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Extract Before Detect

N-signal Complex Approximate Message Passing applied to radar signals

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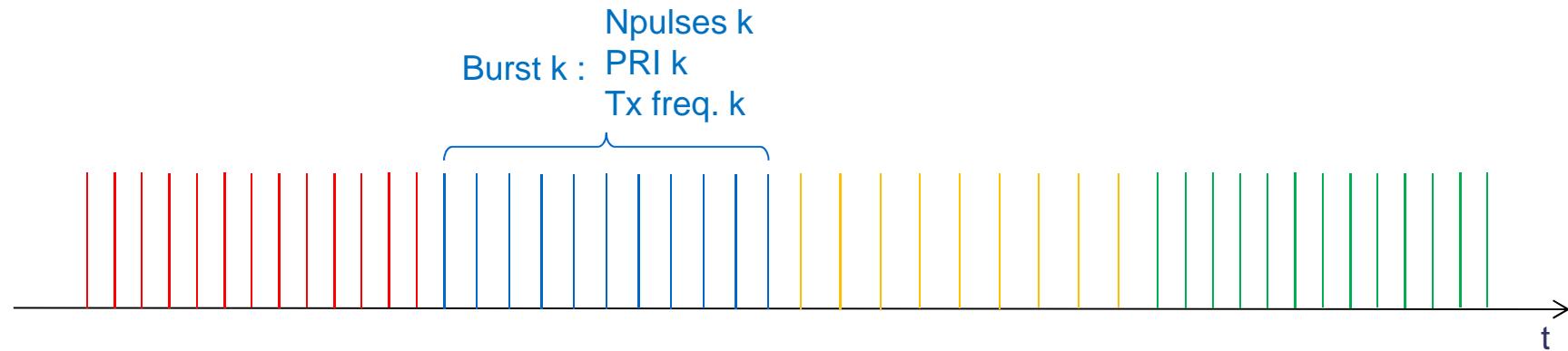
Summary

- ◆ Compressive Sensing radar processing
- ◆ CAMP (Complex Approximate Message Passing) applied to multiple burst signal
- ◆ Simulation results
- ◆ Conclusion

Summary

- ◆ Compressive Sensing radar processing
 - Radar signals
 - Compressed Sensing
 - Radar processing objective in terms of Compressive Sensing
 - Extract Before Detect
- ◆ CAMP (Complex Approximate Message Passing) applied to multiple burst signal
- ◆ Simulation results
- ◆ Conclusion

