



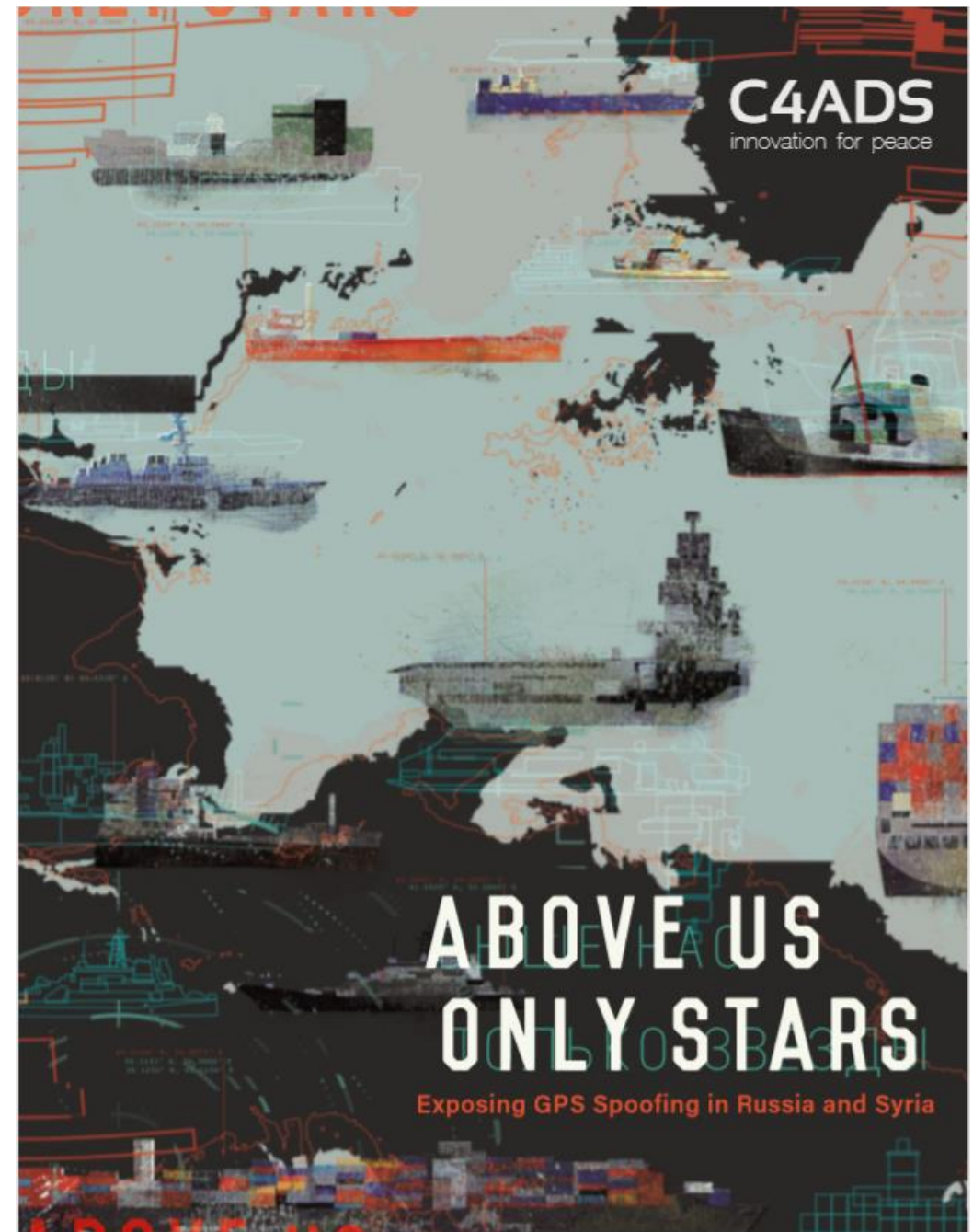
GNSS spoofing detection using two receivers in a very short baseline

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Introduction

- > GNSS Spoofing: realistic threat
- > Not only state actors, also terrorist groups or criminals may spoof GNSS => Danger for both military and civil society
- > Spoofing detection to increase integrity
- > BSc. Thesis on spoofing detection using commercial off the shelf GNSS receivers





Research questions/objectives

- > Is it possible to detect spoofing without knowing or estimating the baseline orientation?
- > If possible, what are the probabilities of missed detection and false alarm?

Constraints:

- Instantaneous
- Commercial off the shelf, (low cost) receivers
- Easy implementation on (military) platform systems

Assumption:

- All satellites used for the position solution are either genuine or counterfeit



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- 2 Theory
- 3 Experimental setup
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- 5 Conclusions



Theoretical observation model

Carrier phase observation model:

$$\phi_r^s \lambda = \rho_r^s + c(dt_r - dt^s) + \lambda(d\phi_r - d\phi^s + N_r^s) + A_r^s + \epsilon_r^s$$

Single differenced carrier phase observation model:

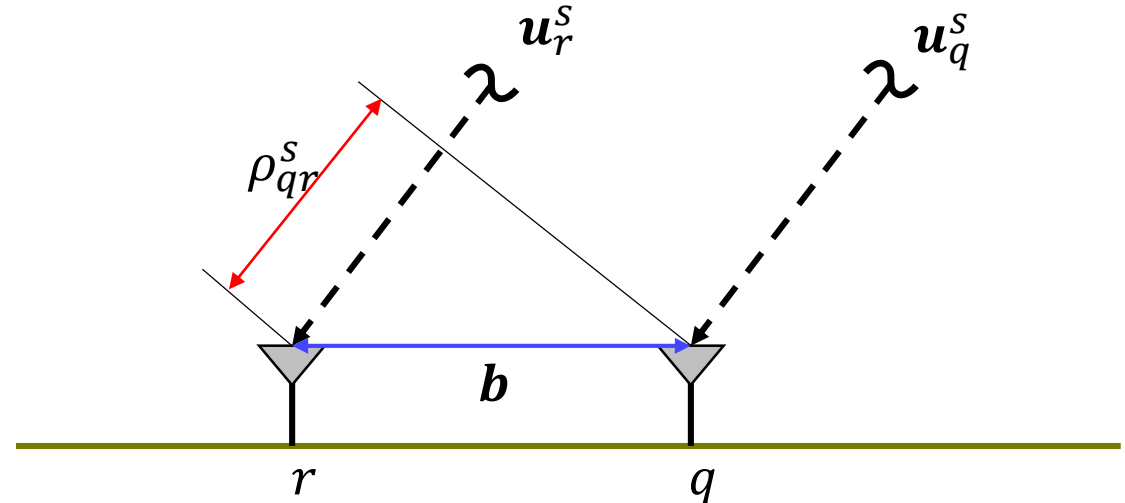
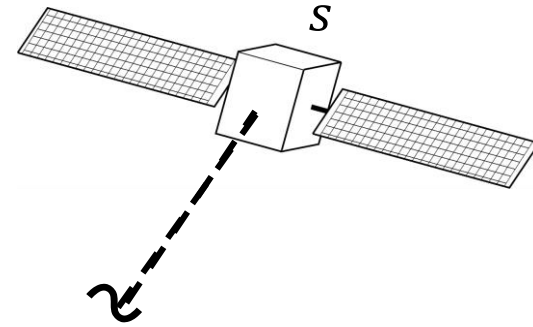
$$\lambda\phi_{qr}^s = \rho_{qr}^s + c dt_{qr} + \lambda(d\phi_{qr} + N_{qr}^s) + \epsilon_{qr}^s$$

