



APPLICAZIONI DELLA TEORIA DELLA DECISIONE E DELLA STIMA

Fabrizio LOMBARDINI, Lucio VERRAZZANI

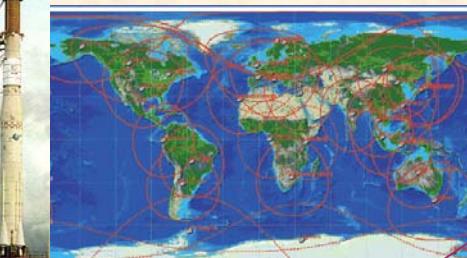
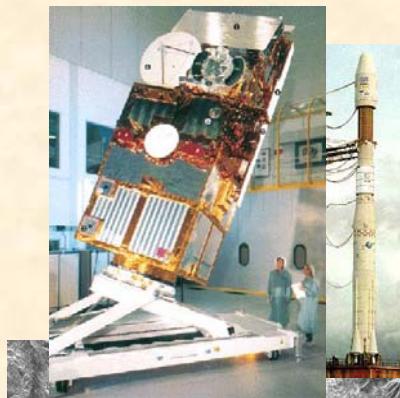
Pisa, 20 Dicembre 2007

<http://www.ing.unipi.it/~d9263/>

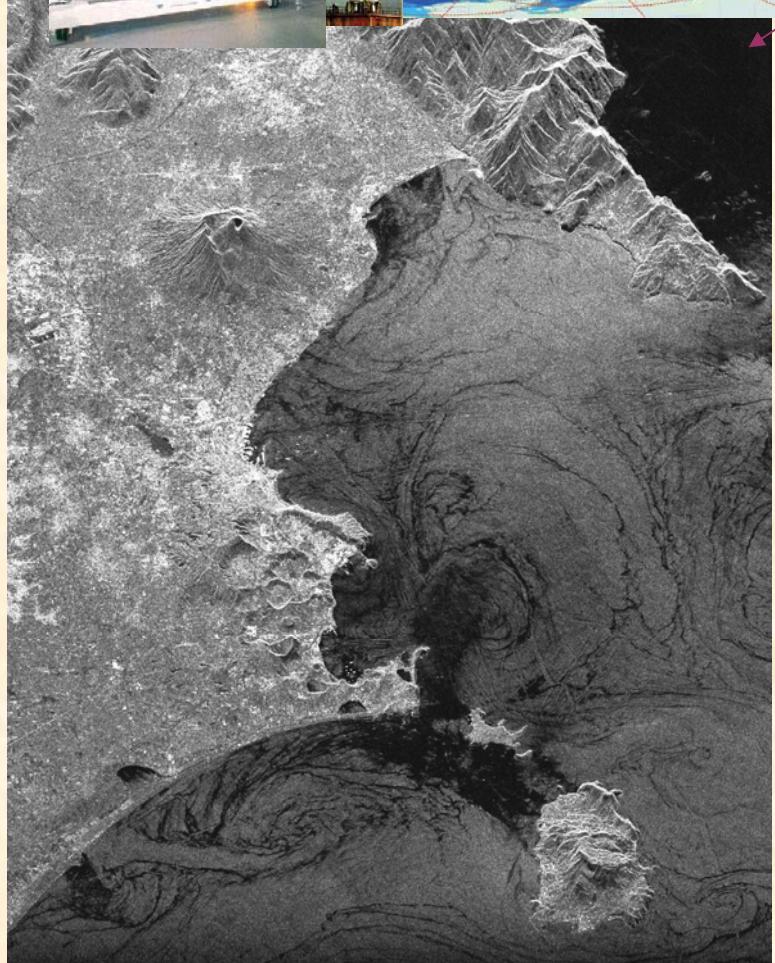
Sommario

- ⌚ Array signal processing
- ⌚ Stima spettrale in rumore moltiplicativo, a super-risoluzione, adattiva
- ⌚ Stima spettrale spazio-temporale
- ⌚ Interpolatori e filtri non stazionari
- ⌚ Bound di Cramér-Rao e di Cramér-Rao ibridi
- ⌚ Robust signal processing
- ⌚ Interferometria SAR
- ⌚ Tomografia radar 3D
- ⌚ Imaging differenziale
- ⌚ Studi di sistema per radar satellitari multicanale

Radar d'immagine (SAR, 2D)