Radar signals

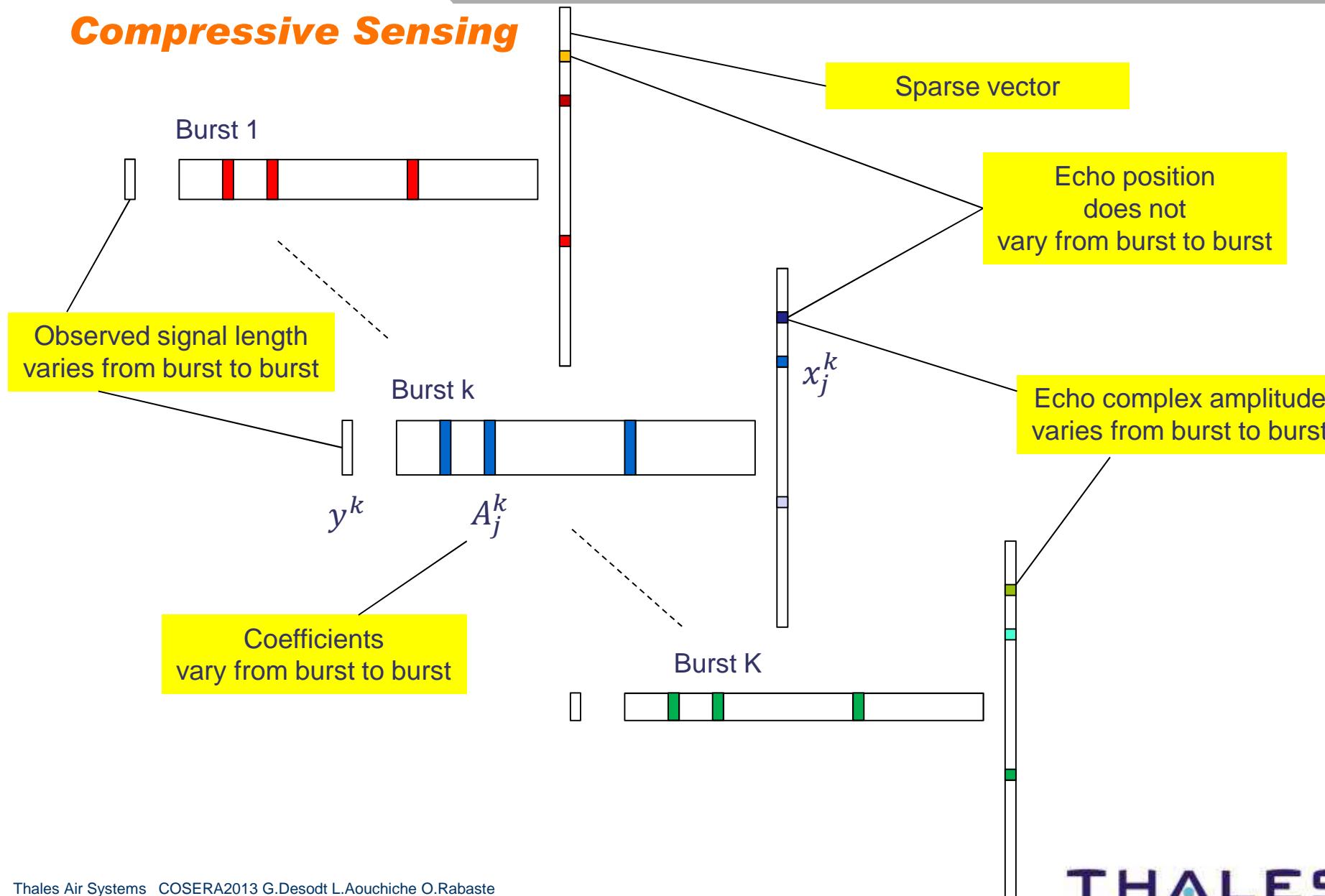


$$y^k = n^k + \sum_i A_i^k \cdot x_i^k$$

Annotations for the equation:

- Radar signal received during 1 burst**
- noise**
- Burst index, $k=1\dots K$**
- Echo index, $i=1\dots I$**
- Echo complex amplitude**
- Echo "signature"**
- Steering vector**
- Receivers (function of echo direction)**
- Coherent pulses (function of echo radial speed)**
- Range bins (function of the range radar-echo)**

Compressive Sensing

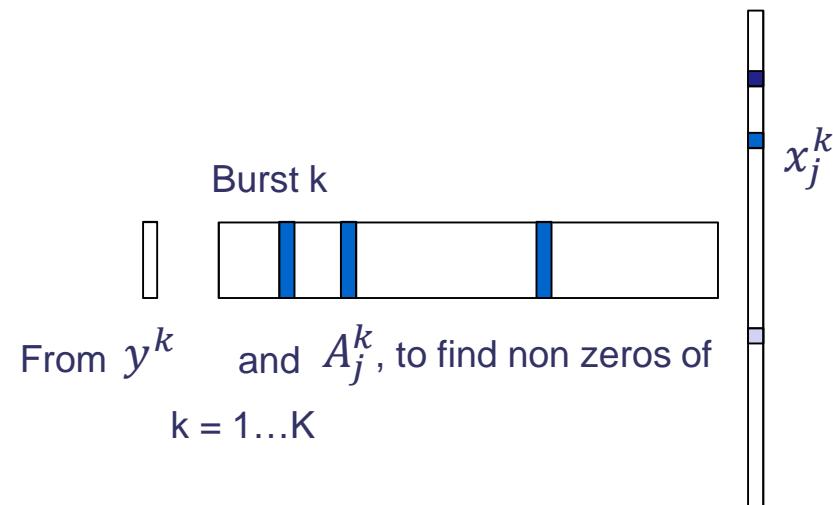


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Radar processing objective ...