With the line of sight vector:

$$\mathbf{u}_r^s \approx \mathbf{u}_q^s \approx \mathbf{u}_0^s$$

We can write:

$$\rho_{qr}^s = (\mathbf{u}_0^s)^T \cdot \mathbf{b}$$





Theoretical observation model

SD genuine: $\lambda\phi_{qr}^s = (\mathbf{u}_0^s)^T \cdot \mathbf{b} + c dt_{qr} + \lambda(d\phi_{qr} + N_{qr}^s) + \epsilon_{qr}^s$

SD spoofed: $\lambda\phi_{qr}^s = (\mathbf{u}_0^f)^T \cdot \mathbf{b} + c dt_{qr} + \lambda(d\phi_{qr} + N_{qr}^{s,f}) + \epsilon_{qr}^{s,f}$

Constructing Double Differenced (DD) observations results in:

DD genuine: $\lambda\phi_{qr}^{st} = (\mathbf{u}_0^{st})^T \cdot \mathbf{b} + \lambda(N_{qr}^{st}) + \epsilon_{qr}^{st}$

DD spoofed: $\lambda\phi_{qr}^{st} = \lambda(N_{qr}^{st,f}) + \epsilon_{qr}^{st,f}$

Introducing fractional phase observations: $\tilde{\phi}_{qr}^{st} = \phi_{qr}^{st} - \text{round}[\phi_{qr}^{st}]$

DD genuine: $\tilde{\phi}_{qr}^{st} = B_{qr}^{st} + e_{qr}^{st}$

DD spoofed: $\tilde{\phi}_{qr}^{st} = 0 + e_{qr}^{st,f}$



Hypothesis testing model

Assumption: $\tilde{\phi}_{qr}^{st} \sim N(B_{qr}^{st}, \sigma_{DD}^2)$

Lets define:

$$\Phi = [\tilde{\phi}_{12}^{12}, \tilde{\phi}_{12}^{13}, \dots \dots \tilde{\phi}_{12}^{1m}]^T$$

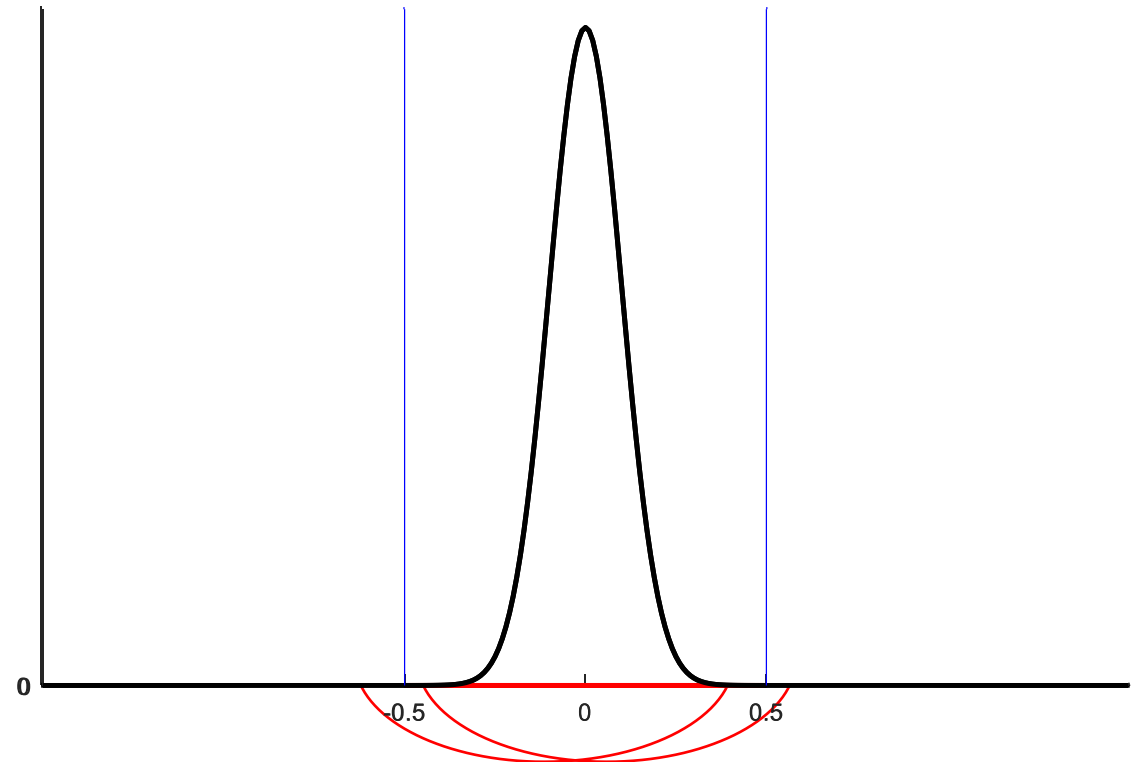
Valid if $|\mathbf{b}| > \lambda$, verified using simulations

Stochastic hypothesis

$$H_0: \Phi \sim U\left(-\frac{1}{2}, \frac{1}{2}\right)$$

$$H_A: \Phi \sim N(0, \sigma_{DD}^2)$$

Assumption $\sigma_{DD} \ll \lambda$





Test statistic

Applying the Simple Likelihood Ratio test leads to the following test statistic:

$$T_s = \Phi^T Q_\phi^{-1} \Phi$$

Reject H_0 if $T_s < k_\alpha$

Test statistic for normally distributed observations (H_A) => m-dimensional central Chi-Squared distribution

However, the H_0 results in the sum of squared uniformly distributed observations => no closed form of the distribution, very complex situation

Monte-Carlo simulation!!