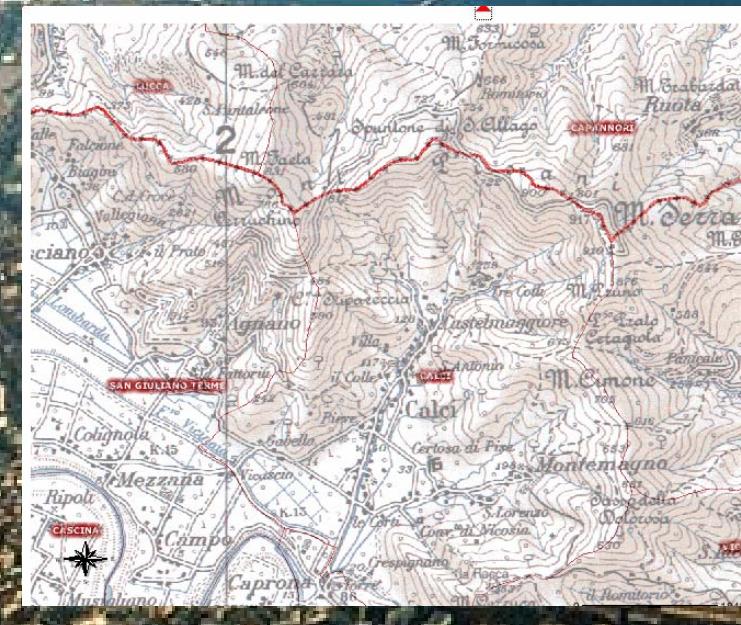


SAR satellitare

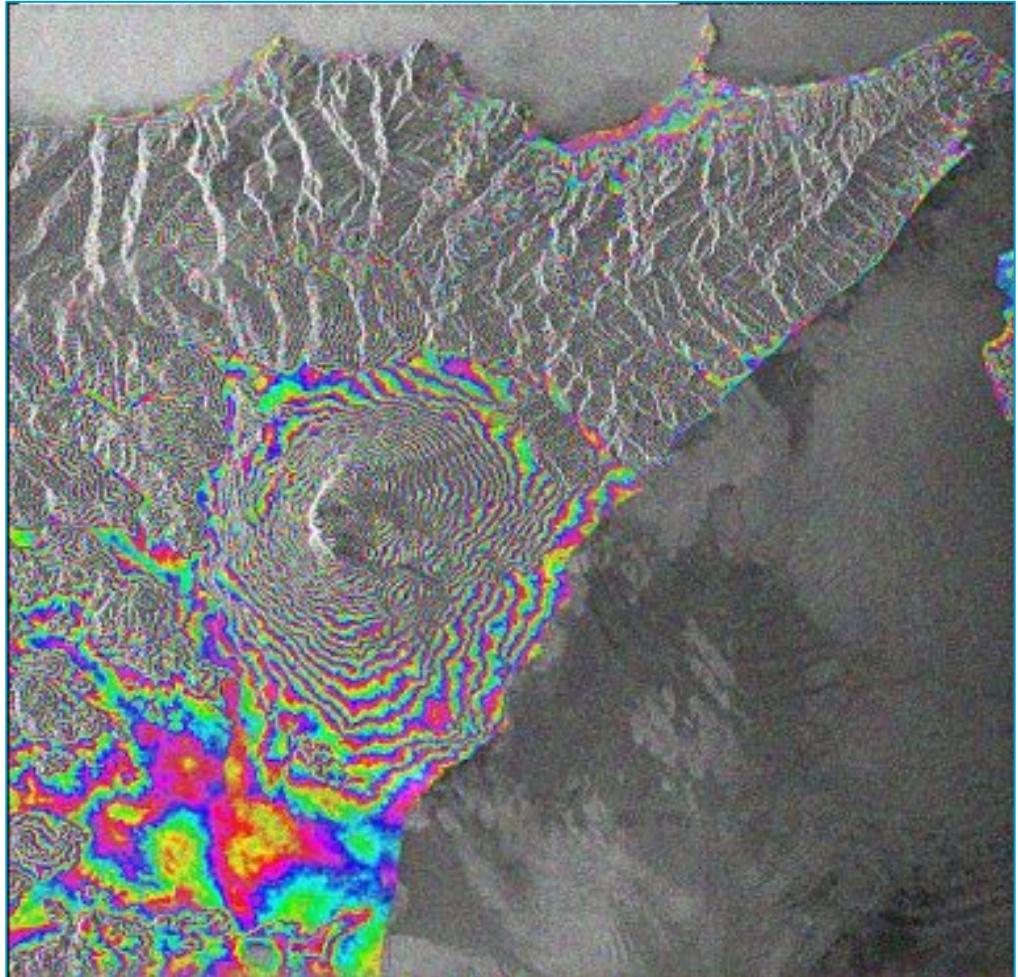
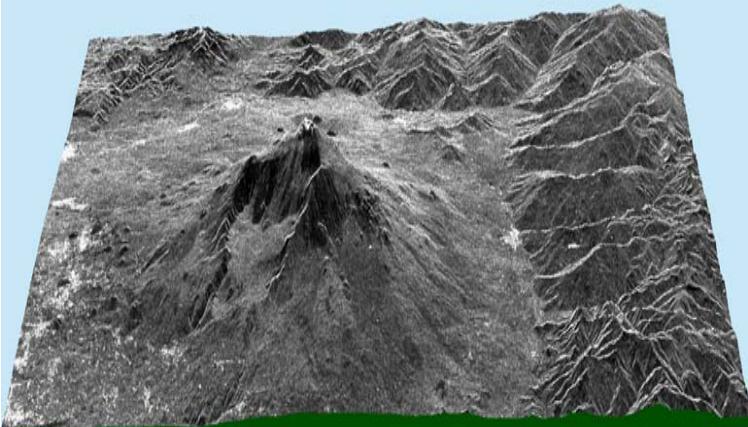
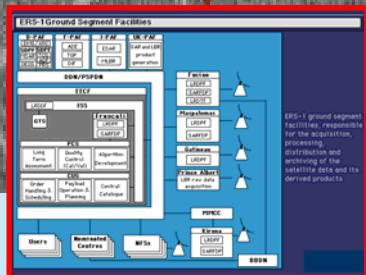
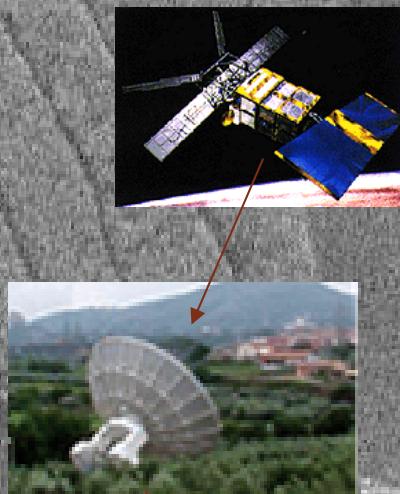


SAR aviotrasportato





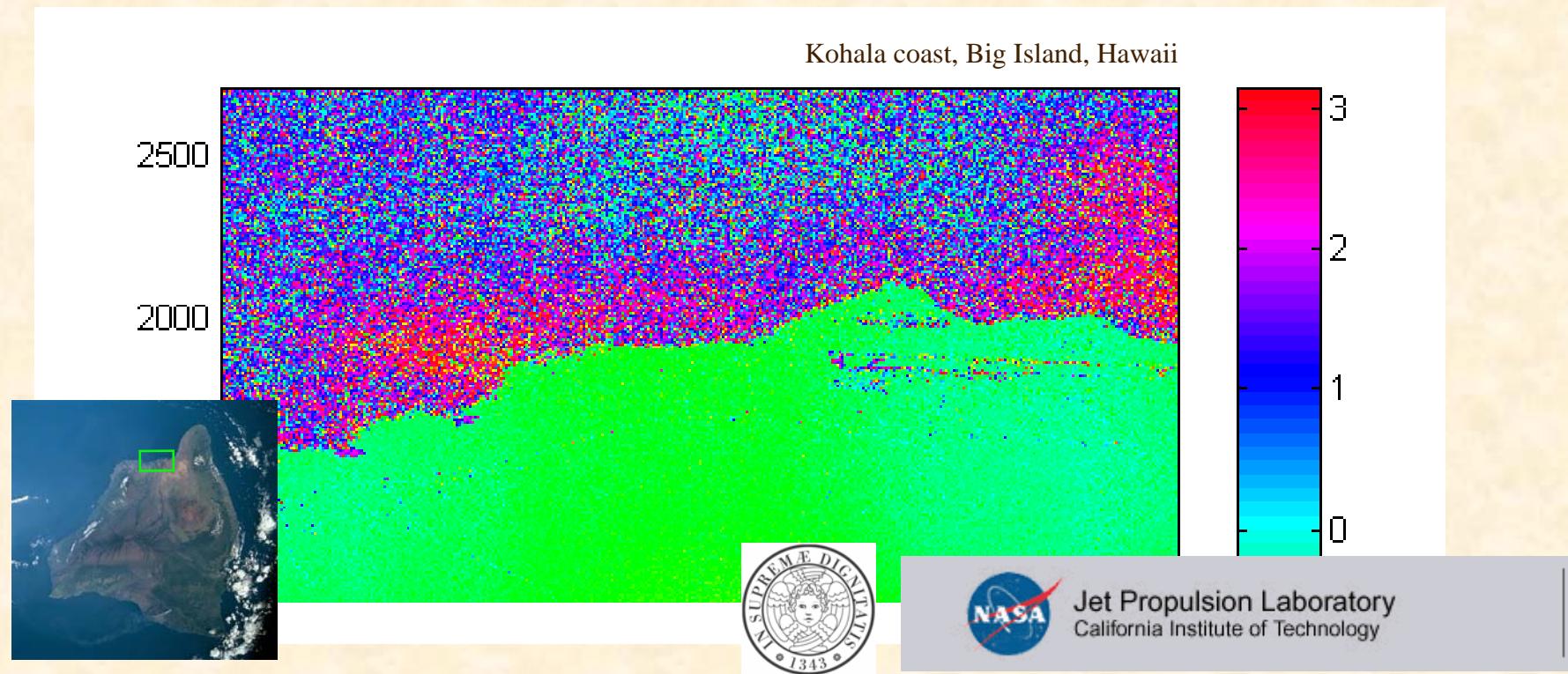
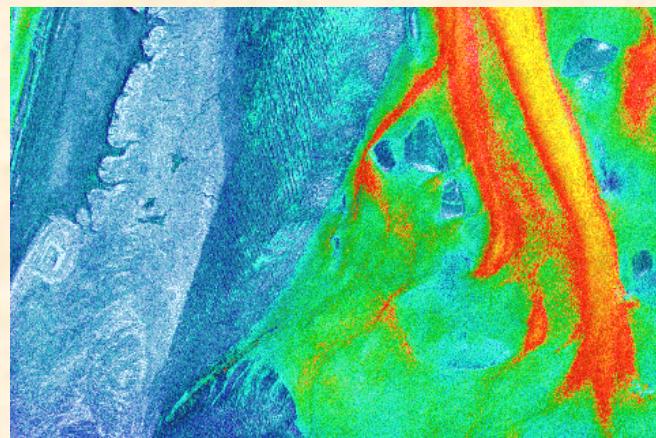
Interferometria SAR – “2.5 D” !



