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**r statistics  
for time-domain cross correlation  
on burst candidate events**

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# Preface

- The burst analysis pipeline uses Event Trigger Generators (ETGs) to flag times when “something” occurs (burst candidates).
- Triggers from different interferometers are brought together in coincidence.
- All cuts in the pipeline are generous, in order to preserve sensitivity: we don't really know what we are looking for...
- We want to use the full power of a coincident analysis:
  - » What confidence can we put on the coincidence candidate?
  - » Are the waveforms consistent?
- The ETG outputs ( $\Delta t$ , BW, amplitude/power/SNR) do not provide enough information, we need to go back to the time series

# Cross correlation in time domain

- Typical duration of coincident events from the burst pipeline: tens of seconds from SLOPE, 0.5-1 sec from TFCLUSTERS.
- Load 5 sec of data from the two interferometers, 2 sec before event start
- High-pass at 100 Hz
- **Whitening/line removal**: train an **adaptive filter** (linear predictor filters – studied by Shourov Chatterji) over the first second of loaded data, apply to the rest. This is fundamental to bypass the problem of non-stationary lines.
- Use the **r-statistics** over “small” time intervals and implement time lags to evaluate the linear cross correlation.

# Pearson's r statistics

$$r = \frac{\sum_i (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_i (x_i - \bar{x})^2} \sqrt{\sum_i (y_i - \bar{y})^2}}$$

## NULL HYPOTHESIS:

the two (finite) series  $\{x_i\}$  and  $\{y_i\}$  are uncorrelated

⇒ Their linear correlation coefficient (**Pearson's r**) is normally distributed around zero, with standard deviation =  $1/\sqrt{N}$

where N is the number of points in the series ( $N \gg 1$ )

$S = \text{erfc}(|r| \sqrt{N/2})$  = double-sided significance of the correlation  
probability that  $|r|$  should be larger than what measured, if  $\{x_i\}$  and  $\{y_i\}$  are uncorrelated

$C = -\log_{10}(S)$  = confidence that the null hypothesis is FALSE  
(that is: confidence that the two series are in fact correlated)

Reference: Numerical Recipes in C

# What does a large confidence mean?

Confidence  $C = 1 \Leftrightarrow$  significance  $S = 0.1$  (10%)

3 different ways to say the same thing:

$\Rightarrow$  10% probability that the null hypothesis is true, OR

$\Rightarrow$  90% probability the hypothesis of no correlation is false (events are correlated) OR

$\Rightarrow$  10% probability this is a false coincidence

If we can assign a confidence to the coincident event pair, we can define a cut on it. The cut defines the false probability in the analysis.

|                |   |                     |                         |
|----------------|---|---------------------|-------------------------|
| confidence=1   | $\Rightarrow$ significance=0.1 (10%)    | $\Rightarrow$ 90%   | correlation probability |
| confidence=1.3 | $\Rightarrow$ significance=0.05 (5.0%)  | $\Rightarrow$ 95%   | correlation probability |
| confidence=1.6 | $\Rightarrow$ significance=0.025 (2.5%) | $\Rightarrow$ 97.5% | correlation probability |
| confidence=2   | $\Rightarrow$ significance=0.01 (1.0%)  | $\Rightarrow$ 99%   | correlation probability |
| confidence=3   | $\Rightarrow$ significance=0.001 (0.1%) | $\Rightarrow$ 99.9% | correlation probability |

# Assign a confidence to the pair of coincident events (burst candidate)

We suspect a burst happened in a 0.5 sec time window. We do not know how long the burst is (1 ms?) WHEN it happens within the interval and at what delay between sites.

Let's shift one of the two series by one data point at a time and calculate a series of:

coefficients  $r_k$   
 significances  $S_k$   
 confidences  $C_k$

$$r_k = \frac{\sum_i (x_i - \bar{x})(y_{i+k} - \bar{y})}{\sqrt{\sum_i (x_i - \bar{x})^2} \sqrt{\sum_i (y_{i+k} - \bar{y})^2}}$$

...and look for a peak in confidence.

In order to reduce fluctuations, decimate the confidence series by 4.