- ◆ To detect \Leftrightarrow To find echo complex amplitudes
- ◆ To locate \Leftrightarrow To find echo positions

... in terms of Compressive Sensing



- ◆ Compressive Sensing processing produces at once
 - echo positions
 - and their amplitudes

Extract Before Detect

Related COSERA 2013 presentation :

R. Pribic, I Kyriakides, "Bayesian Compressive sensing in radar systems", session A3
J. Ender, De-ghosting before detect, "A CS approach to the fusion of PCL sensors", session A8

Summary

- ◆ Compressive Sensing radar processing
- ◆ CAMP (Complex Approximate Message Passing) applied to multiple burst signal
 - N-burst criterion
 - CAMP (Complex Approximate Message Passing)
 - N-signal complex soft threshold variation
 - N-signal CAMP (N-CAMP)
- ◆ Simulation results
- ◆ Conclusion

Single burst criterion (LASSO)

$$\diamond \min_x \frac{1}{2} \underbrace{(y - A \cdot x)^H \cdot (y - A \cdot x)}_{\text{Quadratic terms}} + \lambda \cdot \sum_j \sqrt{|x_j|^2}$$

N-burst criterion

$$\diamond \min_{x^k} \frac{1}{2} \sum_k (y^k - A^k \cdot x^k)^H \cdot (y^k - A^k \cdot x^k) + \lambda \cdot \sum_j \sqrt{\sum_k |x_j^k|^2}$$

Quadratic terms are summed over the bursts

L1-2 norm

Related COSERA 2013 presentations :

S. Wenger, M. Magnor, L₁-∞ norm, "Group sparsity imaging algorithm for transient radio sources", session A2

Z. Bingchen et al, Lq norm, "Azimuth ambiguity suppression for SAR imaging based on group sparse reconstruction", session A4

CAMP : Complex Approximate Message Passing

- ◆ D. Donoho, A. Maleki, A. Montanari, Message Passing algorithms for Compressed Sensing, Proceedings of the National Academy of Sciences 106, 2009
- ◆ A. Maleki, L. Anitori, Z. Yang, R. Baraniuk, Asymptotic analysis of complex LASSO via Complex Approximate Message Passing Algorithm (CAMP), IEEE Trans. Information Theory, 2012
- ◆ CAMP is based on complex soft threshold variation

N-signal complex soft threshold variation

◆ N-signal complex soft threshold

$$\bullet \eta^k(v; \lambda) = \max\left(0; 1 - \lambda \cdot \left(\sum_m v^{m*} \cdot v^m\right)^{-\frac{1}{2}}\right) \cdot v^k$$

Quadratic term is summed
over the bursts

◆ N-signal complex soft threshold derivatives

Scaling factor common to all bursts
If it is 0
then all burst amplitudes are 0

$$\bullet \frac{\partial \eta^k(v; \lambda)}{\partial v^n} = \left(1 - \lambda \cdot \left(\sum_m v^{m*} \cdot v^m\right)^{-\frac{1}{2}}\right) \cdot \delta(k - n) + \frac{\lambda}{2} \cdot \left(\sum_m v^{m*} \cdot v^m\right)^{-\frac{3}{2}} \cdot v^{n*} \cdot v^k$$

$$\bullet \frac{\partial \eta^k(v; \lambda)}{\partial v^{n*}} = \frac{\lambda}{2} \cdot \left(\sum_m v^{m*} \cdot v^m\right)^{-\frac{3}{2}} \cdot v^n \cdot v^k$$

◆ N-signal complex soft threshold differential

$$\bullet \eta^k(v + dv; \lambda) - \eta^k(v; \lambda) = \\ \left(1 - \frac{\lambda}{\left(\sum_m v^{m*} \cdot v^m\right)^{\frac{1}{2}}}\right) \cdot dv^k + \frac{\lambda}{2} \cdot \frac{v^k}{\left(\sum_m v^{m*} \cdot v^m\right)^{\frac{1}{2}}} \cdot \sum_n \frac{v^{n*}}{\left(\sum_m v^{m*} \cdot v^m\right)^{\frac{1}{2}}} \cdot \frac{dv^n}{\left(\sum_m v^{m*} \cdot v^m\right)^{\frac{1}{2}}} + \\ \frac{\lambda}{2} \cdot \frac{v^k}{\left(\sum_m v^{m*} \cdot v^m\right)^{\frac{1}{2}}} \cdot \sum_n \frac{dv^{n*}}{\left(\sum_m v^{m*} \cdot v^m\right)^{\frac{1}{2}}} \cdot \frac{v^n}{\left(\sum_m v^{m*} \cdot v^m\right)^{\frac{1}{2}}}$$

N-signal CAMP (N-CAMP)