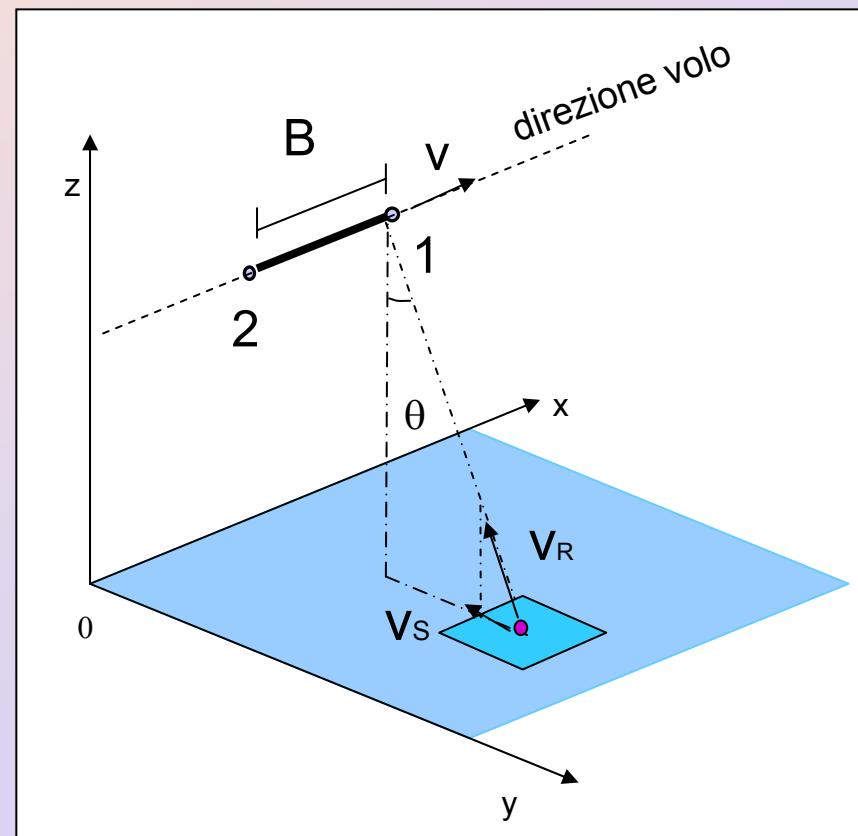
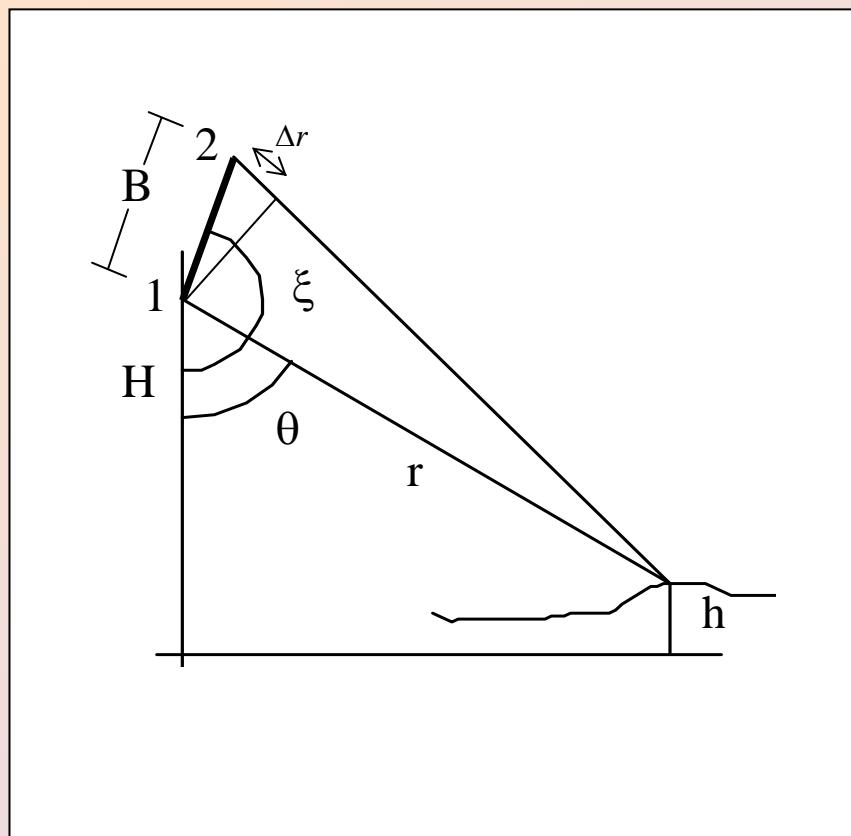
Etna, Sicilia
satellite europeo ERS-1/2



Interferometria SAR – 2D + T !



Tecnica InSAR e ATI-SAR

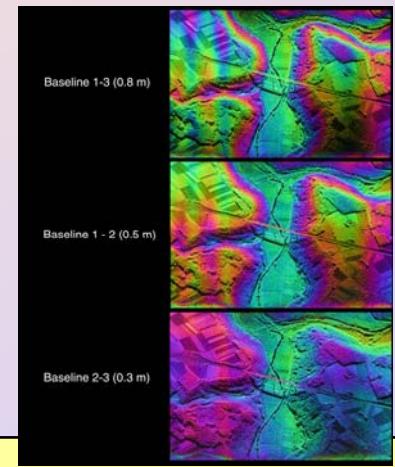
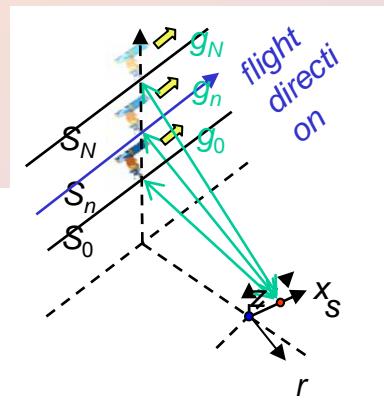
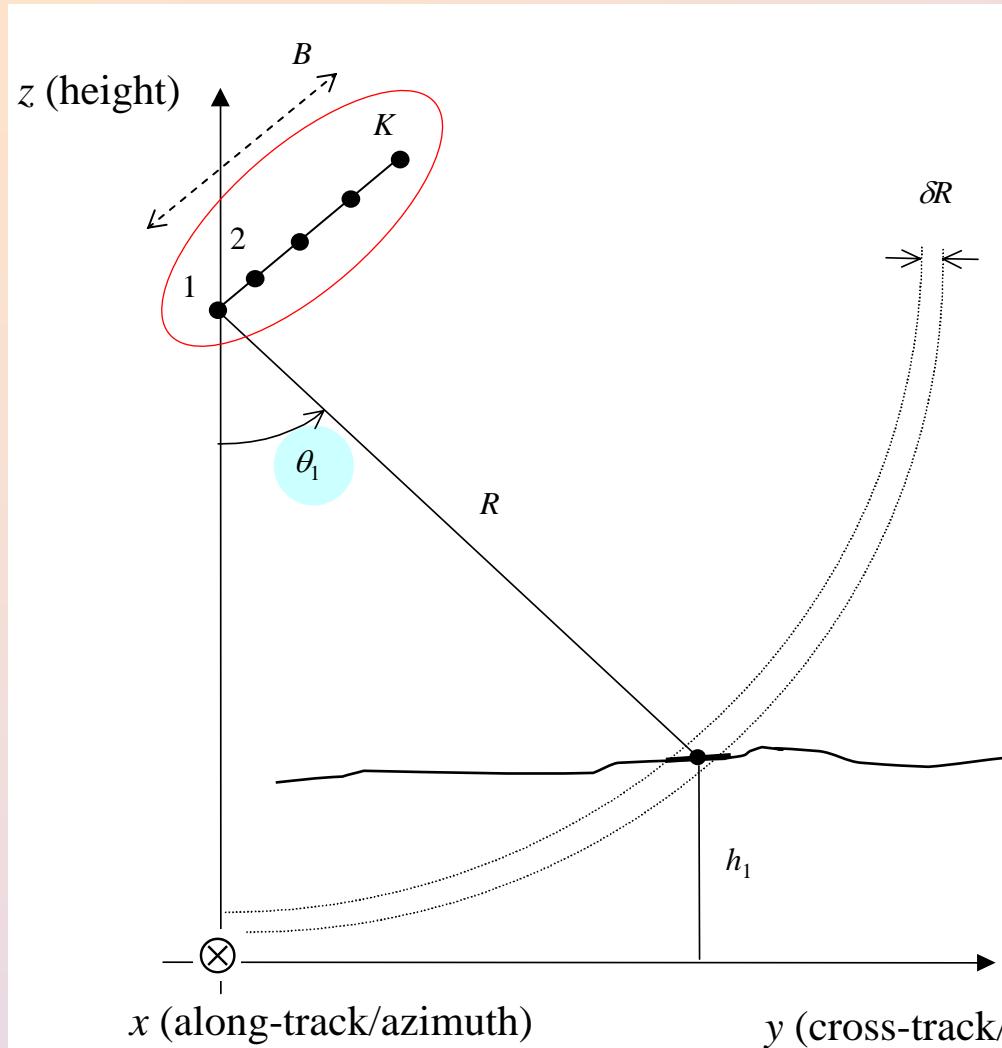