Max confidence = confidence in the correlation of the event pair

Time shift for max confidence = delay between IFOs

**Magic number: 10 ms**

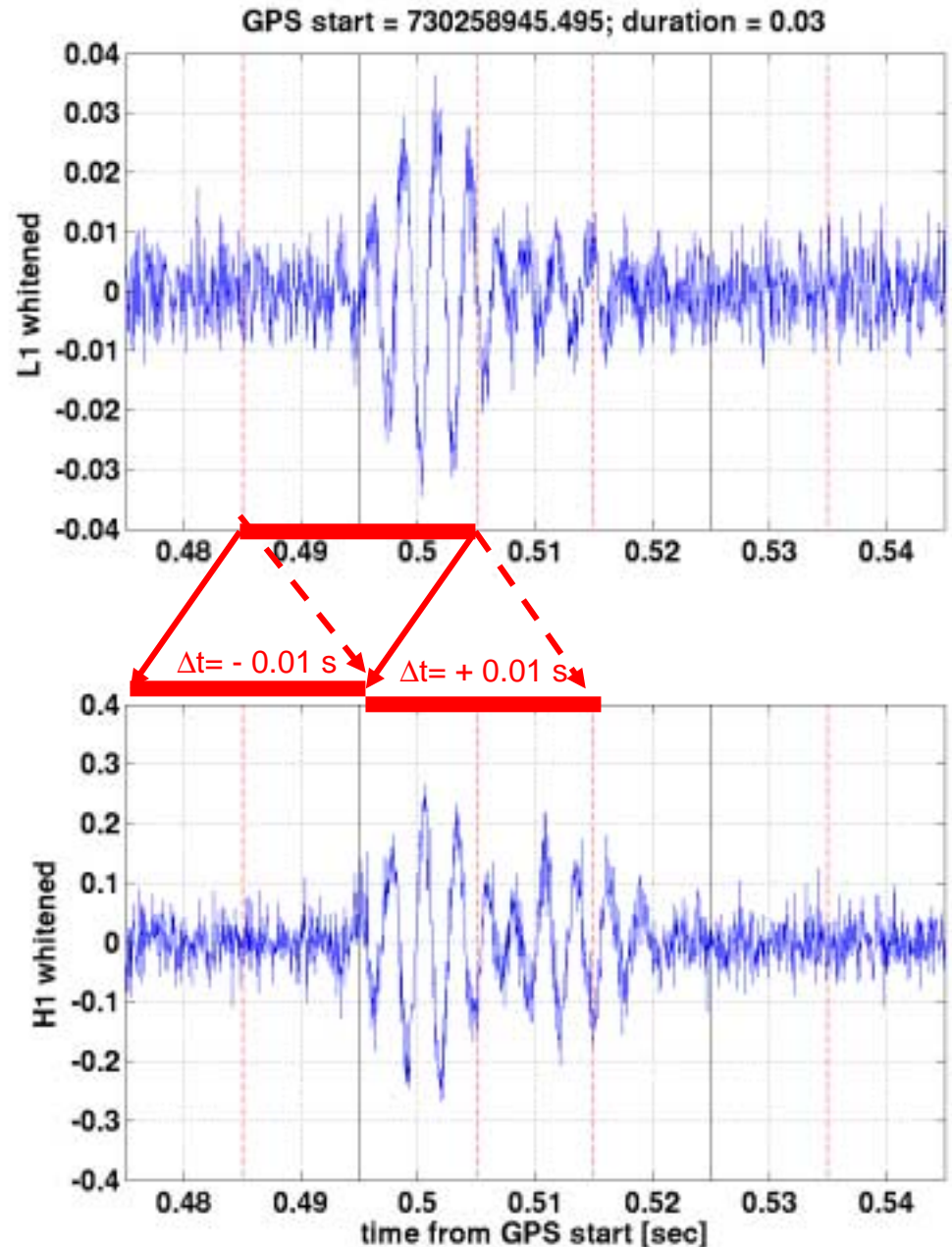
Example: S2 hardware injection  
From Feb 25, 2003 – H1-L1  
(sine gaussian, 361Hz, Q=9)

Divide in 20 ms data segments,  
shift by +/- 10 ms and find  $C_{\max}$   
Move 10 sec forwards and repeat  
the iteration

This way we allow for bursts  
with separation up to 10 ms at  
any point within the trigger  
duration.

20 ms intervals work well for  
burst injections – needs some  
tuning, though.

If the segment is much larger,  
the correlation washes out.