Probability of false alarm and missed detection

Type I error:

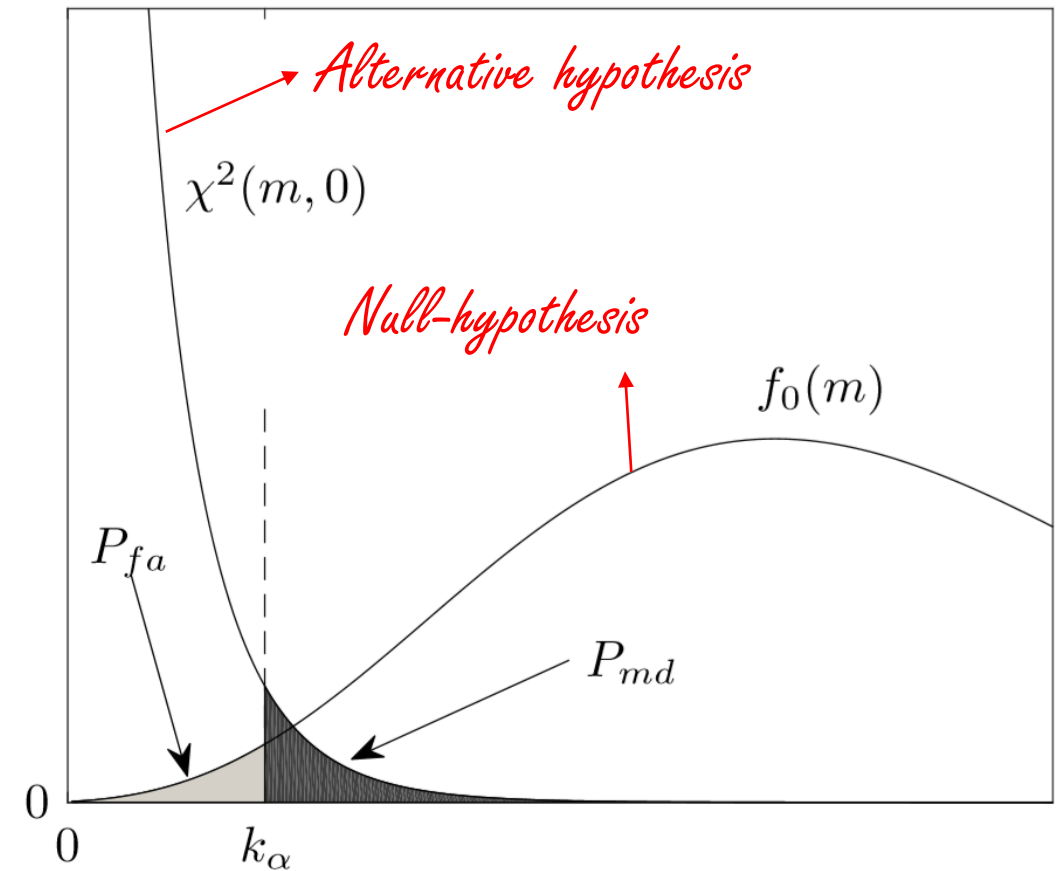
- > Rejection of H_0 when in fact H_0 is true.
- > false alarm

Continuity & availability

Type II error:

- > Acceptance of H_0 when in fact H_0 is false.
- > missed detection

Integrity





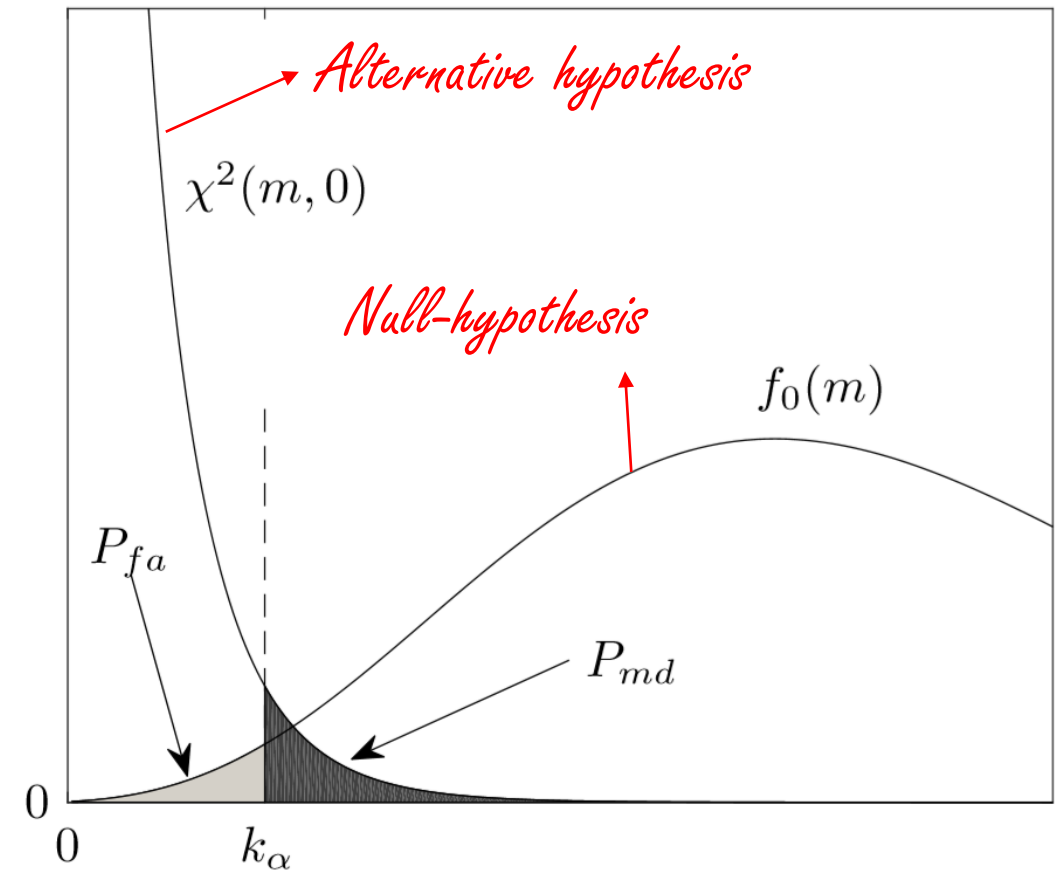
Probability of false alarm and missed detection

k_α directly influences the P_{md} and P_{fa} .

So, how to determine k_α ?