Prodotti principali:

mappe topografiche

mappe correnti marine

Multichannel (MB/MF) InSAR



Multich. improved functionalities:

- MB/MF phase unwrapping
- Reduction of data noise

Array processing

$$\mathbf{y}(n) = [y_1(n) \quad y_2(n) \quad \dots \quad y_K(n)]^T$$

data vector, $n=1,2,\dots,N$ looks

$$\mathbf{a}(\omega) = [1 \quad e^{jl_2\omega} \quad \dots \quad e^{jl_K\omega}]^T$$

steering vector

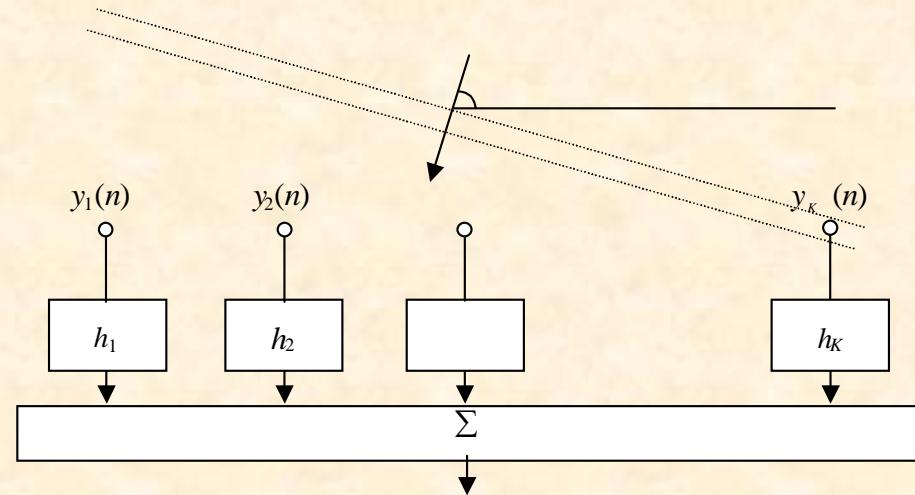
Spatial frequency!

$$\hat{P}_B(\omega) = \frac{1}{N} \sum_{i=1}^N \left| \underbrace{\frac{1}{K} \mathbf{a}^H(\omega) \mathbf{y}(n)}_{\mathbf{h}^H} \right|^2$$

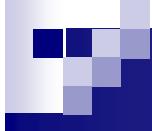
FT

conventional Beamforming
(multilook Periodogram)

$$y_F(n) = \mathbf{h}^H \mathbf{y}(n)$$

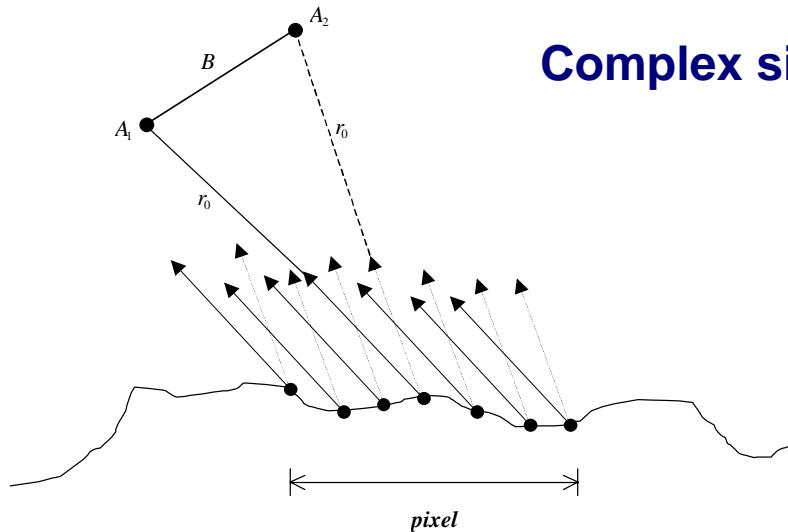


$$\hat{P}_B(\omega) = \frac{\mathbf{a}^H(\omega) \hat{\mathbf{R}}_y \mathbf{a}(\omega)}{K^2}$$



Multiplicative noise

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Complex signal (I & Q) at each pixel of the SAR image

Fully developed speckle: complex circular Gaussian distributed

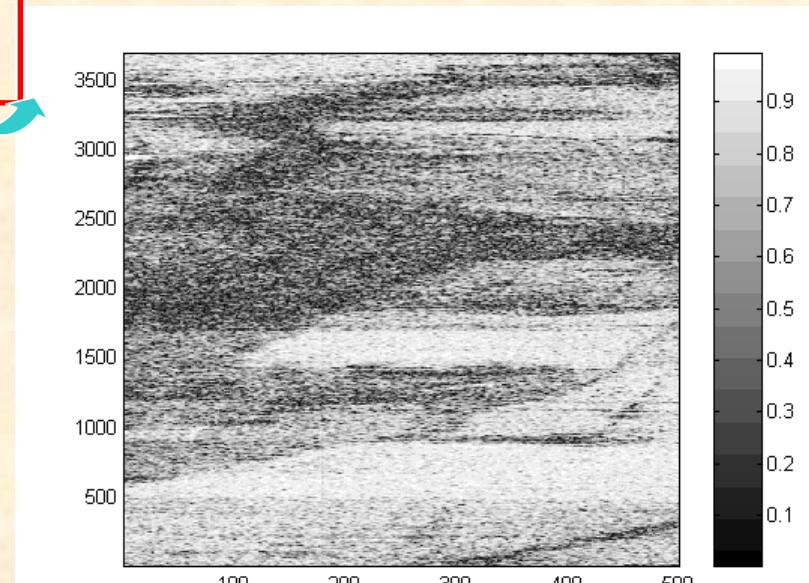
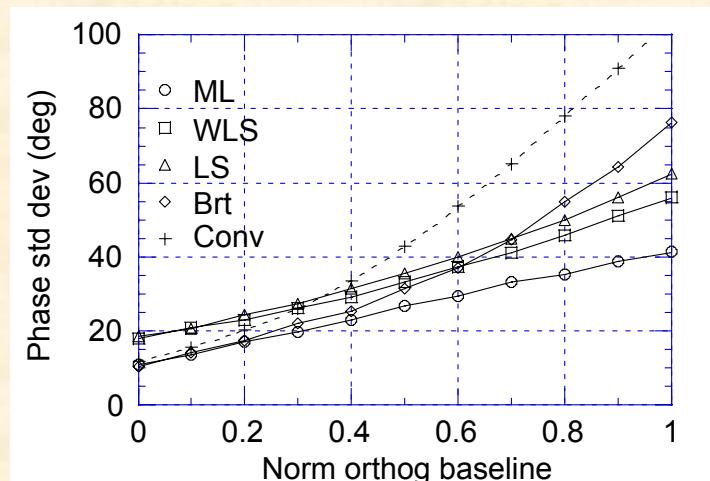
Spatial (baseline) decorrelation: non-line spatial spectrum!

