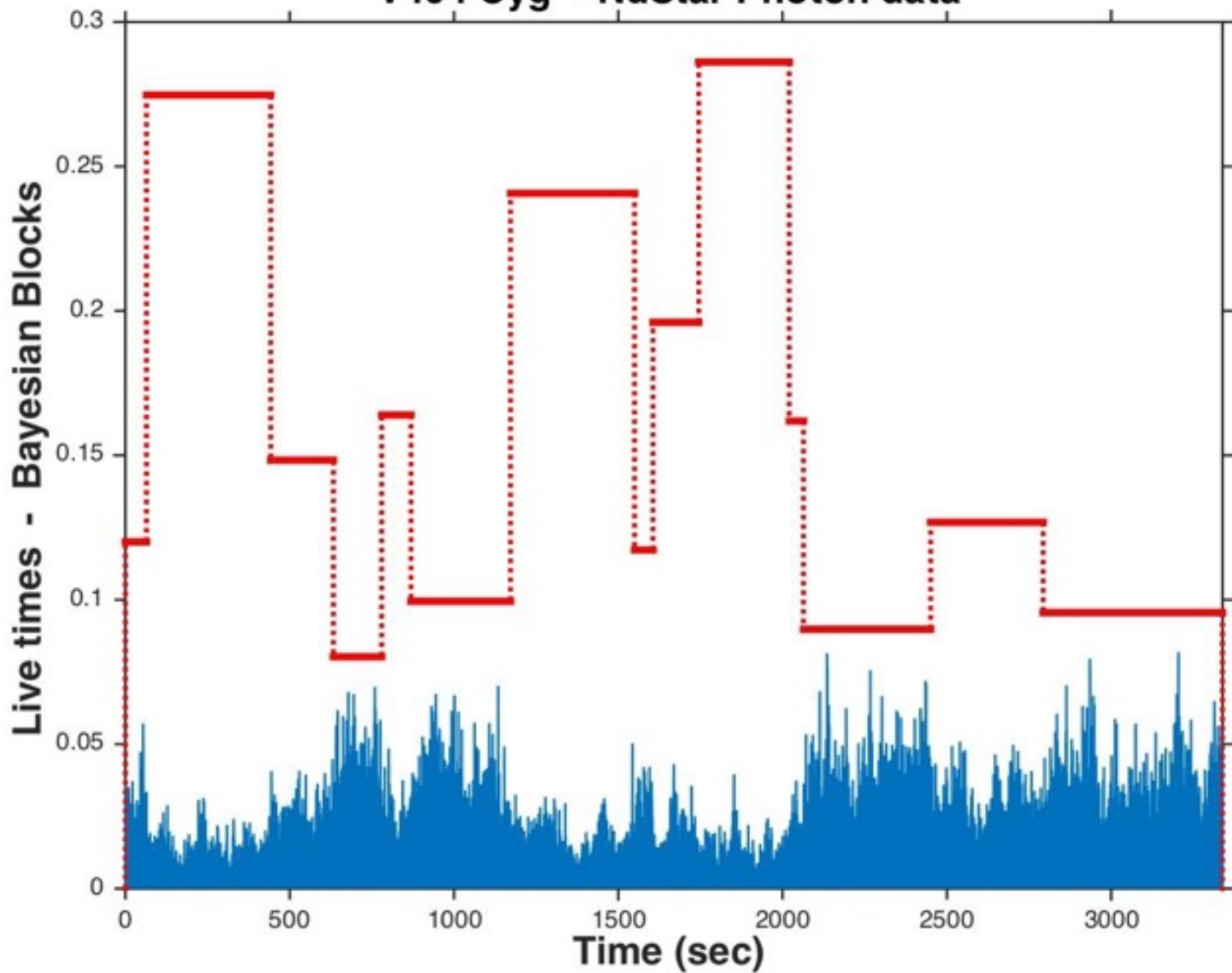


Crab Nebula





V404 Cyg - NuStar Photon data

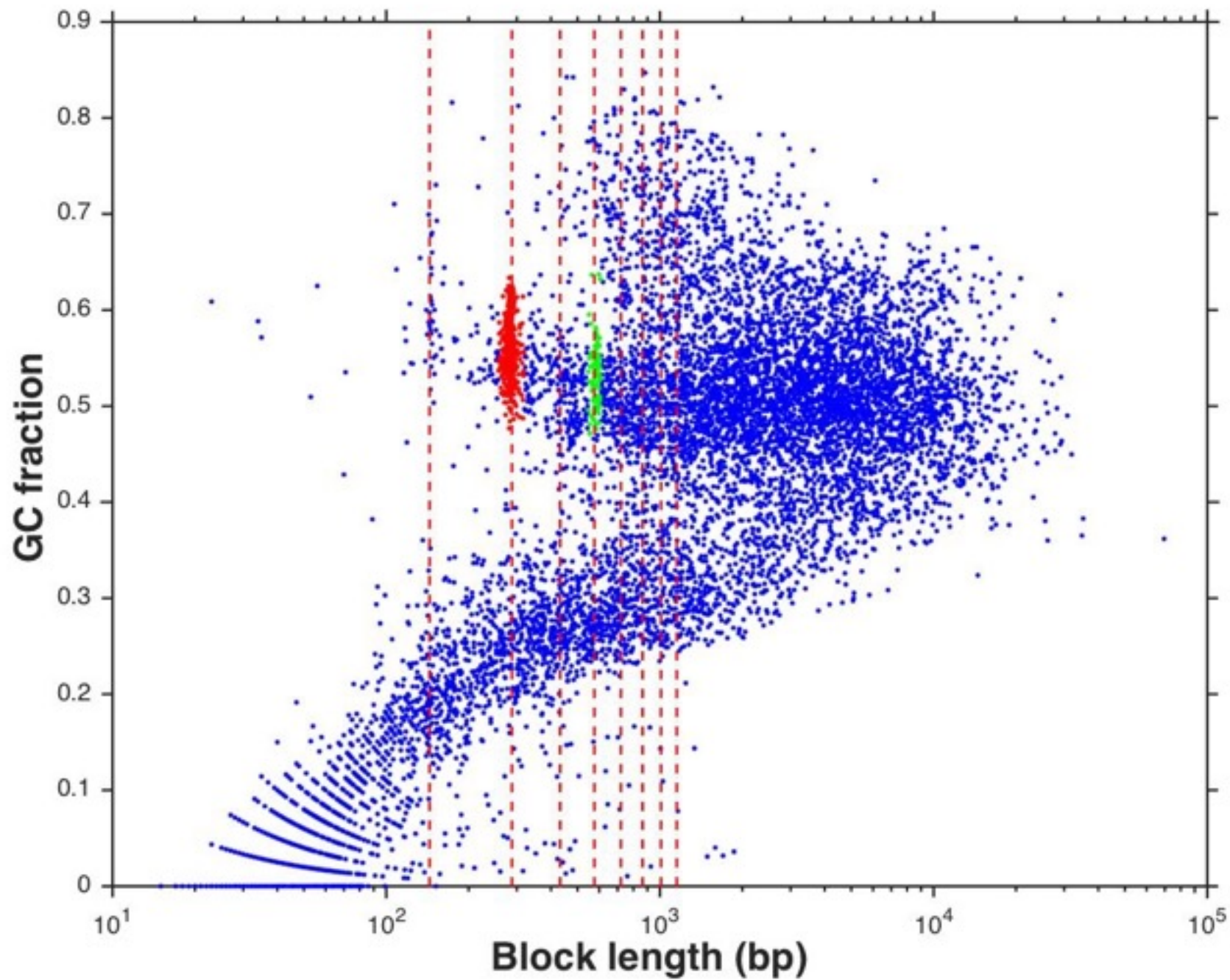


The Human Genome

ATCAGAGACTCGTACCCCGTGCCTGAGCAGTGC
TGAAGAGGGCACTCGTTTGGAAAGGGGGCCCAT
CCCCCGGGACTTCGGACACTCCTGGCTGAAGC
ATAACGTGTAGGCGCTCTAGAGGGCTCGCCTA
CCCGAGCCACATACCAGCGATGATGAGATCG
ACCCGCACCGACGCATGGAGCGCAGGCGTGCT
GCTAGTCGAGGGCACGGTCGCCCCAACAAACG
ACCGCCA ... N ~100,000,000 bp

“GC islands”: G or C \rightarrow 1, A or T \rightarrow 0; two data modes:

- 0-1 on fixed grid
- intervals between 1's



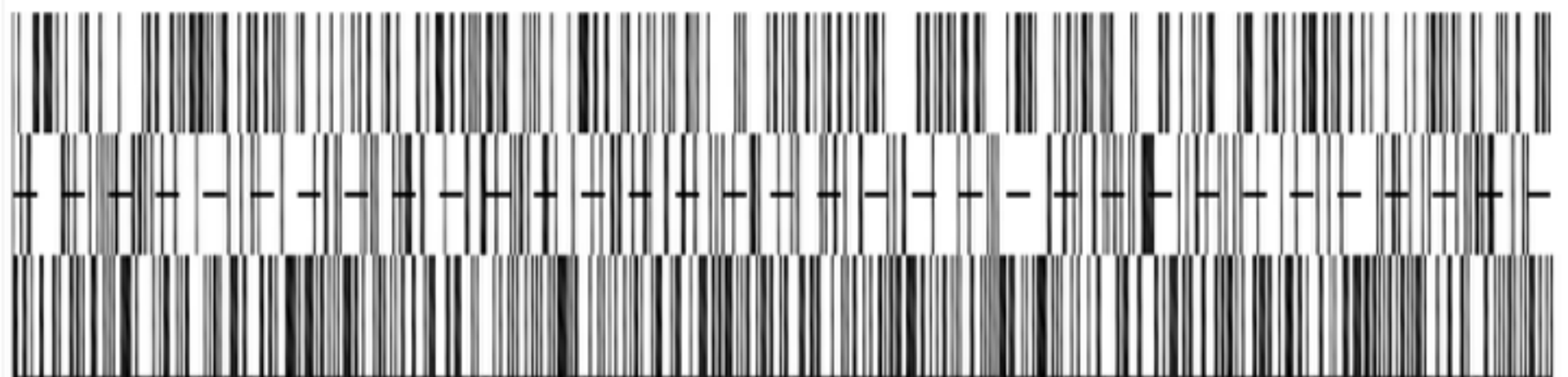
Series #3

Series #2

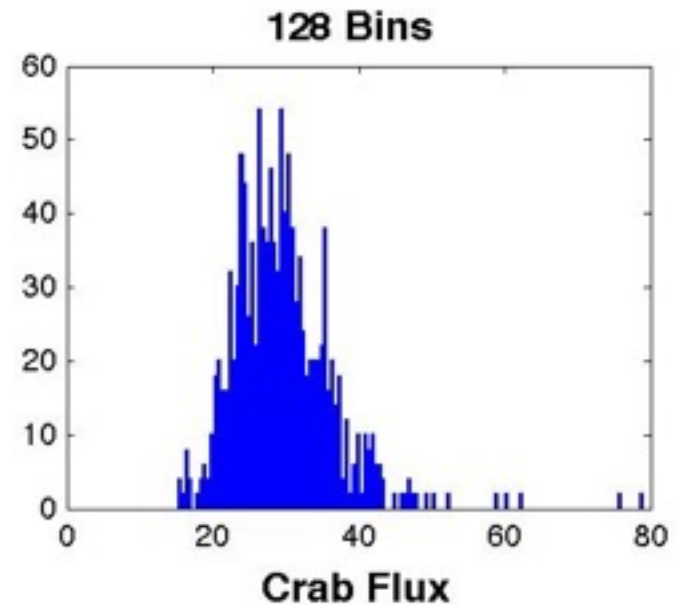
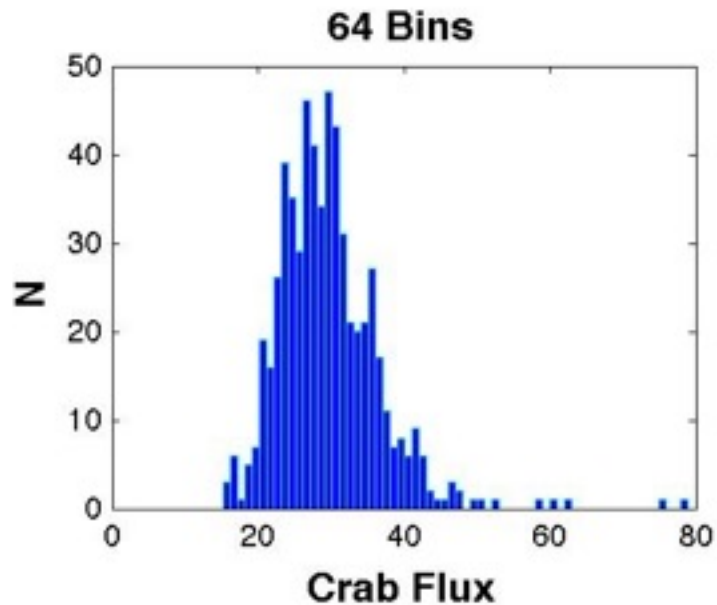
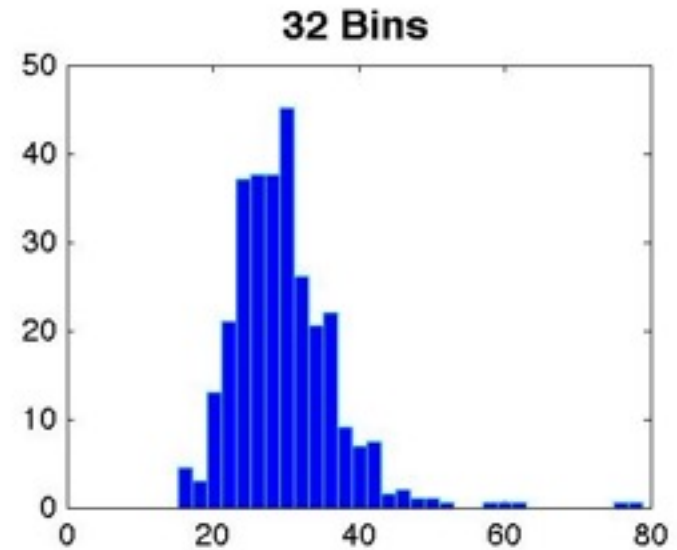
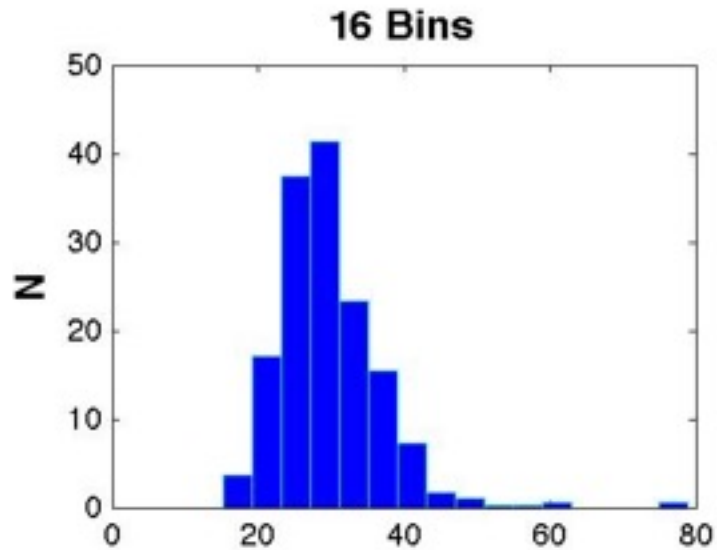
Series #1