Two options:

- > Determine k_α based on a fixed P_{md} then determine P_{fa}
- > Determine k_α based on a fixed P_{fa} then determine P_{md}





Monte-Carlo simulation

- › Estimation of the false alarm rate
 - Calculate test statistic for each combination of P_{md} , σ_f and m
 - 10^9 random uniformly distributed samples per combination
 - Calculate the test statistic T_s
 - Determine for how many samples $T_s < k_\alpha$ (false alarm)

$P_{md} = 0.001$						
m	$\sigma_f = 0.01$	$\sigma_f = 0.02$	$\sigma_f = 0.03$	$\sigma_f = 0.04$	$\sigma_f = 0.05$	$\sigma_f = 0.10$
3	2.8e-04	2.2e-03	7.4e-03	1.8e-02	3.4e-02	2.7e-01
4	1.7e-05	2.7e-04	1.4e-03	4.3e-03	1.1e-02	1.7e-01
5	9.6e-07	3.2e-05	2.4e-04	1.0e-03	3.1e-03	1.0e-01
6	5.1e-08	3.8e-06	4.2e-05	2.4e-04	9.1e-04	5.9e-02
7	4.0e-09	4.2e-07	7.3e-06	5.5e-05	2.6e-04	3.4e-02
8	<1.0e-9	5.6e-08	1.2e-06	1.2e-05	7.4e-05	1.9e-02
9	<1.0e-9	6.0e-09	2.0e-07	2.8e-06	2.1e-05	1.1e-02
10	<1.0e-9	<1.0e-9	4.5e-08	6.2e-07	5.7e-06	5.8e-03
11	<1.0e-9	1.0e-09	4.0e-09	1.1e-07	1.5e-06	3.1e-03
12	<1.0e-9	<1.0e-9	1.0e-09	3.6e-08	4.0e-07	1.7e-03
13	<1.0e-9	<1.0e-9	<1.0e-9	7.0e-09	1.0e-07	9.1e-04
14	<1.0e-9	<1.0e-9	<1.0e-9	<1.0e-9	3.4e-08	4.8e-04
15	<1.0e-9	<1.0e-9	<1.0e-9	<1.0e-9	8.0e-09	2.5e-04