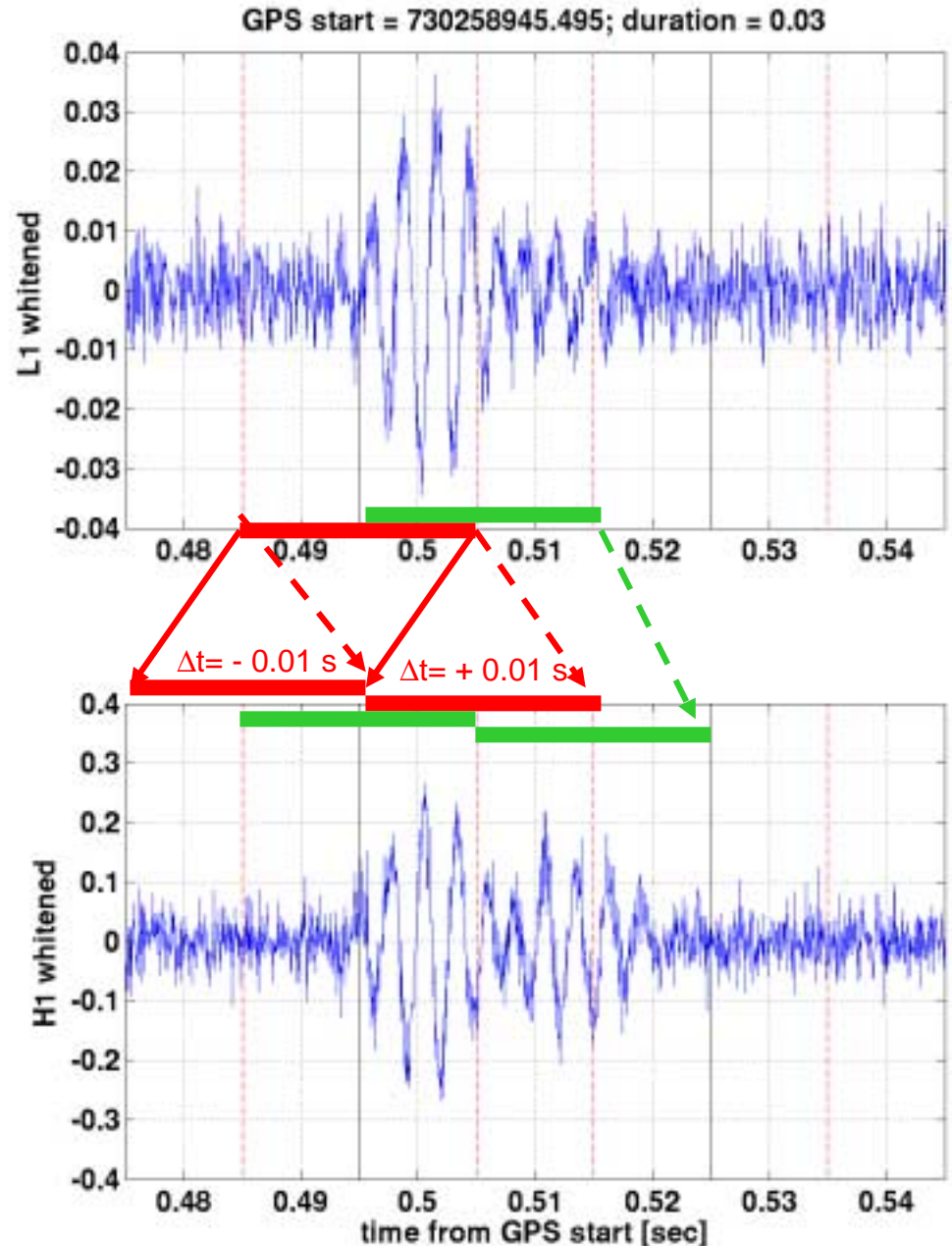
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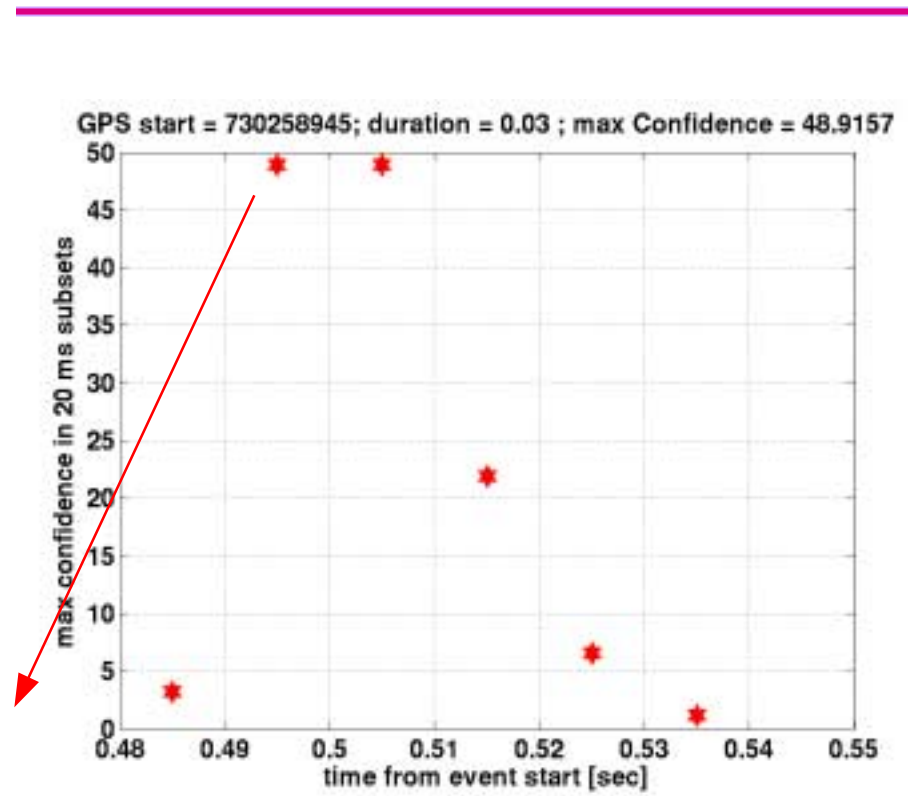