- ◆ **CAMP expressions applied to K vectors of observed signals**

- $v_j^{k,t} = \sum_b A_{b,j}^k \cdot z_b^{k,t-1} + x_j^{k,t-1}$
- $x_j^k = \eta^k(v_j^k; \tau_t)$
- $z_a^{k,t} = y_a^k - \sum_j A_{a,j}^k \cdot x_j^{k,t} - \sum_j A_{a,j}^k \cdot (\eta^k(v_j^t + dv_{a,j}^t; \tau_t) - \eta^k(v_j^t; \tau_t))$

- ◆ **where the kth term of $dv_{a,j}^t$ in η^k variation is the only non zero term**

- $dv_{a,j}^t = (0, \dots, 0, -A_{a,j}^{k,*} \cdot z_a^{k,t-1}, 0, \dots, 0)$

Summary

- ◆ Compressive Sensing radar processing
- ◆ CAMP (Complex Approximate Message Passing) applied to multiple burst signal
- ◆ Simulation results
 - Domain of application : Doppler axis
 - Scenario 1, targets separated by about 1 ambiguous speed
 - Scenario 2, targets close each other
 - Scenario 3, target separated by exactly 1 ambiguous speed
 - Overall results analysis
- ◆ Conclusion

Domain of application : Doppler axis

- ◆ 5 bursts
- ◆ Same Doppler resolution : 10 m/s
- ◆ Ambiguous speeds : 170, 190, 210, 230, 250 m/s
- ◆ Radial speed range : [-1000 ; 1000] m/s
- ◆ Grid step : 5 m/s
 - Oversampling: factor 2
 - Targets are on the grid

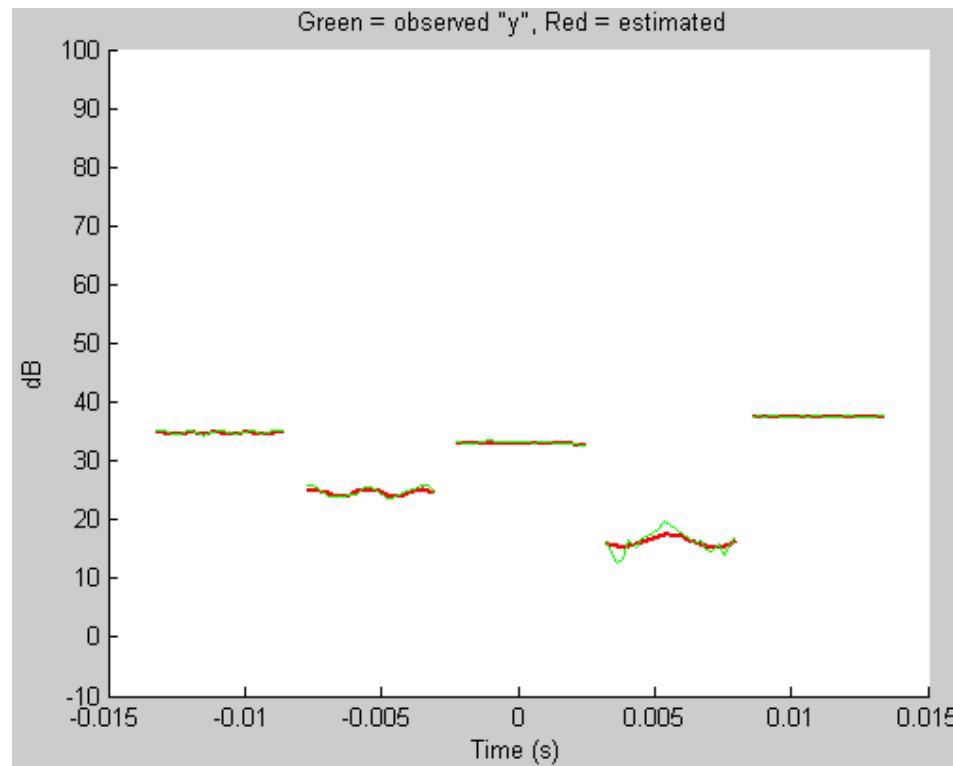
Objective

- ◆ To detect and to locate on the Doppler axis
 - Including Doppler ambiguity solving
 - At least 2 targets simultaneously

Scenario 1: targets separated by 1 ambiguous speed + $\frac{1}{2}\delta v$

- ◆ Target 1: 60 m/s, 50 dB per burst, Swerling 2 fluctuation
- ◆ Target 2: 275 m/s, 20 dB per burst, Swerling 2 fluctuation