Radar d'immagine (Interferometria, 2.5D)

Frequency MLE
in multiplicative noise

Stima di fase interferometrica (frequenza spaziale)

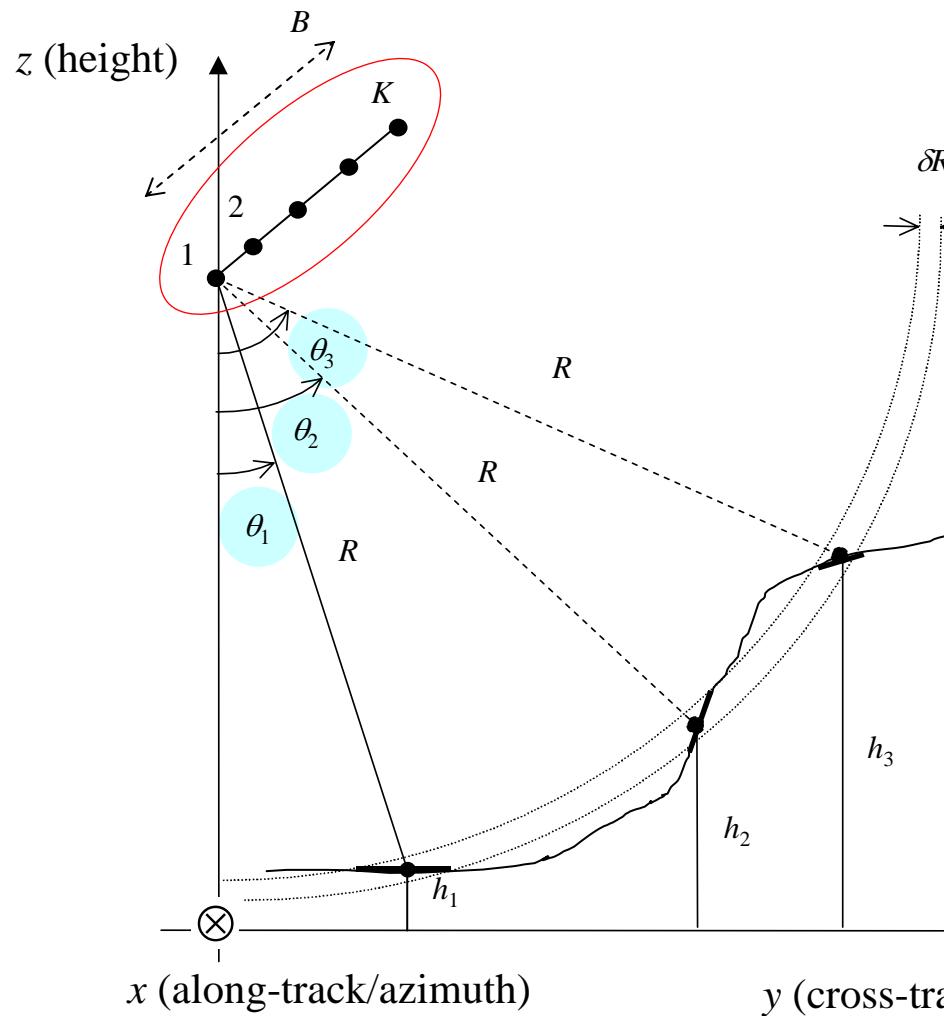
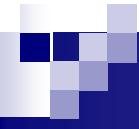
$$J(\varphi) = (\rho_{12} - \rho_{23}\rho_{13}) \operatorname{Re}\{\exp[-j\varphi(1-p)] \sum_{i=1}^N \tilde{P}_1^{*(i)} \tilde{P}_2^{(i)}\} + \\ (\rho_{13} - \rho_{12}\rho_{23}) \operatorname{Re}\{\exp[-j\varphi] \sum_{i=1}^N \tilde{P}_1^{*(i)} \tilde{P}_3^{(i)}\} + \\ (\rho_{23} - \rho_{12}\rho_{13}) \operatorname{Re}\{\exp[-j\varphi p] \sum_{i=1}^N \tilde{P}_2^{*(i)} \tilde{P}_3^{(i)}\}$$



Immagini con rivelazione spaziale

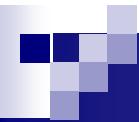


OrbiSat
da Amazônia SA



MB improved/new functionalities:

- MB phase unwrapping
- Reduction of data noise
- **Layover overcoming: multiple DOAs estimation**



Superresolution

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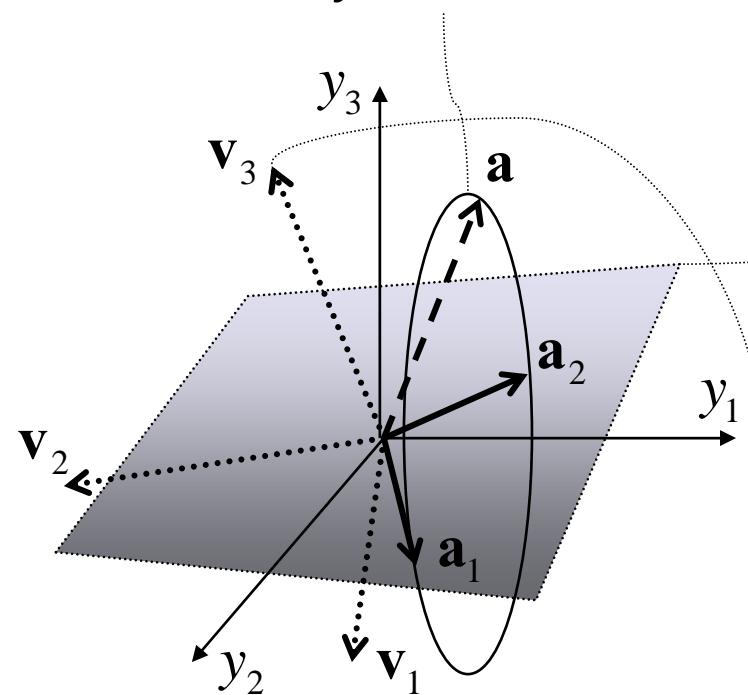
Zero spatial bandwidth

$$K = 3, N_s = 2$$

$$\sigma_v^2 = 0$$

$$\mathbf{y}(n) = \alpha_1(n)\mathbf{a}(\varphi_1) + \alpha_2(n)\mathbf{a}(\varphi_2) = \begin{bmatrix} | \\ \mathbf{a}_1 | \mathbf{a}_2 | \\ | \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \end{bmatrix}$$

Array manifold



Signal subspace

$$N_s, \{\varphi_m\}$$

$$\mathbf{C}_y = \begin{bmatrix} | \\ \mathbf{a}_1 | \mathbf{a}_2 | \\ | \end{bmatrix} \begin{bmatrix} \tau_1 \\ \tau_2 \end{bmatrix} \left[- \frac{\mathbf{a}_1^H}{\mathbf{a}_2^H} - \right]$$

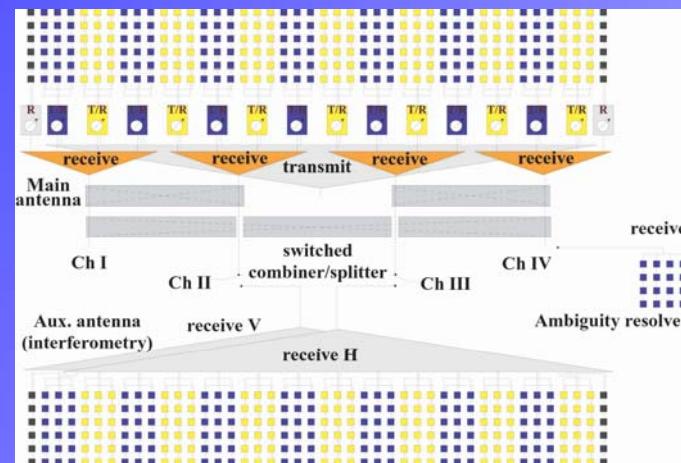
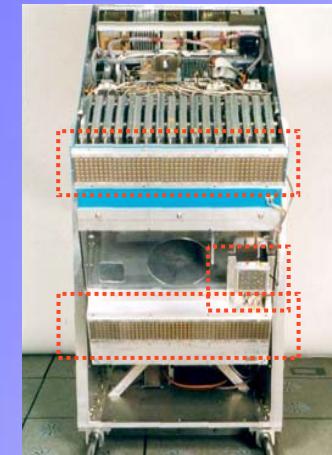
Range[\mathbf{C}_y] = Signal subspace

Parametric spectral estimation

“Noise” subspace

Airborne experimental results: FGAN-FHR AER-II

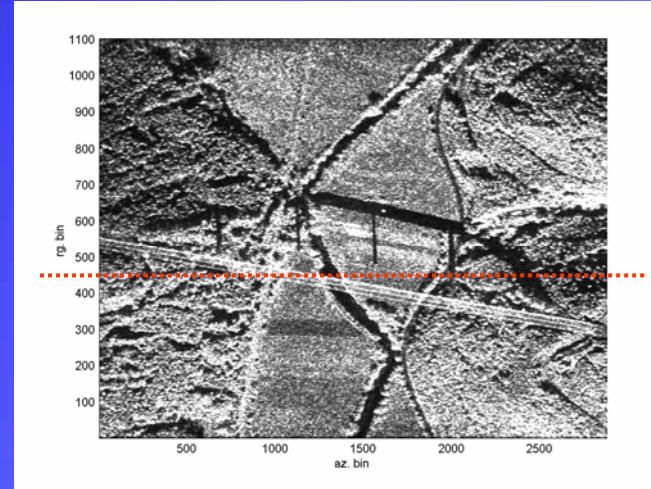
Center frequency	10.0 GHz
Resolution	1 m
Range	20 km
XTI channels	3 parallel receive channels, overall baseline 0.8 m
Antenna	Active phased array, 16 T/R modules + auxiliary antennas
Transmit power	80 W
Polarisation	VV (HH / HV / VH)