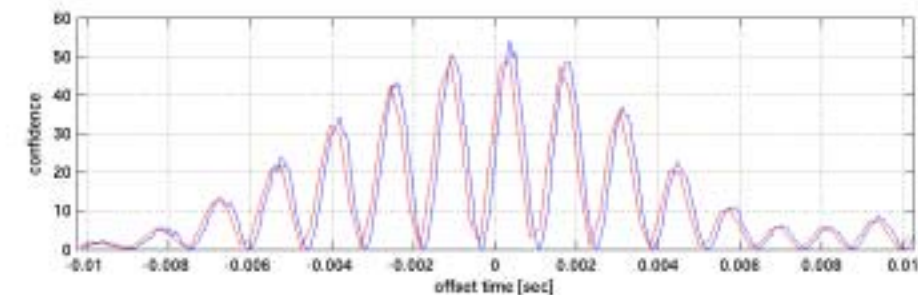
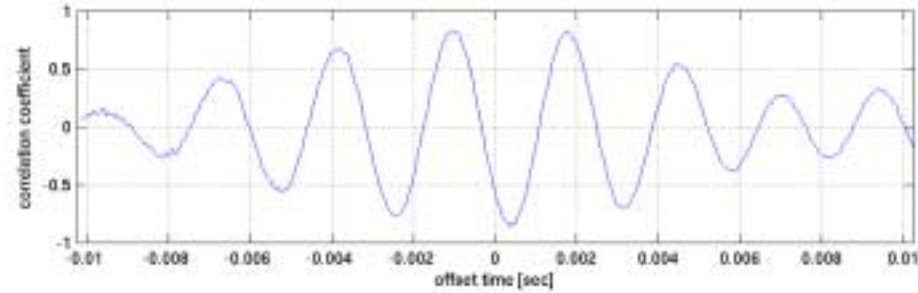
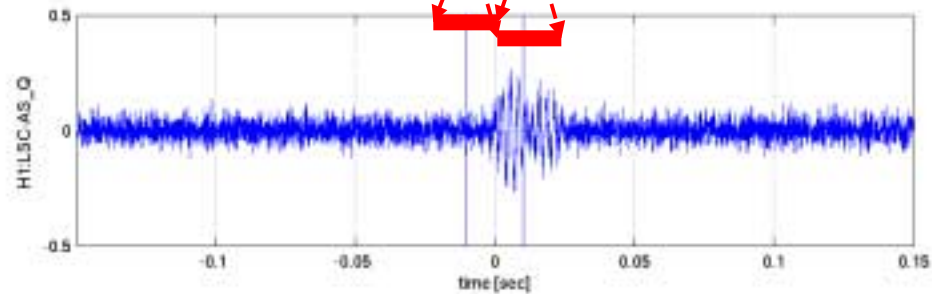
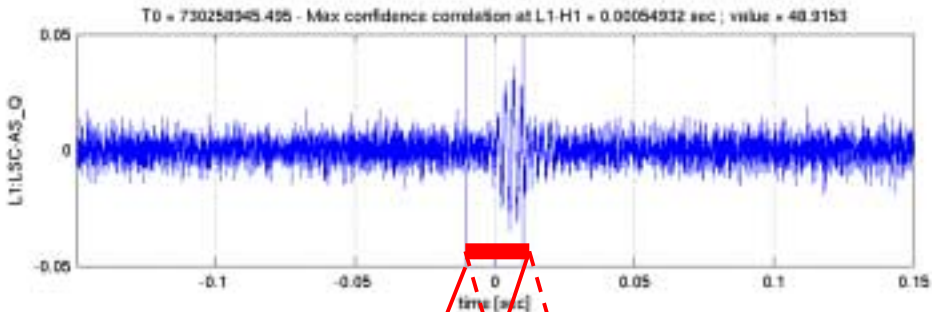