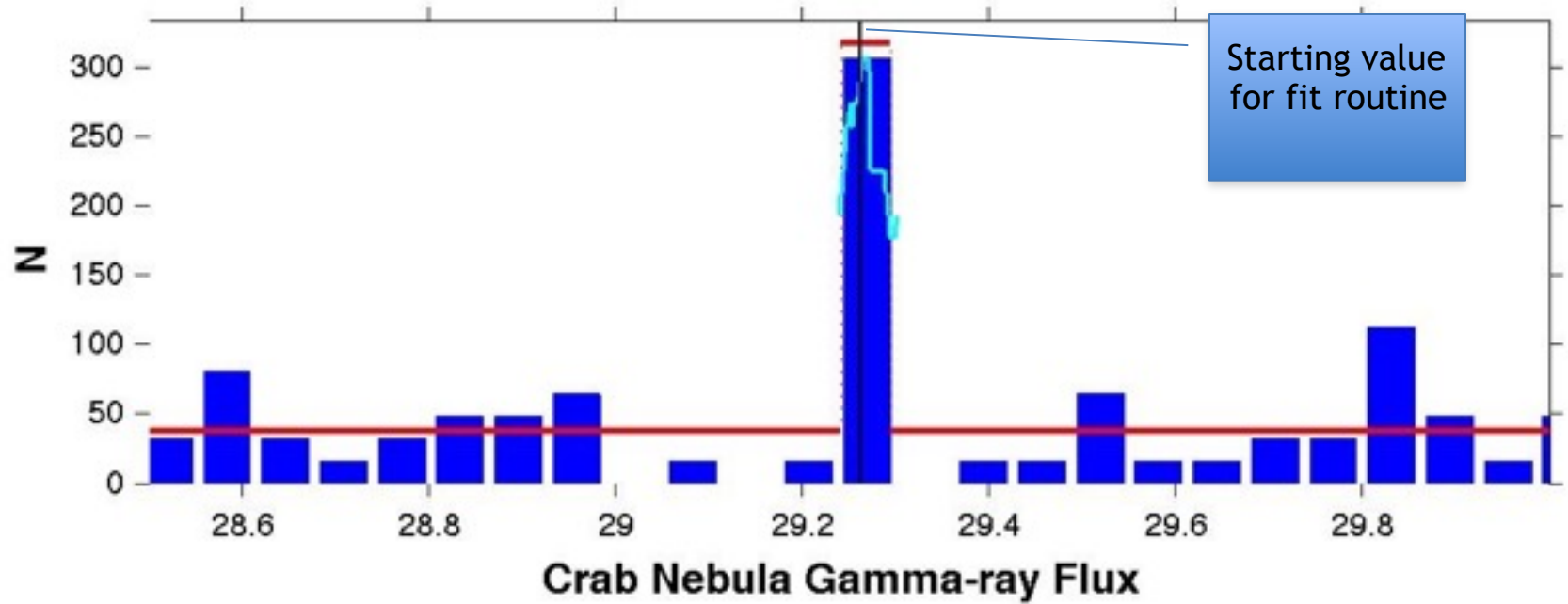
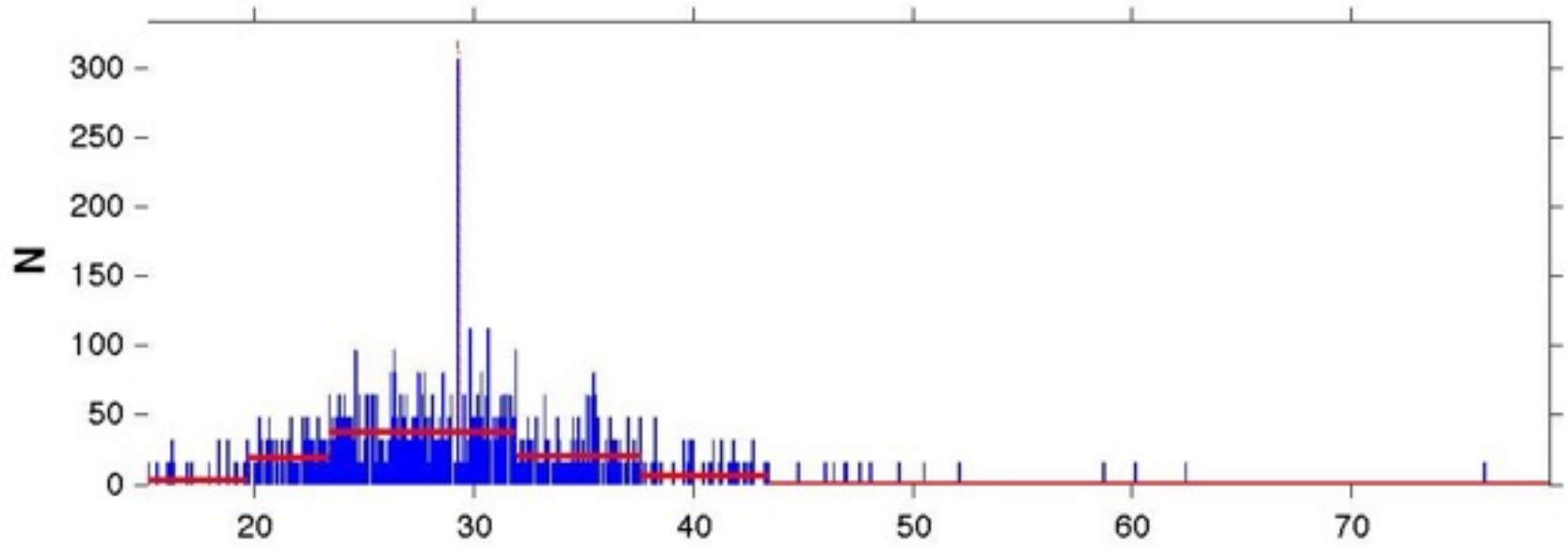
Concatenated Times ...