Airborne multiantenna results (FGAN AER-II)



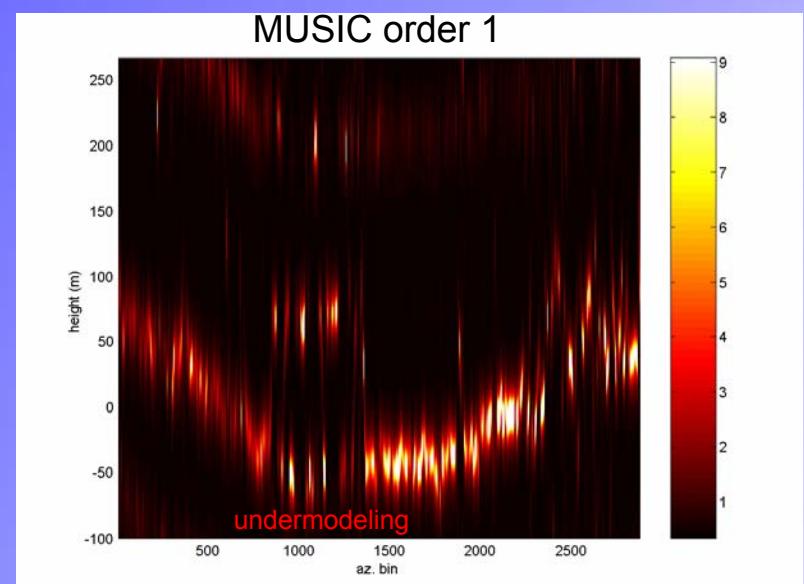
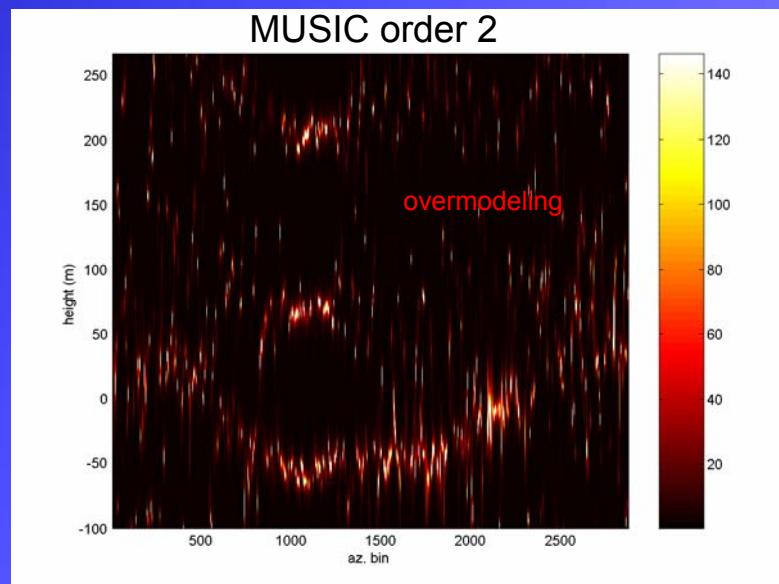
Neckar river valley and highway bridge,
Weitingen, Germany