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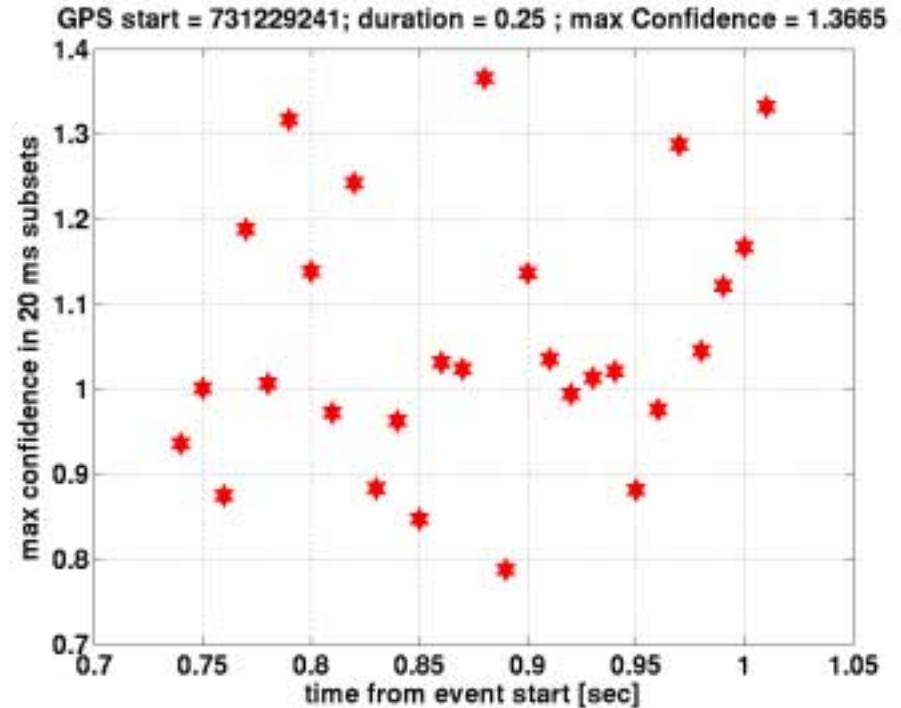
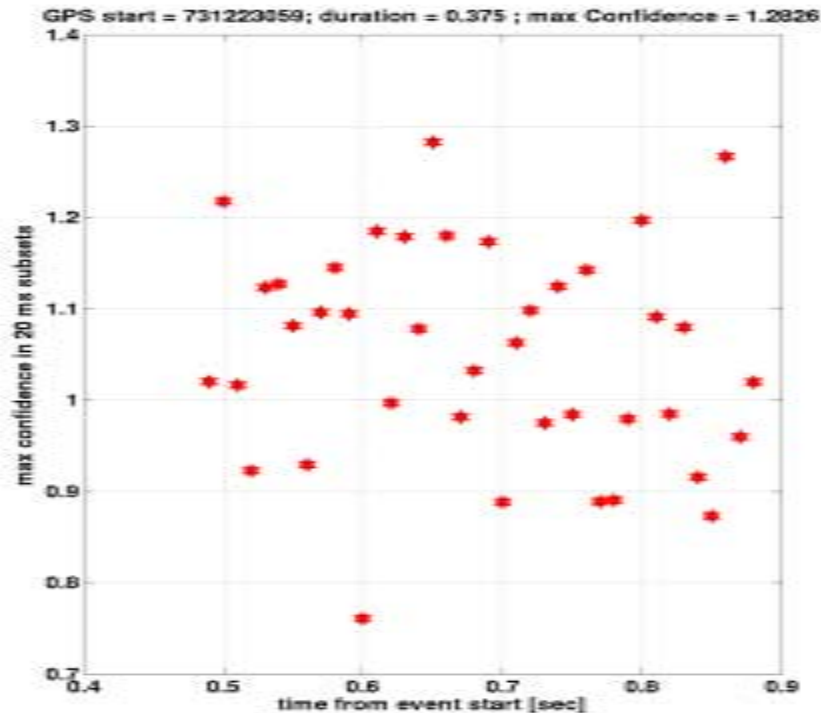