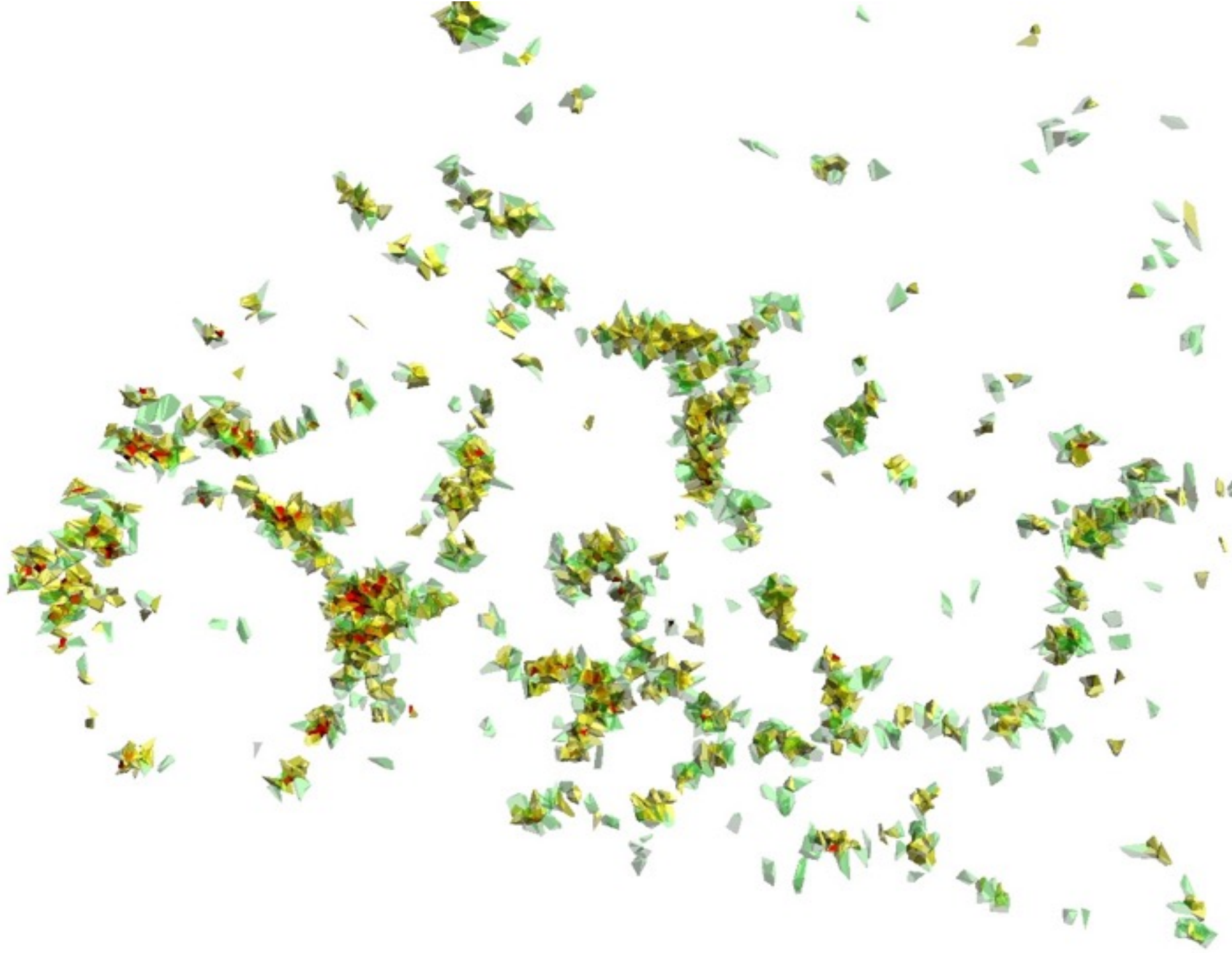
Ordered Times

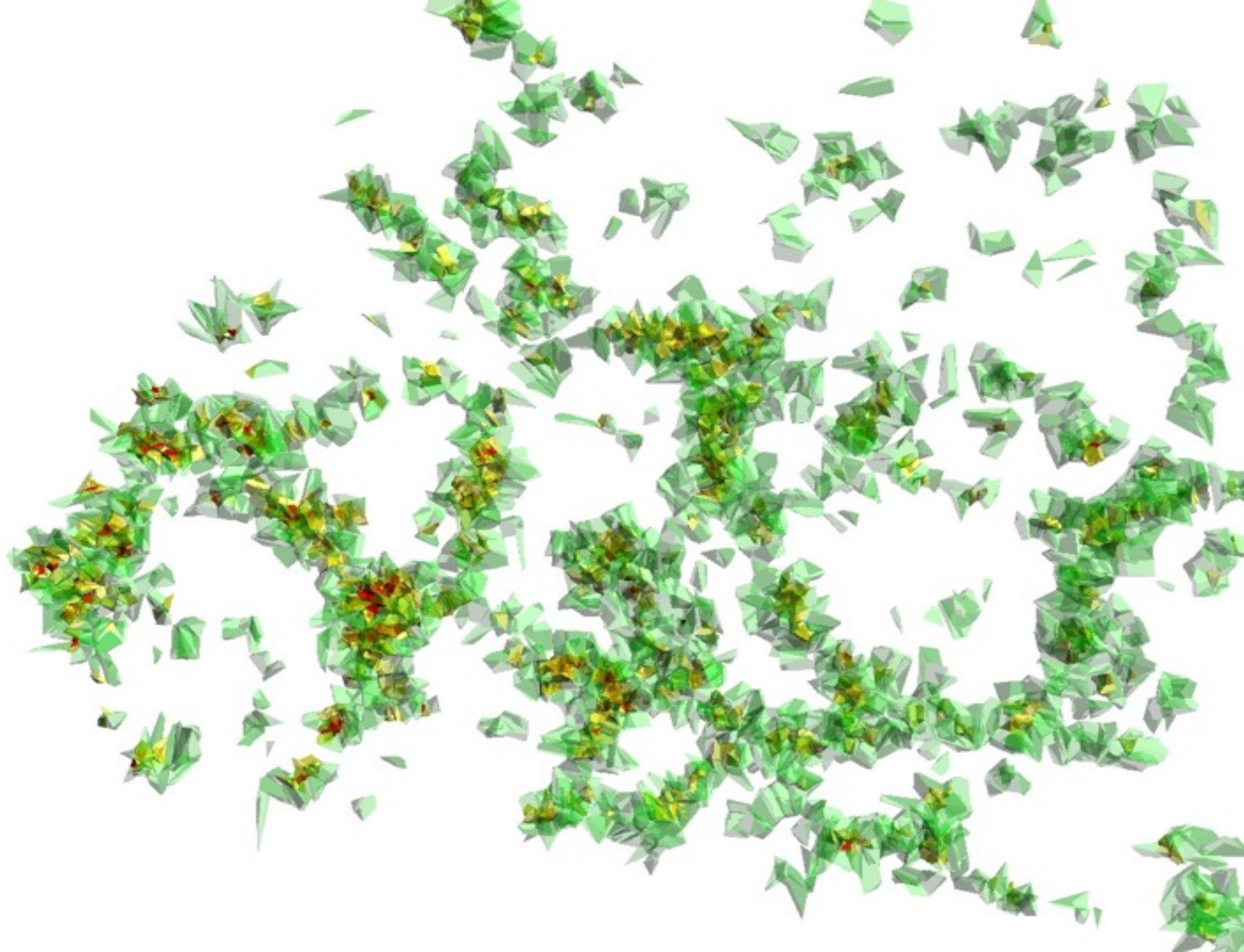


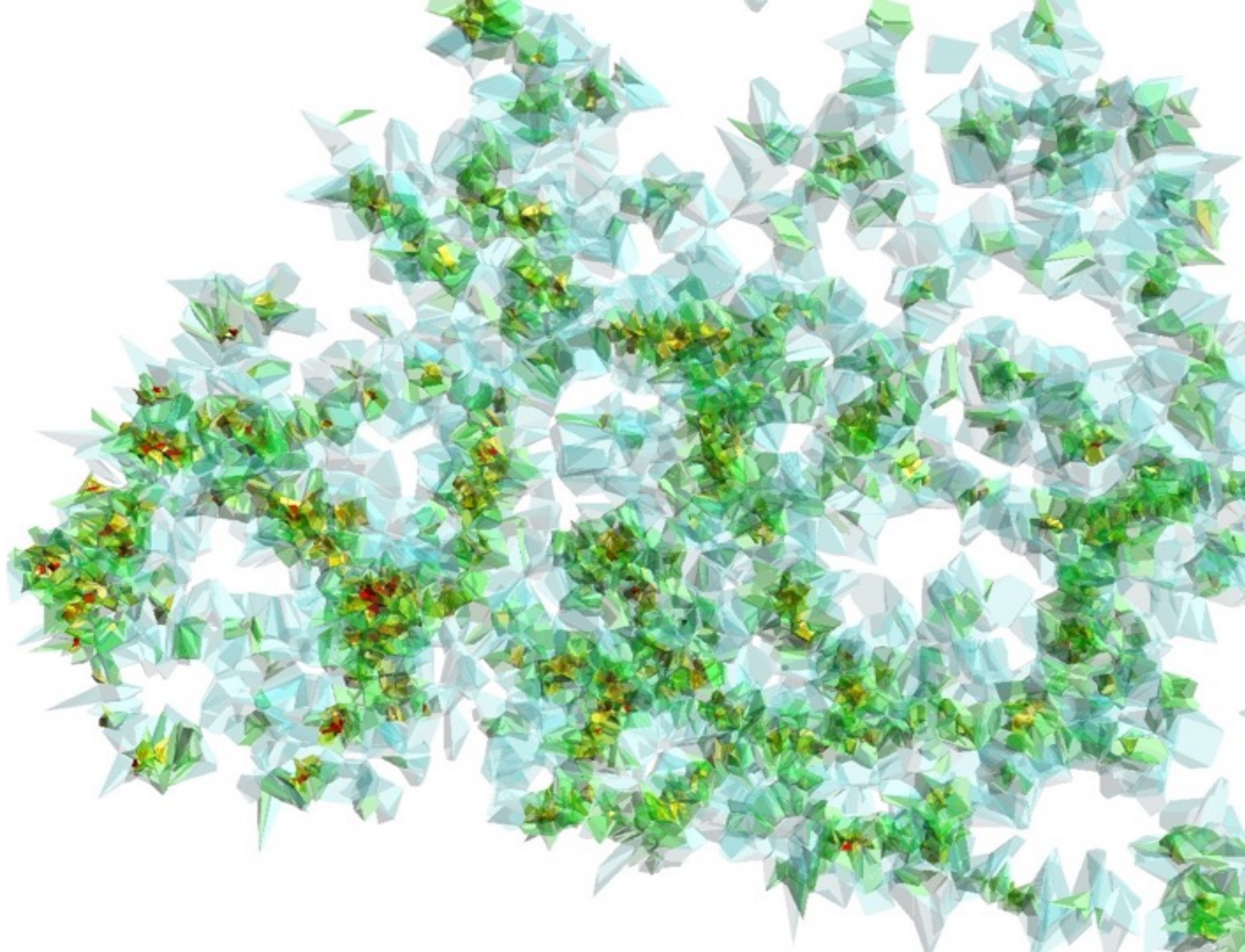
Bayesian Block analysis of event time series is mathematically the same problems as constructing (optimal) [histograms](#)!











Maximum (log) Likelihood Block Costs

Event Data: $C_n = N_n \log(N_n / T_n)$

N_n = number of events in block n

T_n = length of block n

Point Measurements:

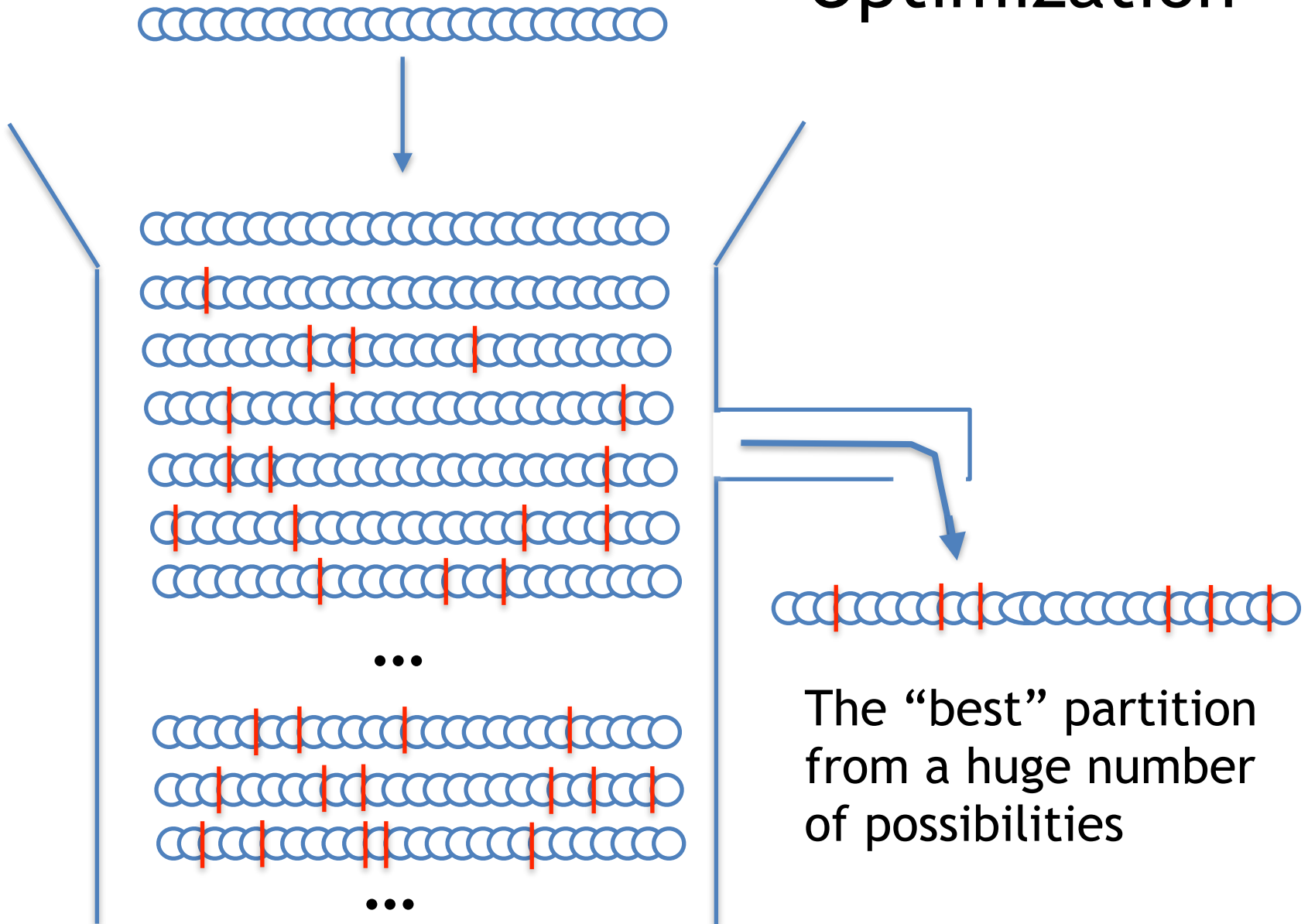
$$C_n = (\sum w_n x_n)^2 / (\sum w_n) \quad \text{variance!}$$

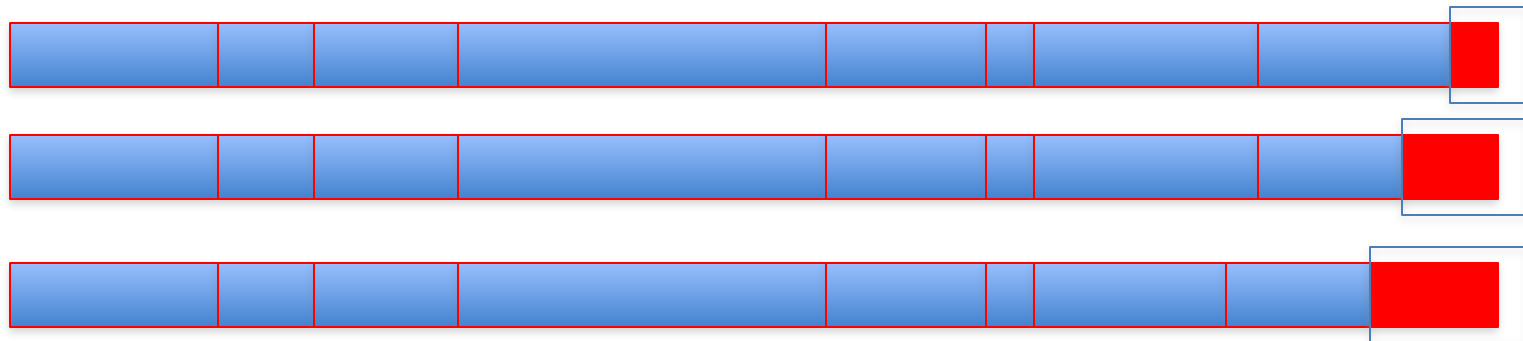
x_n = measured value

w_n = weight = $1 / \sigma_n^2$

Total model cost: $\sum C_n$

Optimization

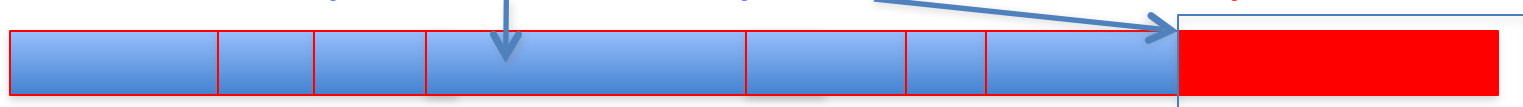




...

Optimum Partition Up To This Point

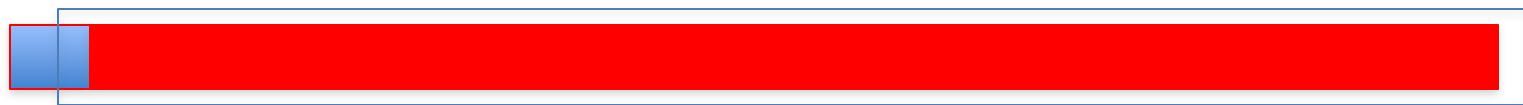
Prospective Last Block



...



...