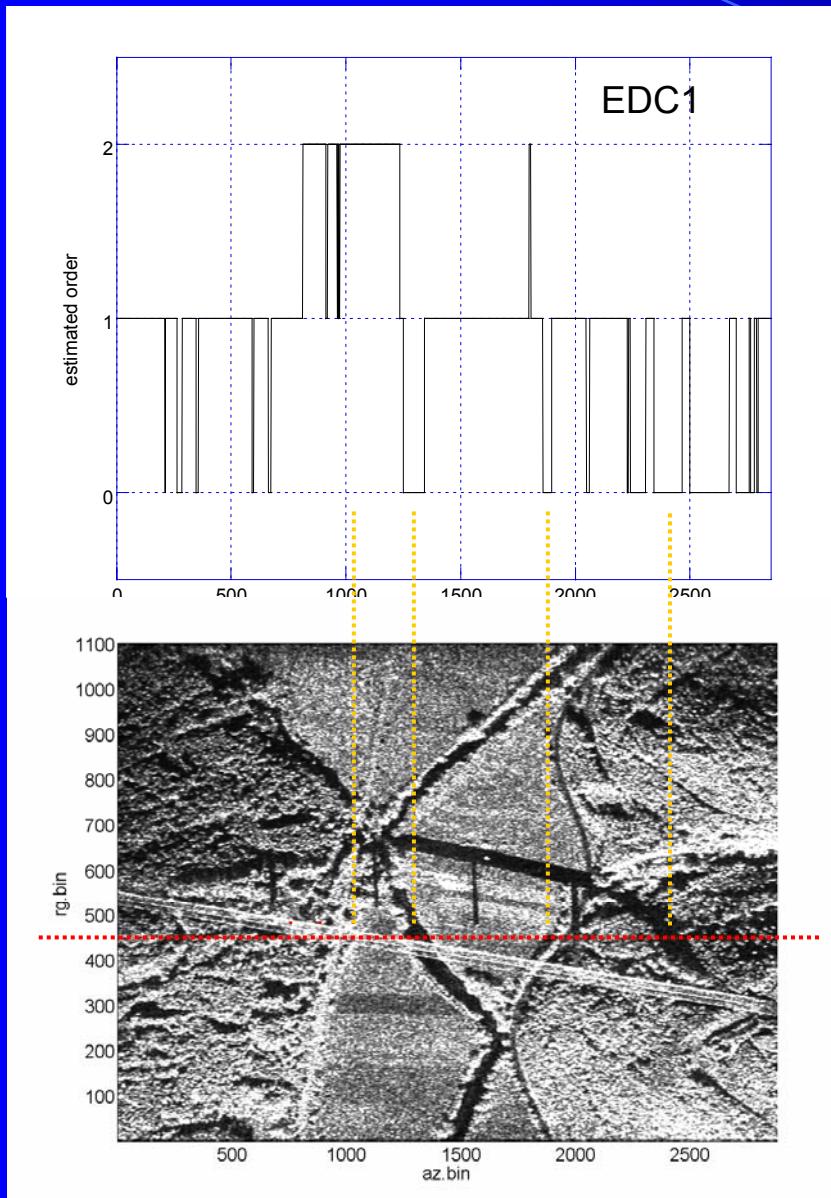


- THERMAL NOISE EQUALIZATION
- PHASE CALIBRATION
- AMPLITUDE CALIBRATION
- DERAMPING AND ORTHO BASELINE EVALUATION



*Height
resolution
98 m
Ambiguous
height
370 m*

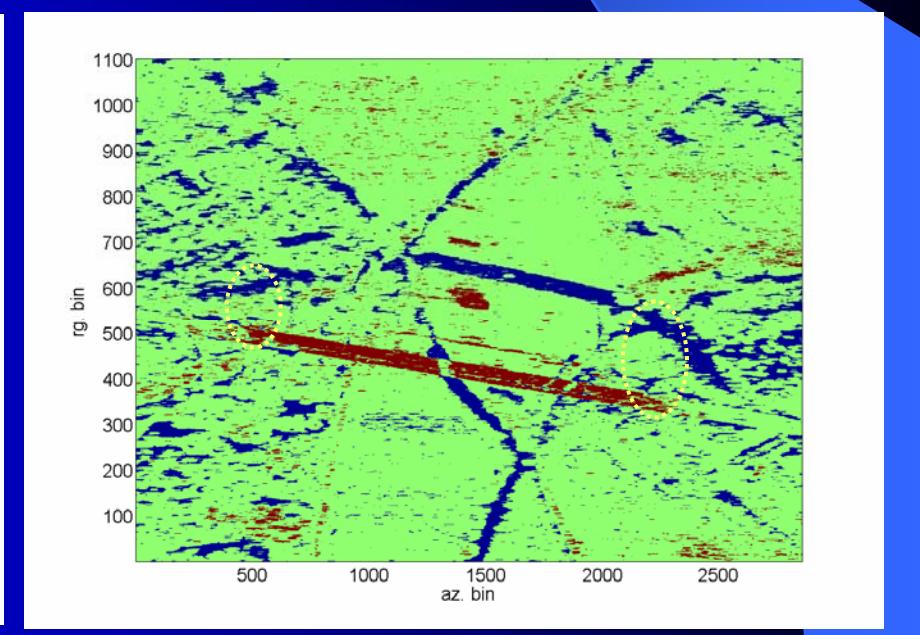




Numero di look
pari a 33 looks,
stabilizzazione
degli autovalori

Prestazione della stima dell'ordine:
90%

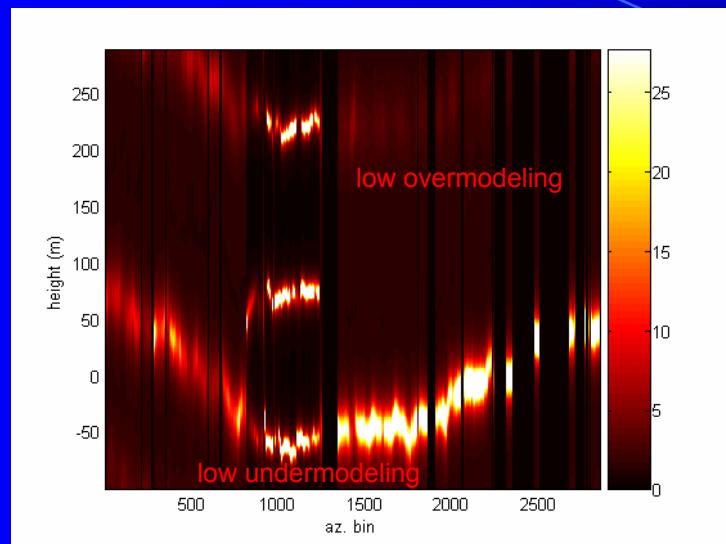
$$r, g, b: \hat{N}_s = 2, 1, 0$$



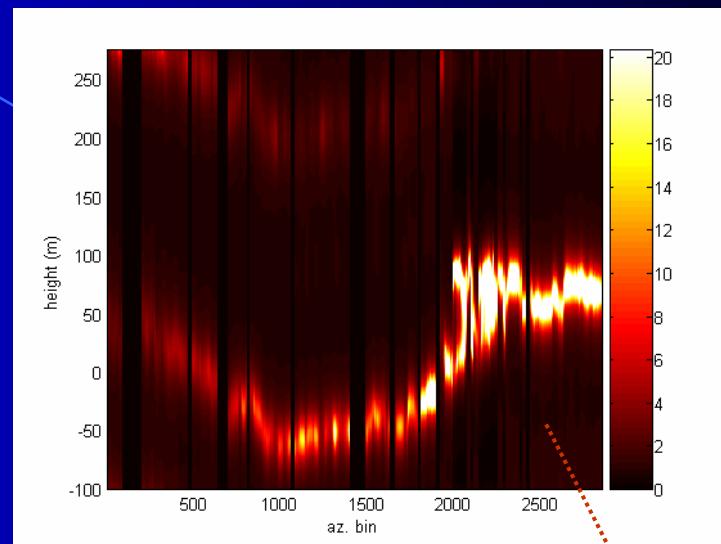


Metodo integrato

EDC-MUSIC

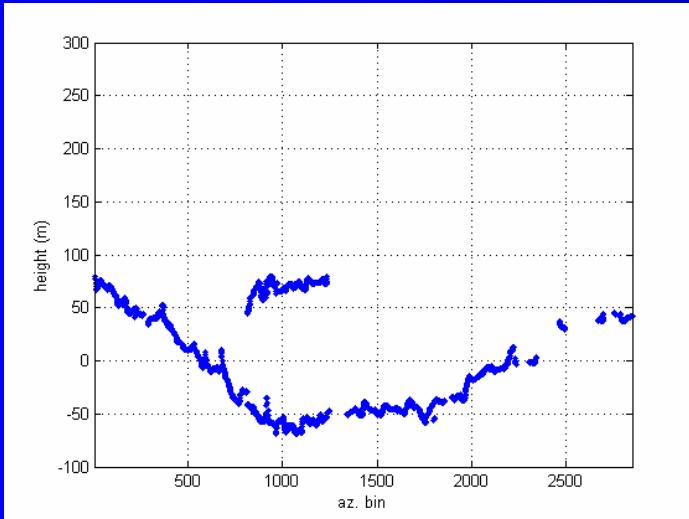


EDC-MUSIC

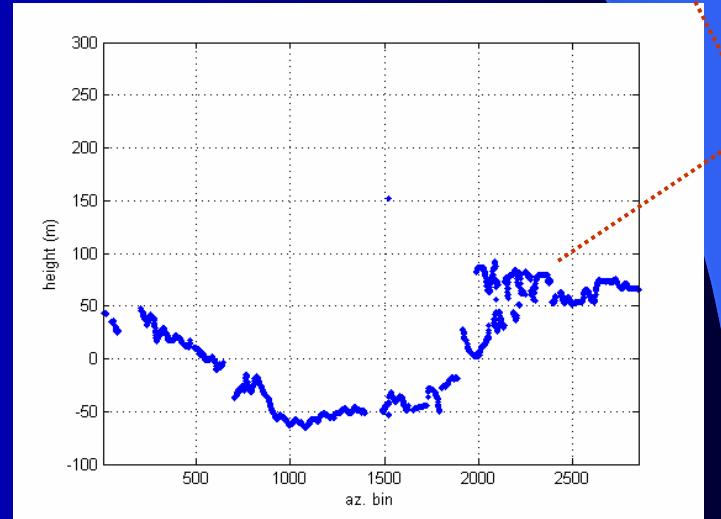


Post-processing: si sfruttano le info a priori sul settore in quota in cui sono contenuti i picchi da MUSIC