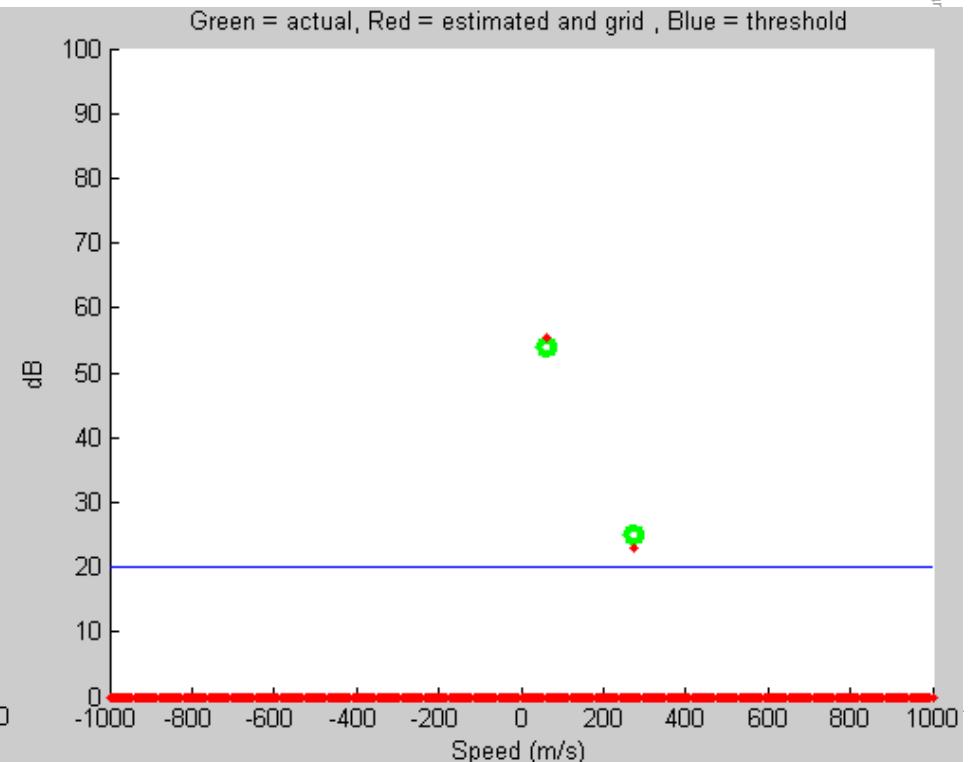
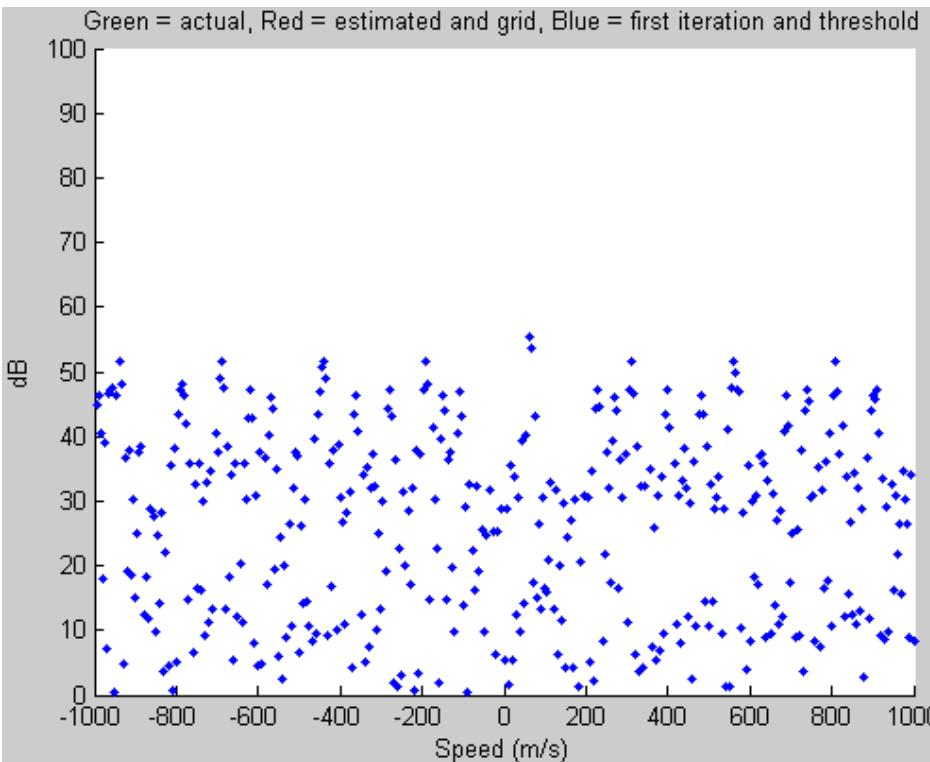
• $v_{target_2} = v_{target_1} + V_{ambiguous_3} + \frac{1}{2}\delta v$