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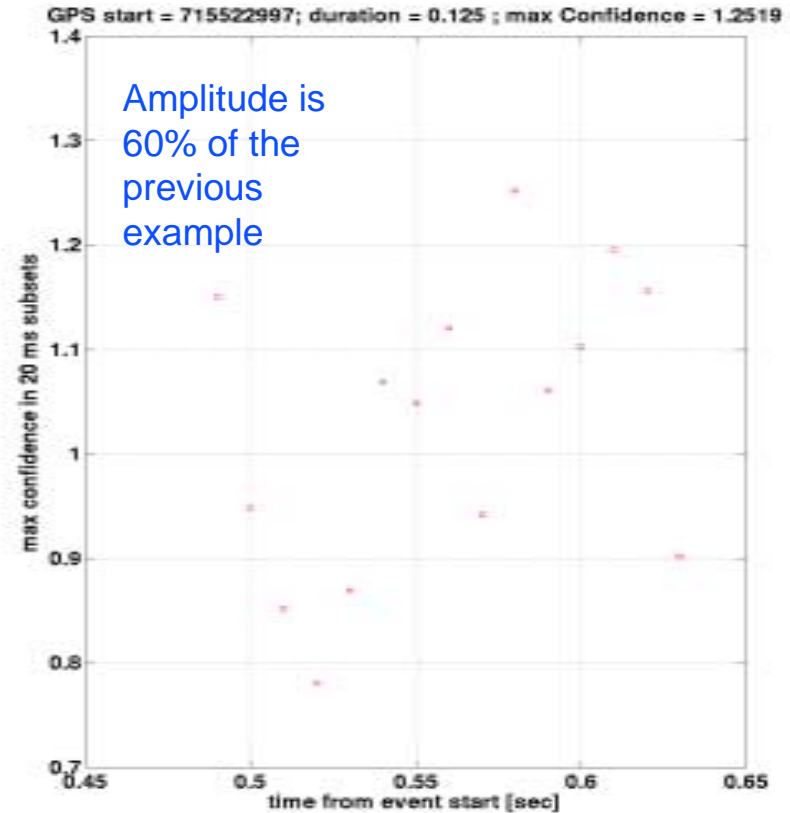
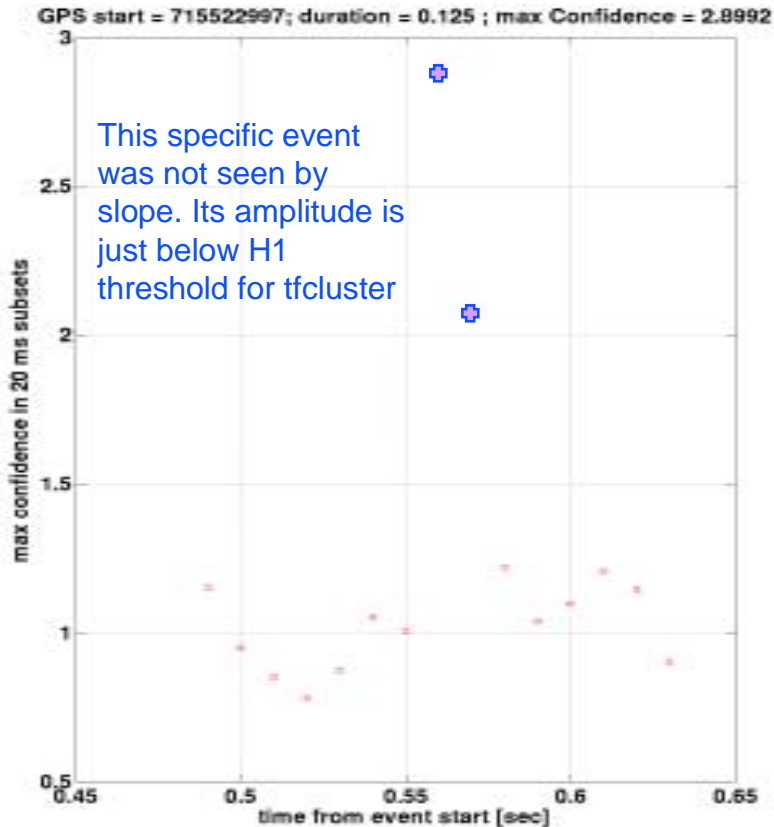
# Background events