Dynamic Programming Algorithm

```
for ii = 1:num_points

    sum_num          = sum_num          + num_vec( ii );%count
    sum_len          = sum_len          + len_vec( ii );%length

    % Cost of last block:
    cost_last        = cost_function(sum_num          , ...
                                     sum_len          );

    % Compute and maximize total model cost:
    cost_total       = opt              + cost_last      ;
    [ opt(ii+1) last ] = max( cost_total          );

    last_change( ii ) =          last ; % Store last change-point

end
```

Pruned Dynamic Programming Algorithm

```
for ii = 1:num_points

    unpruned( ii ) = ii; % New point unpruned until found otherwise

    sum_num( unpruned ) = sum_num( unpruned ) + num_vec( ii );%count
    sum_len( unpruned ) = sum_len( unpruned ) + len_vec( ii );%length

    % Cost of last block:
    cost_last( unpruned ) = cost_function(sum_num( unpruned ), ...
                                         sum_len( unpruned ));

    % Compute and maximize total model cost:
    cost_total( unpruned ) = opt( unpruned ) + cost_last( unpruned );
    [ opt(ii+1) last ] = max( cost_total( unpruned ) );

    last_change( ii ) = unpruned( last ); % Store last change-point

    % update pruning
    id_unpruned = find( [ cost_total(unpruned) > opt(ii+1) - ncp_prior ] );
    unpruned = unpruned( id_unpruned );

end
```

What good are Segmented Time Series Representations? (2)

*Detect and characterize,
without bins or smoothing:*

- ◆ *Pulses* (aka “flares”)
- ◆ *Pulse shapes* (including the *Arrow of Time*)
- ◆ *Variability index*
- ◆ *Variability time scales* (min, max, dist, ...)
- ◆ *Transient event triggers* (real-time mode)

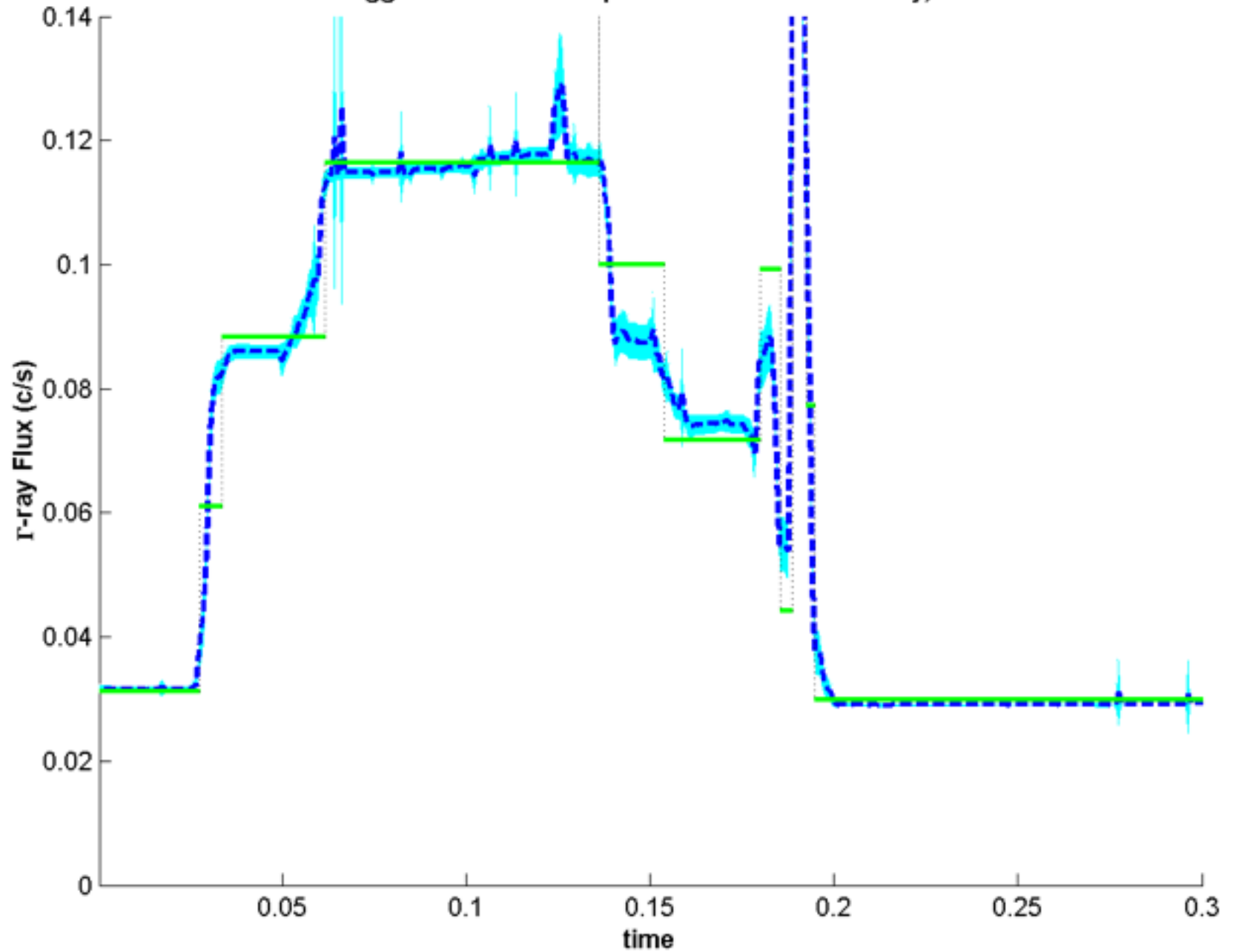
What good are Segmented Time Series Representations? (3)

Implement:

- ◆ *Exploratory Data Analysis*
- ◆ *Time series classification*
- ◆ *Noise suppression*
- ◆ *Visual displays*
- ◆ *Data compression*
- ◆ *Data adaptive histograms*

Thanks

BATSE Trigger 1453: Bootstrap mean and 5σ Uncertainty, ML Blocks



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