Scenario 1: targets separated by 1 ambiguous speed + $\frac{1}{2}\delta\nu$

- ◆ Target 1: 60 m/s, 50 dB per burst, Swerling 2 fluctuation
- ◆ Target 2: 275 m/s, 20 dB per burst, Swerling 2 fluctuation

$$\textcircled{v}_{\text{target_2}} = v_{\text{target}_1} + V_{\text{ambiguous_3}} + \frac{1}{2}\delta\nu$$

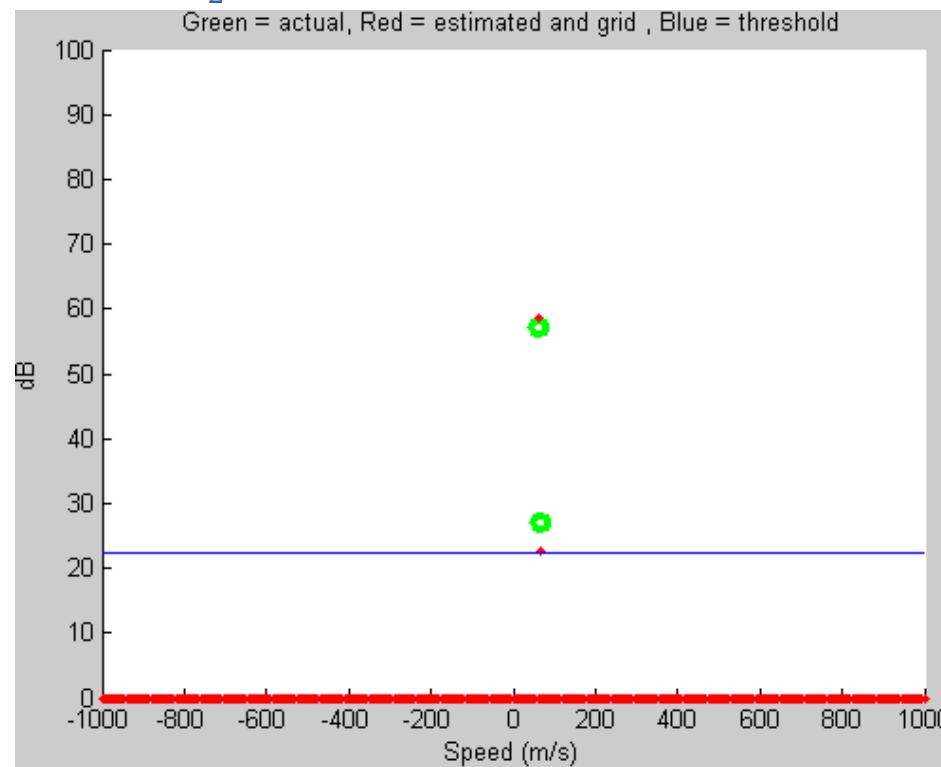


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Scenario 2 : targets separated by $\frac{1}{2} \delta v$