Experiment

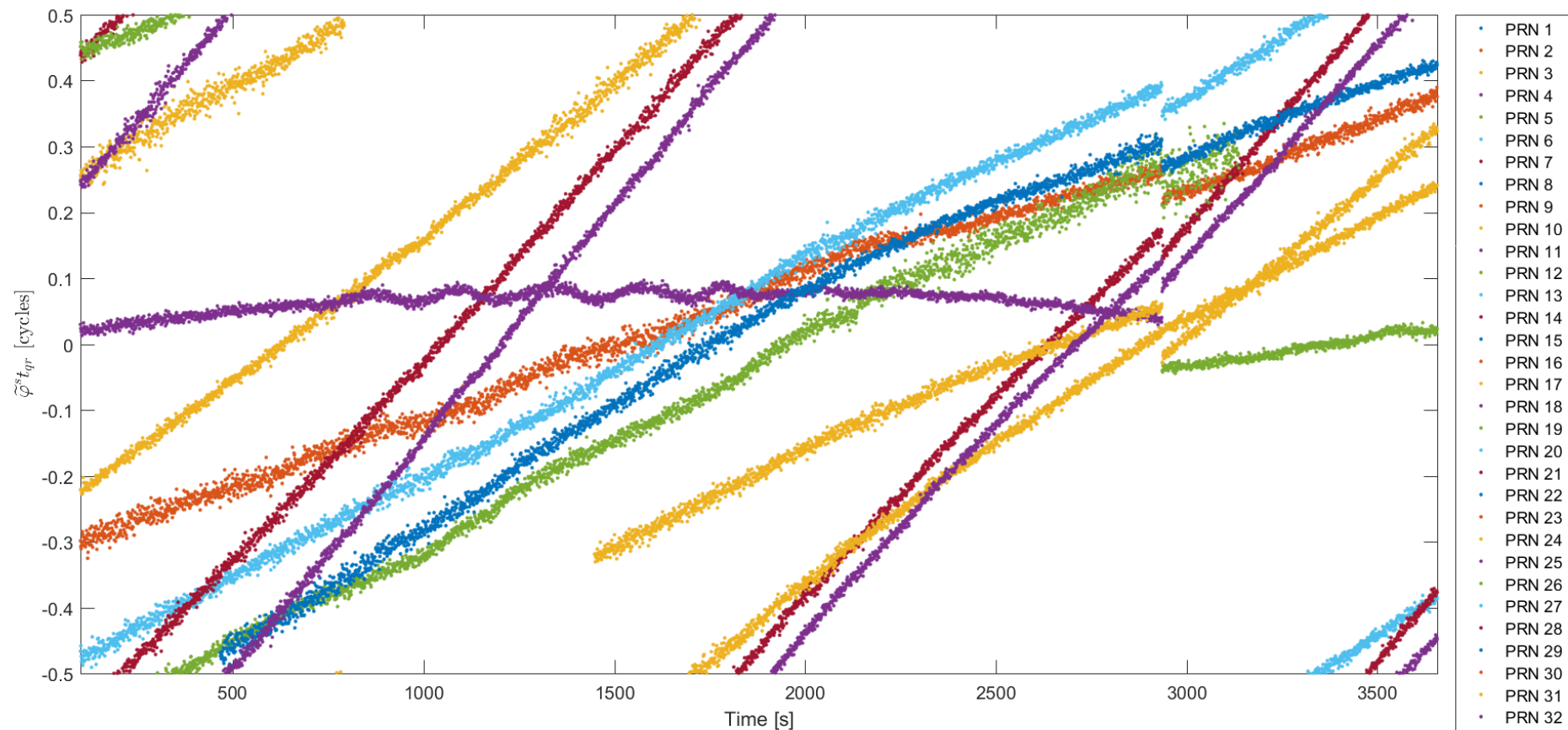


TEST LOCATIONS



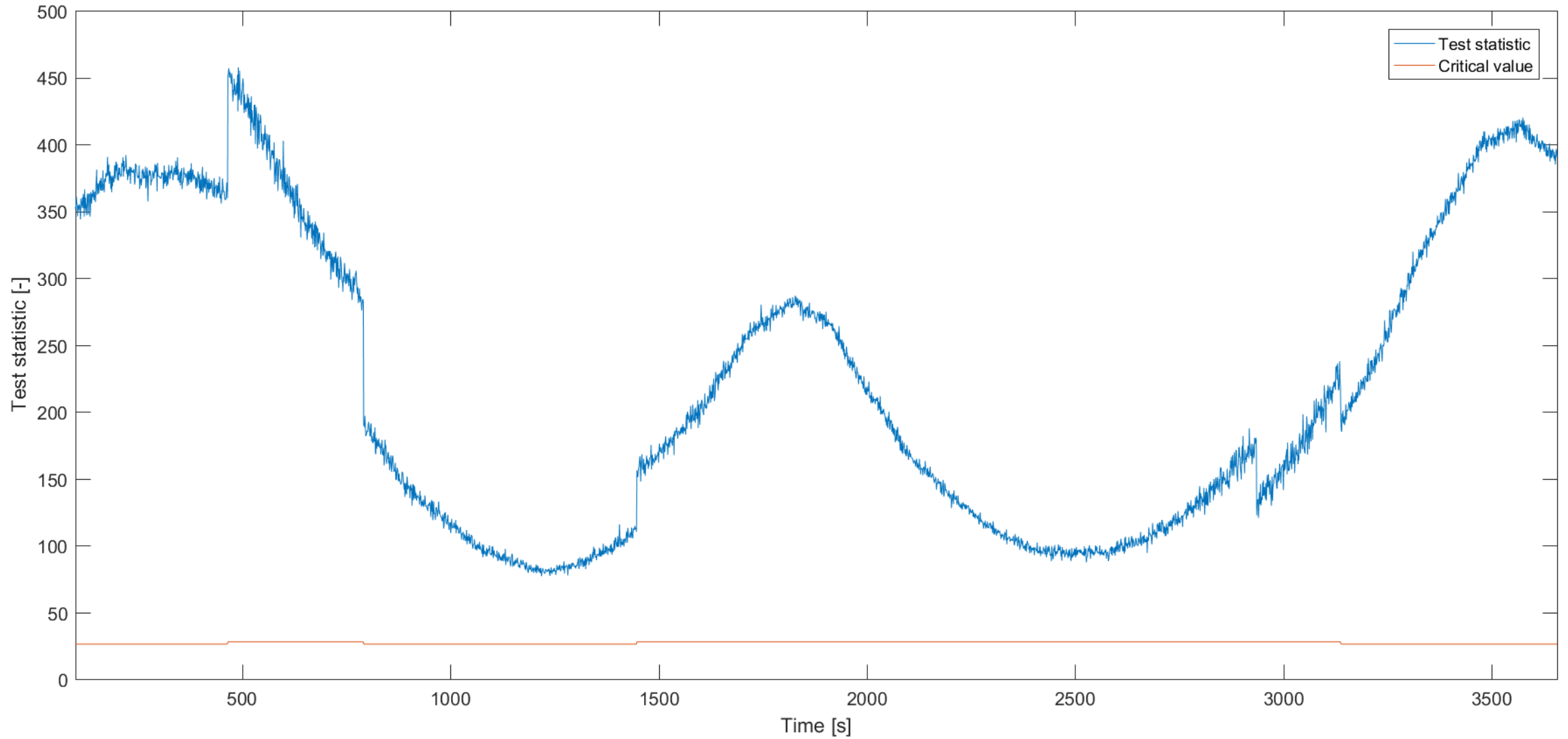


Experimental results – Sea dyke





Results: Dyke, Test statistic





Observations based on more genuine observations

> In total 216 minutes of genuine observations @ 1Hz (~ 13000 epochs)