2 instances of L1H1 coincidence in the S2 playground: max confidence = 1.3  
(significance = 0.05)

Need to sample more of these to tune the cut. 1.6-2 seems reasonable on the basis of what seen so far.

# S1 software injections (1 ms Gaussian)



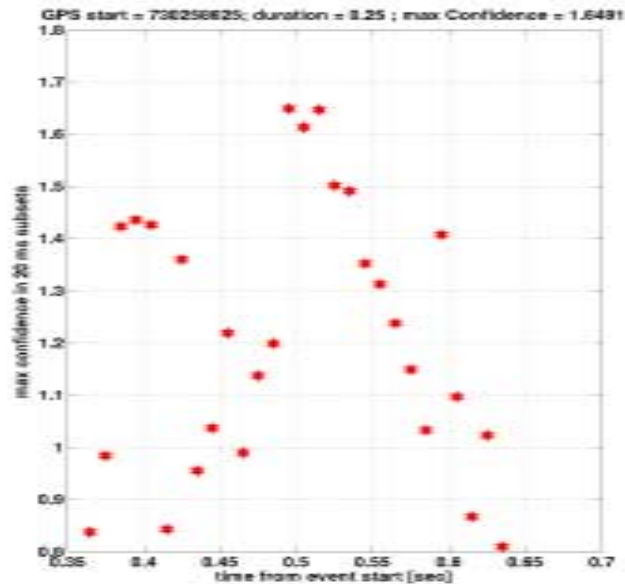
This would pass confidence cut

No correlation here

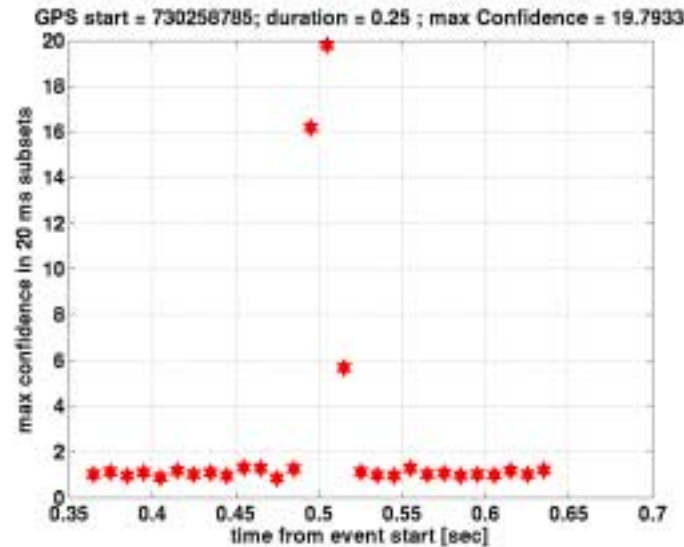
# S2 hardware injections (Feb 25, 2003)