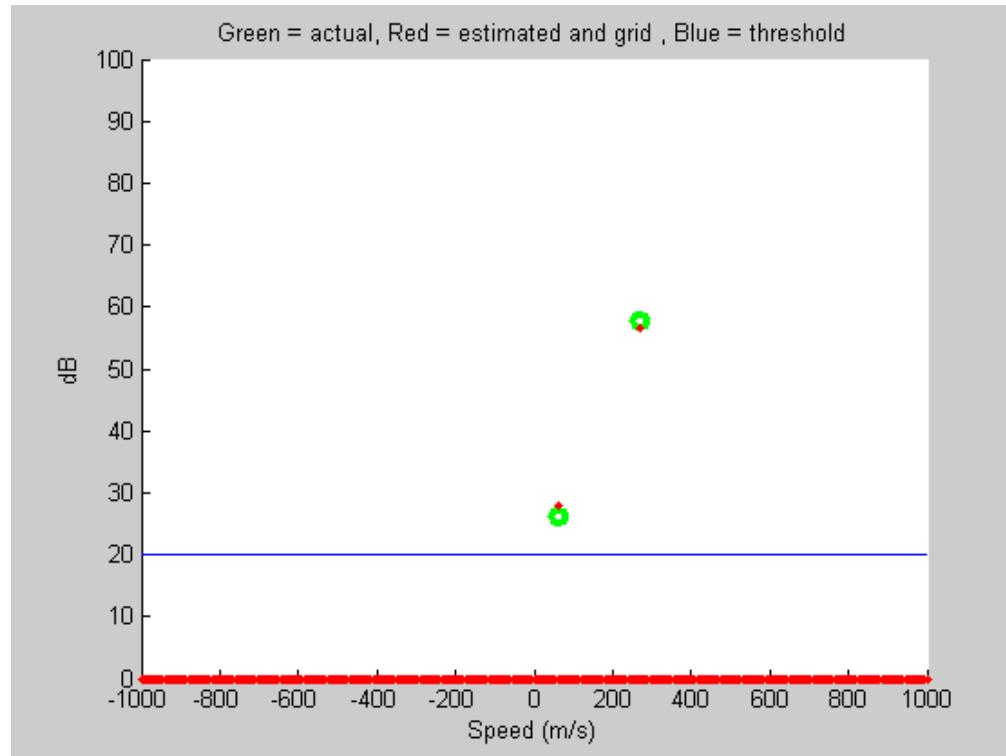
- ◆ Target 1: 60 m/s, 50 dB per burst, Swerling 2 fluctuation
- ◆ Target 2: 65 m/s, 20 dB per burst, Swerling 2 fluctuation

• $v_{target_2} = v_{target_1} + \frac{1}{2} \cdot \delta v$



Scenario 3 : 2 targets separated by 1 ambiguous speed

- ◆ Target 1: 60 m/s, 50 dB per burst, Swerling 2 fluctuation
 - ◆ Target 2: 270 m/s, 20 dB per burst, Swerling 2 fluctuation
- $v_{target_2} = v_{target_1} + V_{ambiguous_3}$



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Overall results analysis

- ◆ In all cases, Compressive Sensing
- ◆ Detects both targets
- ◆ Estimates perfectly target positions
 - While solving Doppler ambiguity
- ◆ Estimates correctly energy levels
 - Even in the case where both targets have a common ambiguous speed
 - Slight difference with input energies

Summary

- ◆ Compressive Sensing radar processing
- ◆ CAMP (Complex Approximate Message Passing) applied to multiple burst signal
- ◆ Simulation results
- ◆ Conclusion
 - Compressive Sensing reduces complexity
 - Compressive Sensing reduces losses
 - Further works

CAMP has been extrapolated to N-burst signals

◆ **N-CAMP**

- N-signal complex soft threshold
- Complex derivatives
- 1 single non-zero term in $d\nu_{a,j}^t$
- A few lines of code

Compressive Sensing reduces complexity

- ◆ **Compressive Sensing achieves at once detection, measurement, ambiguity solving and multiple targets discrimination**
 - Extract Before Detect
 - Easier to settle than a series of tests to solve ambiguities from hits

Compressive Sensing reduces losses

- ◆ **Detection is based on the sum of the energies over all the bursts**
 - Standard process "K over N" is based on K times a 1-burst detection
 - Reduced detection loss
- ◆ **Weighting functions are not necessary**
 - Compressive Sensing extracts weak signals below strong signal sidelobes
 - Reduced weighting loss
- ◆ **Folded echoes hardly affect target detection**
 - Even in the case where 2 targets have a common ambiguous speed
 - Reduced interfering targets loss

Further works

- ◆ Grid adaptation to target positions
- ◆ Merging Compressive Sensing and coherent functions
 - Digital Beam Forming
 - Doppler Filtering
 - Pulse Compression
- ◆ Merging Compressive Sensing and noise measurement
 - CFAR (Constant False Alarm Rate)
 - Clutter maps

Compressive sensing is a very promising approach
to improve radar processing efficiency

Thank you for your attention

Any question ?