> For testing it is assume that $\sigma_f = 0.10$ cycles

> On average 9 satellites visible (elevation cutoff @ 5 deg)

> $P_{md} = 0.001$

– 94 alerts ($P_{fa} = 7.2 \times 10^{-3}$)

> $P_{md} = 0.01$

– 60 alerts ($P_{fa} = 5.2 \times 10^{-3}$)

> $P_{md} = 0.1$

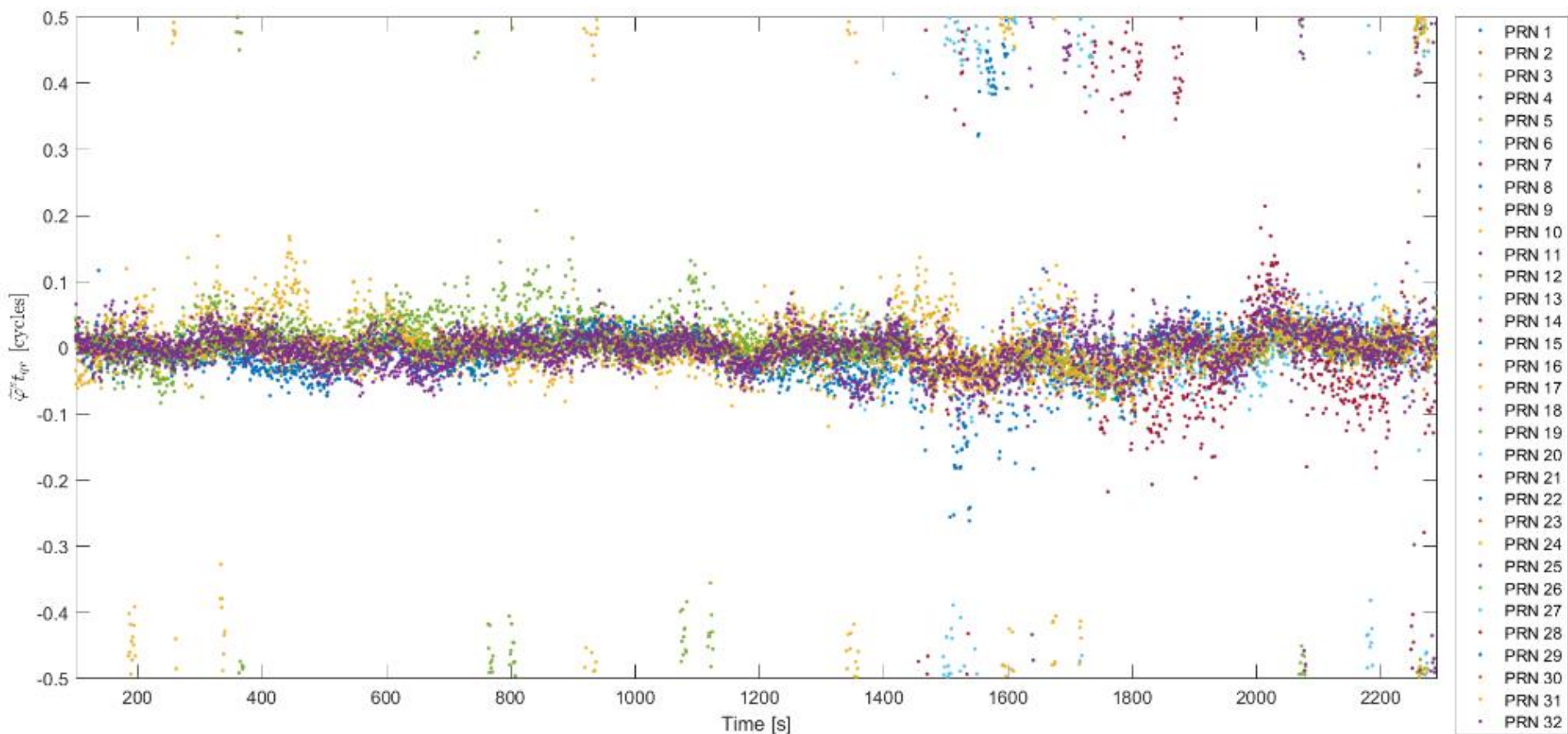
– 9 alerts ($P_{fa} = 6.9 \times 10^{-4}$)