| GPS                       | cycle | frequency | L1tfclu | H1tfclu | L1-H1  | confidence | visible peak? |
|---------------------------|-------|-----------|---------|---------|--------|------------|---------------|
| <a href="#">730258565</a> | 1     | 100       | 0       | 0       | -2.3ms | 0.8        | NO            |
| <a href="#">730258585</a> | 1     | 153       | 1       | 0       | -2.3ms | 11.6       | YES           |
| <a href="#">730258605</a> | 1     | 235       | 1       | 0       | 8.7ms  | 2.2        | YES           |
| <a href="#">730258625</a> | 1     | 361       | -1      | 0       | 8.4ms  | 1.6        | NO            |
| <a href="#">730258645</a> | 1     | 554       | 0       | 0       | -2.8ms | 1.0        | NO            |
| <a href="#">730258665</a> | 1     | 850       | 0       | 0       | -6.5ms | 1.2        | NO            |
| <a href="#">730258685</a> | 1     | 1304      | -1      | 0       | -6.1ms | 2.0        | YES           |
| <a href="#">730258705</a> | 1     | 2000      | 0       | 0       | -7.0ms | 1.8        | NO            |
| <a href="#">730258725</a> | 2     | 100       | -1      | 0       | 9.6ms  | 9.7        | YES           |
| <a href="#">730258745</a> | 2     | 153       | 1       | 1       | 1.6ms  | 36.8       | YES           |
| <a href="#">730258765</a> | 2     | 235       | 1       | 1       | 0.1ms  | 23.7       | YES           |
| <a href="#">730258785</a> | 2     | 361       | 1       | 1       | 0.2ms  | 19.3       | YES           |
| <a href="#">730258805</a> | 2     | 554       | 1       | -1      | -0.1ms | 12.9       | YES           |
| <a href="#">730258825</a> | 2     | 850       | -1      | 0       | 2.1ms  | 3.0        | YES           |
| <a href="#">730258845</a> | 2     | 1304      | 1       | 0       | 0.4ms  | 14.6       | YES           |
| <a href="#">730258865</a> | 2     | 2000      | 1       | 0       | 0.1ms  | 6.6        | YES           |
| <a href="#">730258885</a> | 3     | 100       | 0       | 0       | 3.5ms  | 43.0       | YES           |
| <a href="#">730258905</a> | 3     | 153       | 1       | 1       | -1.5ms | 67.6       | YES           |
| <a href="#">730258925</a> | 3     | 235       | 1       | 1       | -3.7ms | 55.8       | YES           |
| <a href="#">730258945</a> | 3     | 361       | 1       | 1       | 0.6ms  | 52.3       | YES           |
| <a href="#">730258965</a> | 3     | 554       | 1       | 1       | -0.9ms | 42.3       | YES           |
| <a href="#">730258985</a> | 3     | 850       | -1      | 1       | 0.4ms  | 25.8       | YES           |
| <a href="#">730259005</a> | 3     | 1304      | 1       | 1       | 0.4ms  | 31.0       | YES           |
| <a href="#">730259025</a> | 3     | 2000      | 1       | 1       | 0.1ms  | 25.9       | YES           |

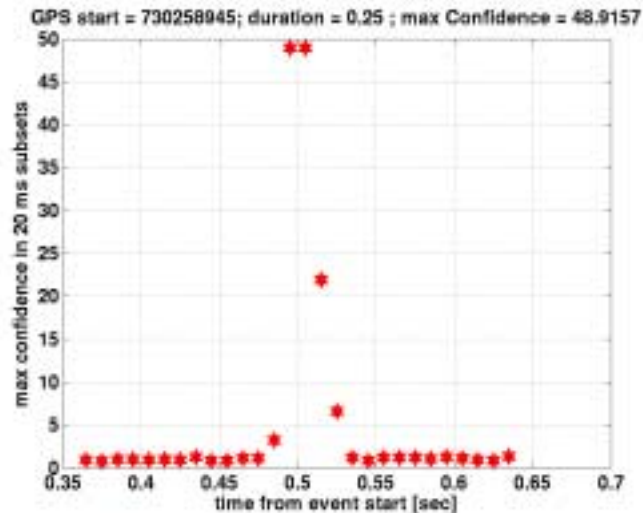
# S2 hardware injections (02/25/03) (554 Hz Sine Gaussian)



554Hz , weak  
Not detected



554Hz  
Power x10  
detected



554Hz  
Power x100  
detected

# Summary

## Suggested method: cross correlation in time domain.

- » Assigns a confidence to coincidence events at the end of the burst pipeline.
- » Verifies the waveforms are consistent.
- » Computationally expensive (present MATLAB implementation: 10 minutes for a 0.5 sec event – could do better), but manageable, since it does not act on the raw data flow but to a finite number of (short) time intervals.
- » Reduces false rate in the burst analysis
  
- » TO-DO list:
  - Run over a suite of S2 playground events
  - Tune a confidence cut (S1 as playground for S2?)
  - Verify effect on burst detection sensitivity, with burst Montecarlo
  - Try more interesting waveforms