Worst case scenario

$P_{md} = 0.1$						
m	$\sigma_f = 0.01$	$\sigma_f = 0.02$	$\sigma_f = 0.03$	$\sigma_f = 0.04$	$\sigma_f = 0.05$	$\sigma_f = 0.10$
3	6.5e-05	5.2e-04	1.8e-03	4.2e-03	8.2e-03	6.5e-02
4	3.0e-06	4.8e-05	2.4e-04	7.6e-04	1.9e-03	3.0e-02
5	1.5e-07	4.5e-06	3.3e-05	1.4e-04	4.3e-04	1.4e-02
6	1.1e-08	3.9e-07	4.6e-06	2.6e-05	9.7e-05	6.2e-03
7	<1.0e-9	3.5e-08	6.4e-07	4.6e-06	2.2e-05	2.8e-03
8	<1.0e-9	3.0e-09	1.0e-07	8.6e-07	5.0e-06	1.3e-03
9	<1.0e-9	1.0e-09	1.2e-08	1.5e-07	1.2e-06	5.9e-04
10	<1.0e-9	<1.0e-9	2.0e-09	3.4e-08	2.6e-07	2.7e-04
11	<1.0e-9	<1.0e-9	<1.0e-9	1.1e-08	5.2e-08	1.2e-04
12	<1.0e-9	<1.0e-9	<1.0e-9	5.0e-09	1.6e-08	5.5e-05
13	<1.0e-9	<1.0e-9	<1.0e-9	<1.0e-9	5.0e-09	2.4e-05
14	<1.0e-9	<1.0e-9	<1.0e-9	<1.0e-9	<1.0e-9	1.1e-05
15	<1.0e-9	<1.0e-9	<1.0e-9	<1.0e-9	<1.0e-9	4.9e-06



Experimental results – Patrol vessel





Outlier detection and removal

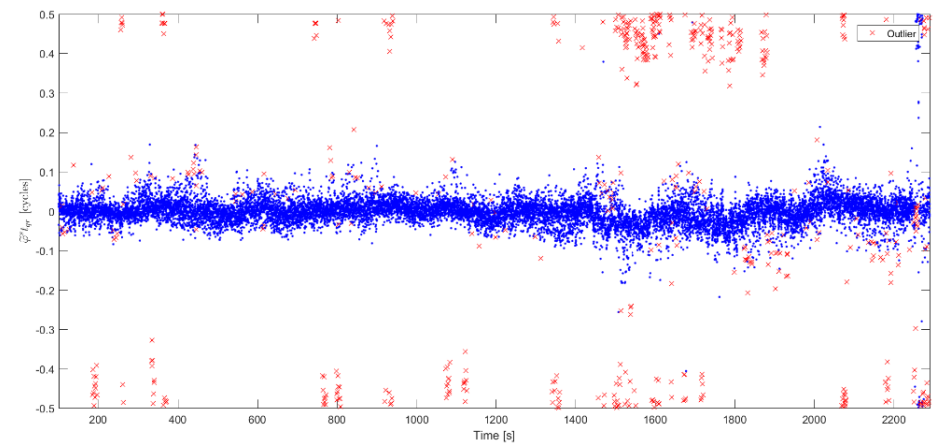
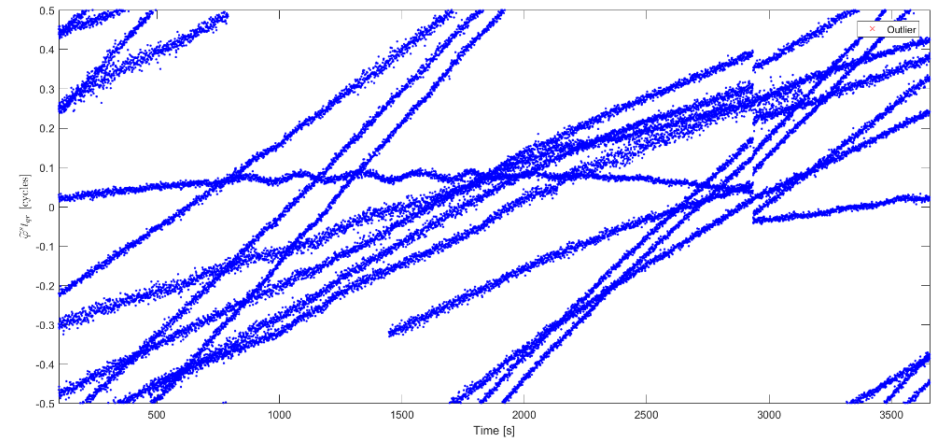
- > Z-score
 - Standard deviation required
- > More robust: Modified Z-score

$$M_{qr}^{ts} = \frac{\tilde{\phi}_{qr}^{ts} - \text{median}(\tilde{\Phi})}{MAD}$$

With

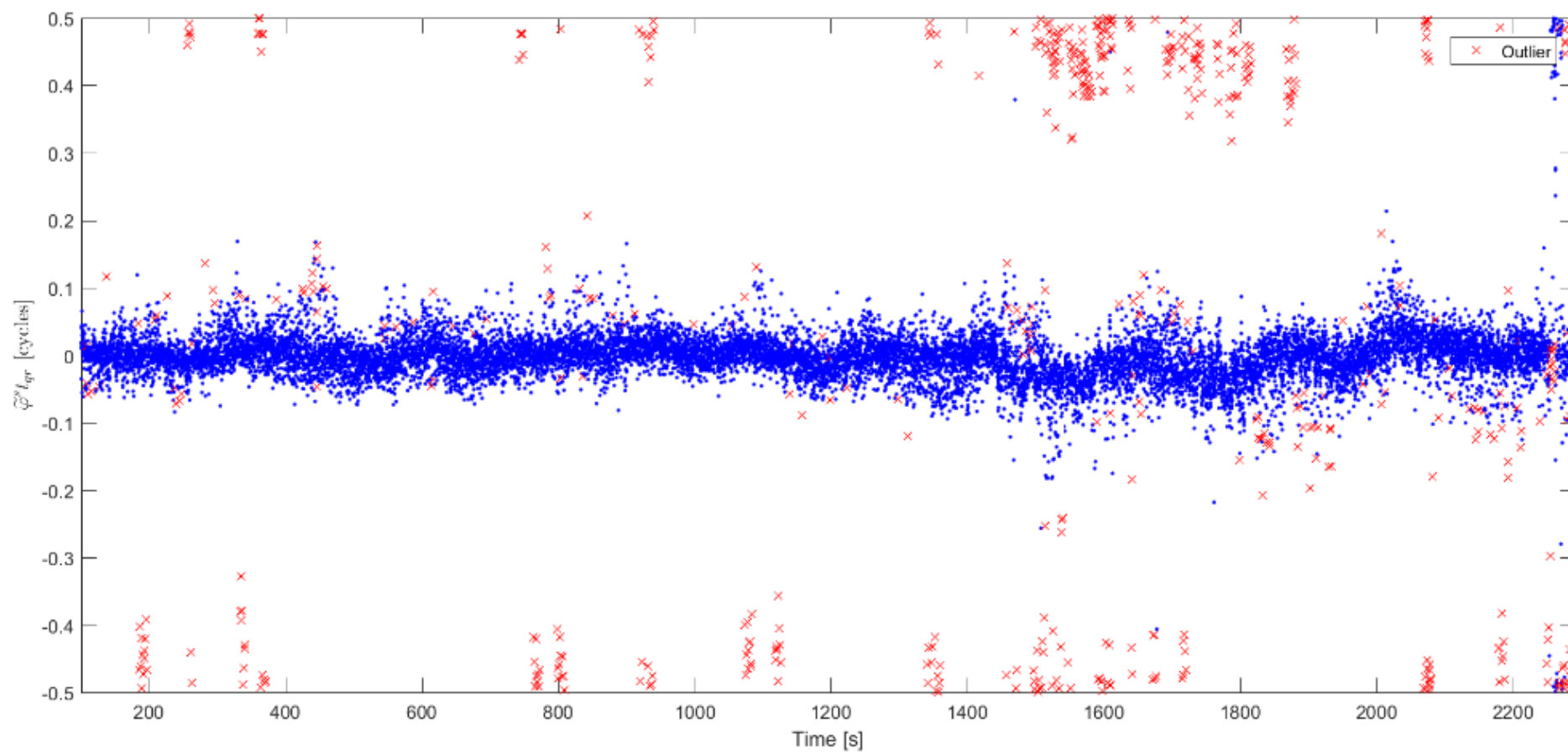
$$MAD = 1.483 \text{ median}|\phi_{qr}^{ts} - \text{median}(\tilde{\Phi})|$$

Remove observation if $M_{qr}^{ts} > \text{threshold}$



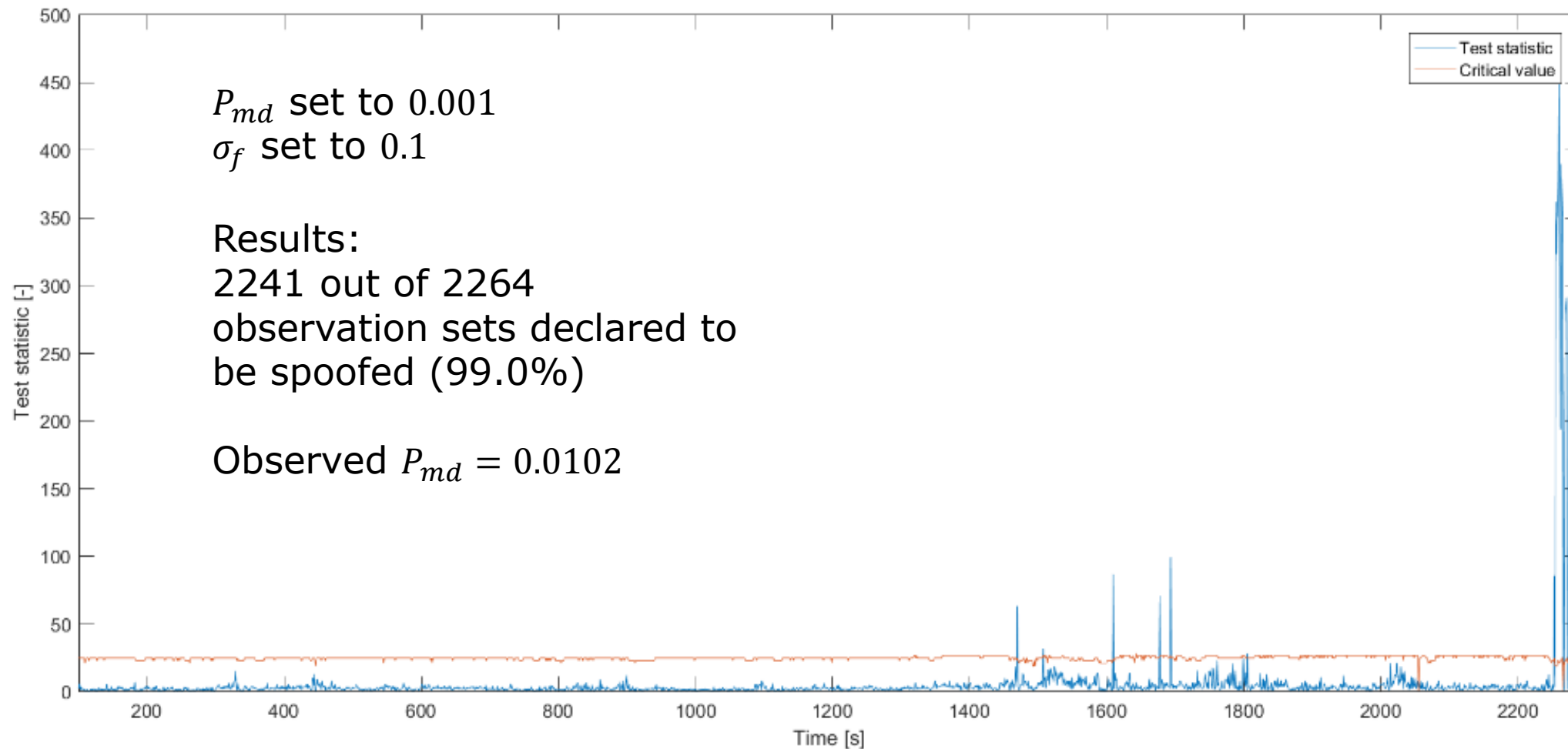


Experimental results – Patrol vessel



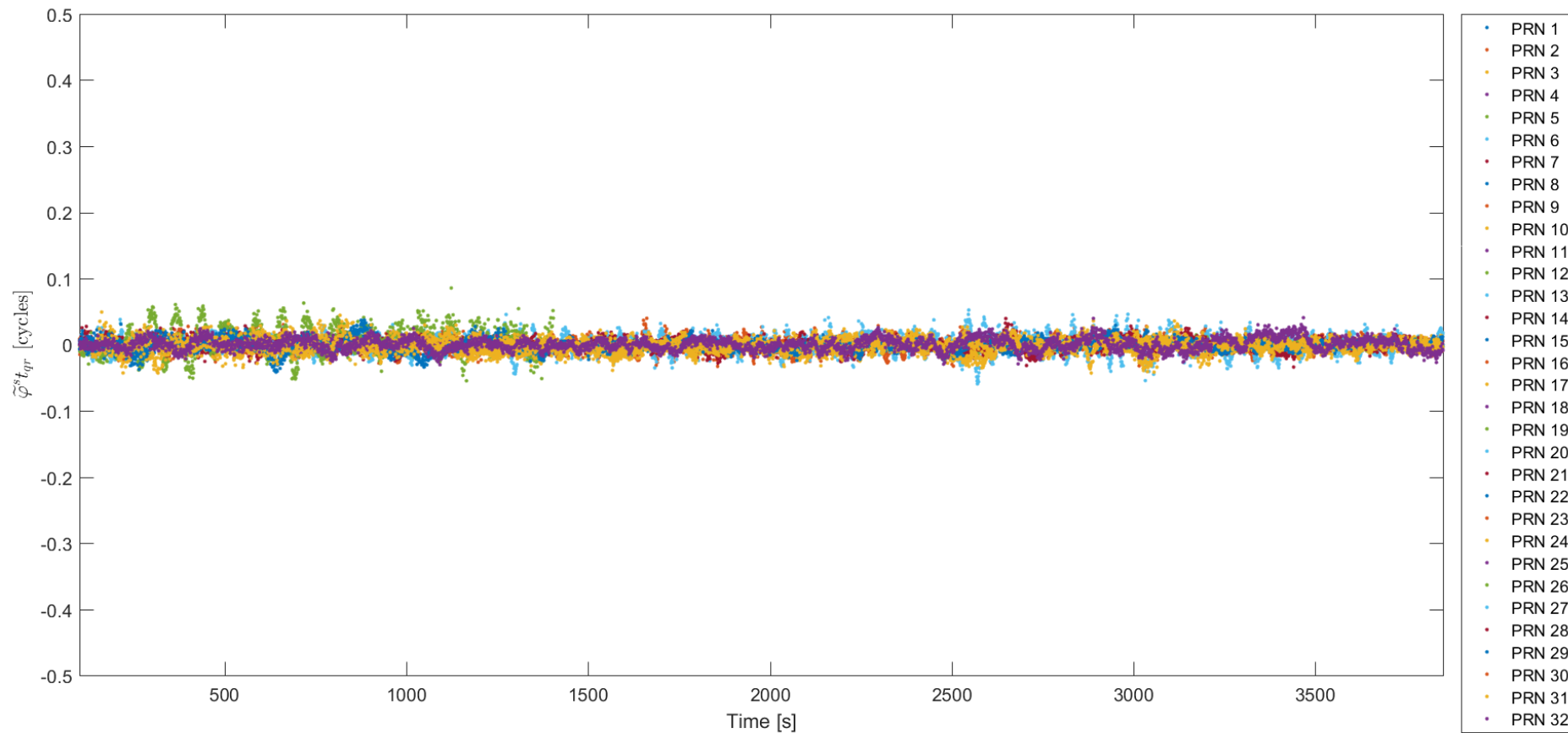


Experimental results – Patrol vessel



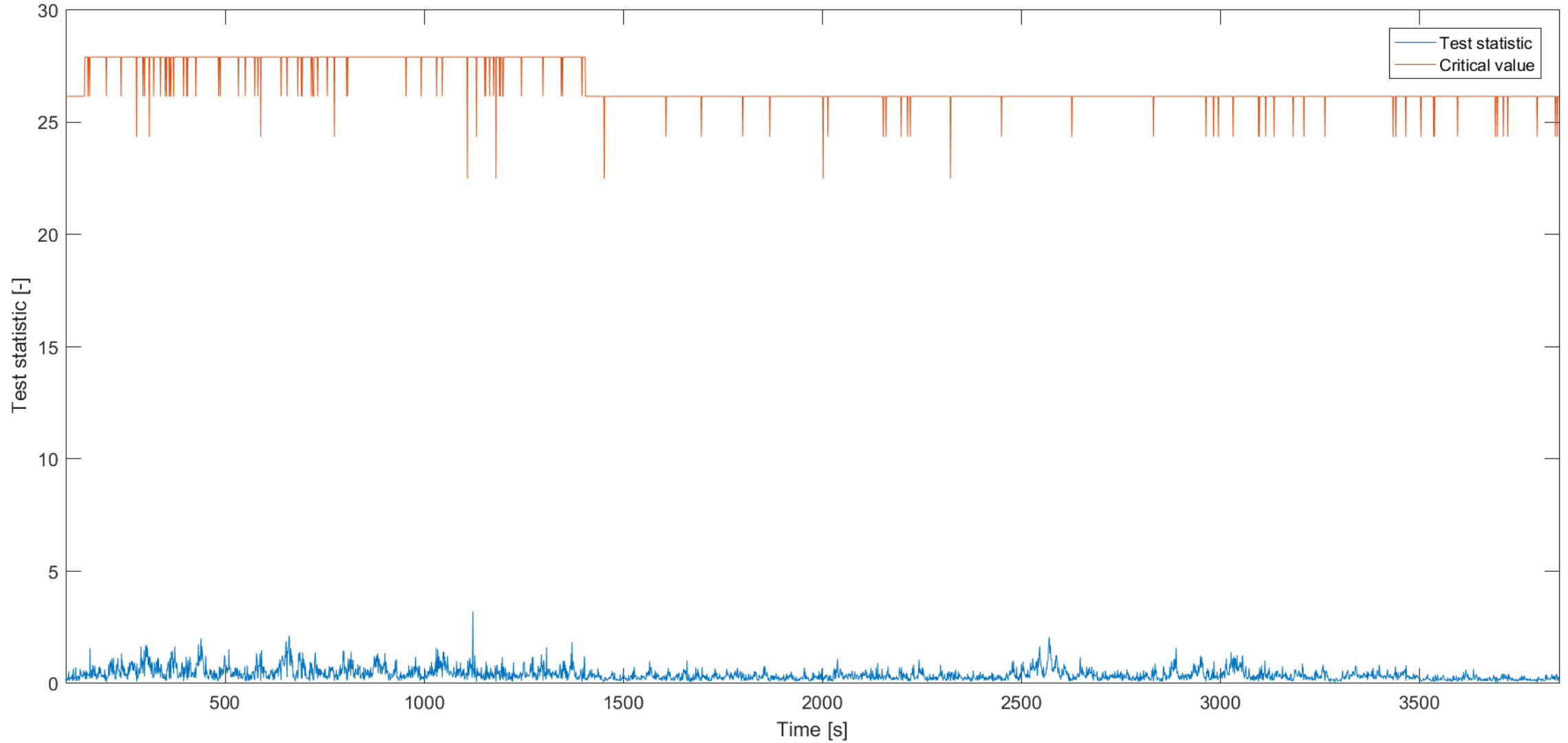


Experimental results – Schiphol





Results: Hangar, Test statistic





Conclusions

- > Measurements confirm theory
- > Method seems to be a very simple but effective spoofing detection method
- > Reasonable probabilities of missed detection and false alarm possible

Future work

- > Validate method with more data
- > Expand method from static to dynamic
- > Determine realistic noise levels for spoofed observations



Thank you for your attention!