Altezze in layover estratte

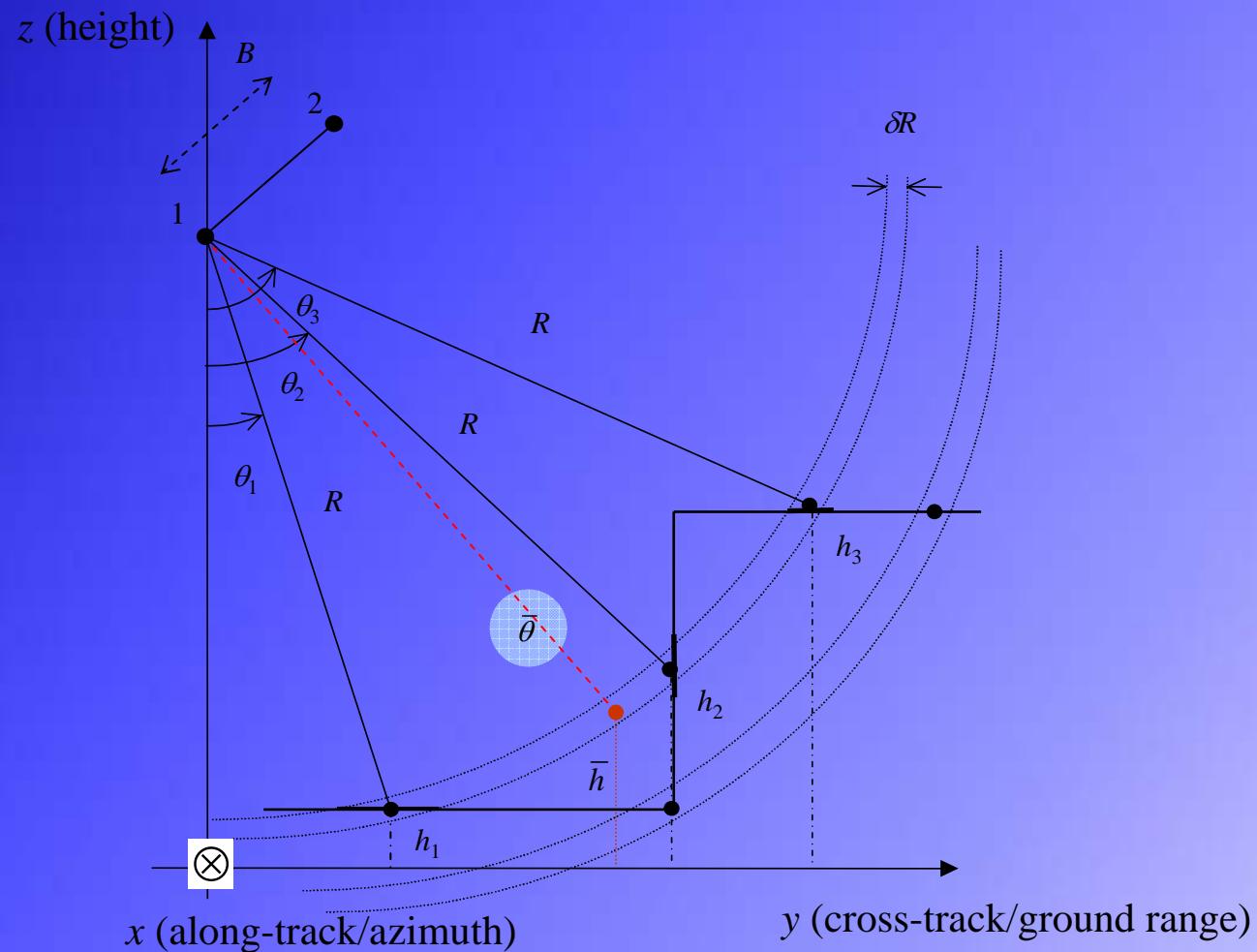


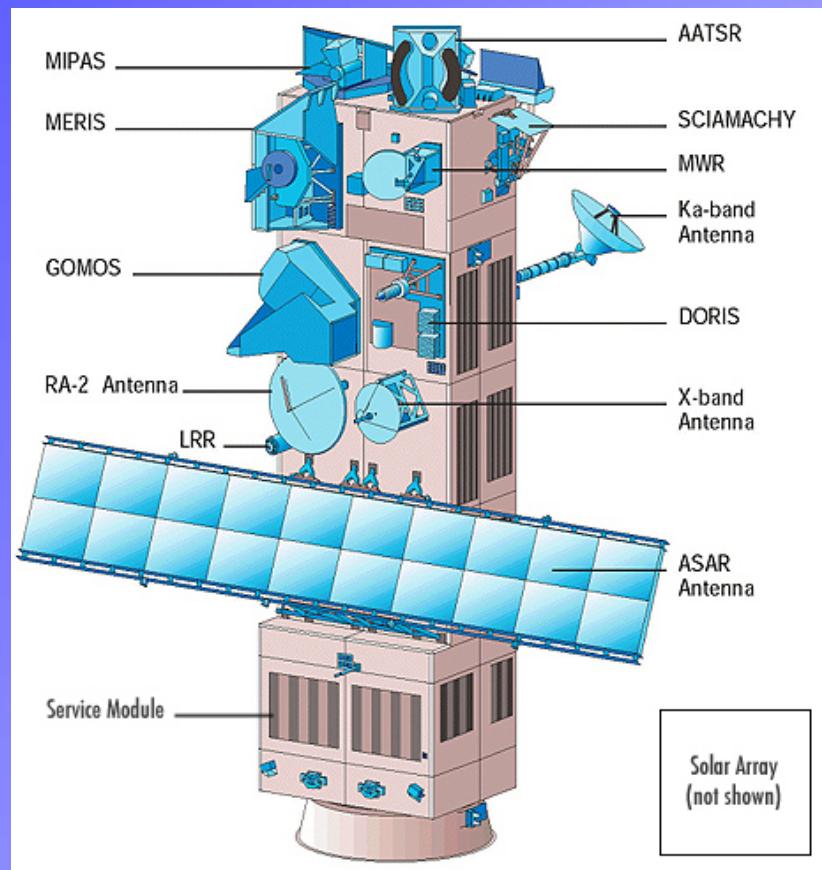
Altezze in layover estratte



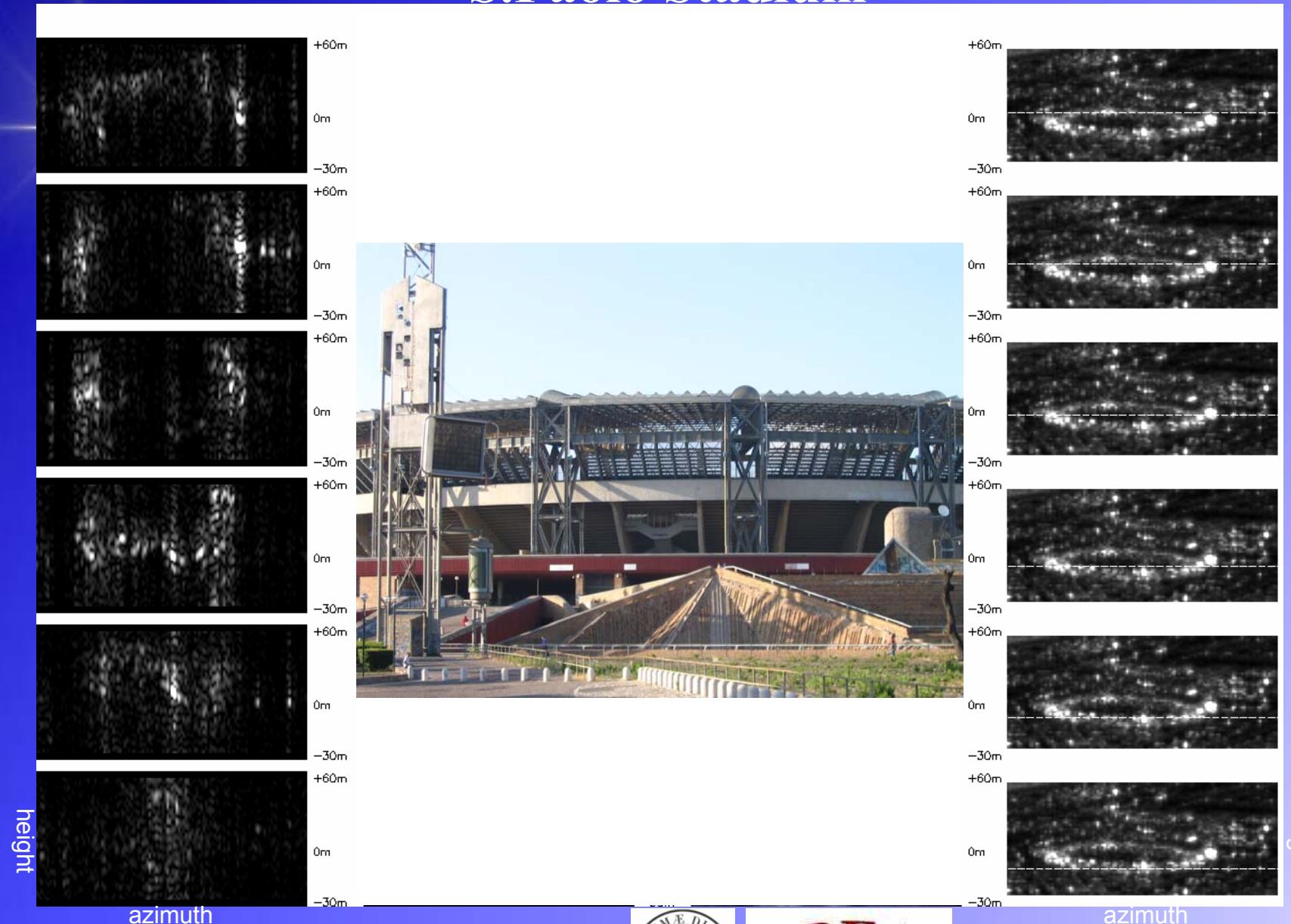
Urban DEM layover problem

Complex structures/discontinuous surfaces → **layover** → multicomponent signal





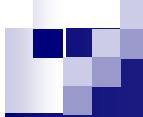
S.Paolo Stadium



IREA
Istituto per il Rilevamento
Elettromagnetico dell'Ambiente
CONSIGLIO NAZIONALE DELLE RICERCHE



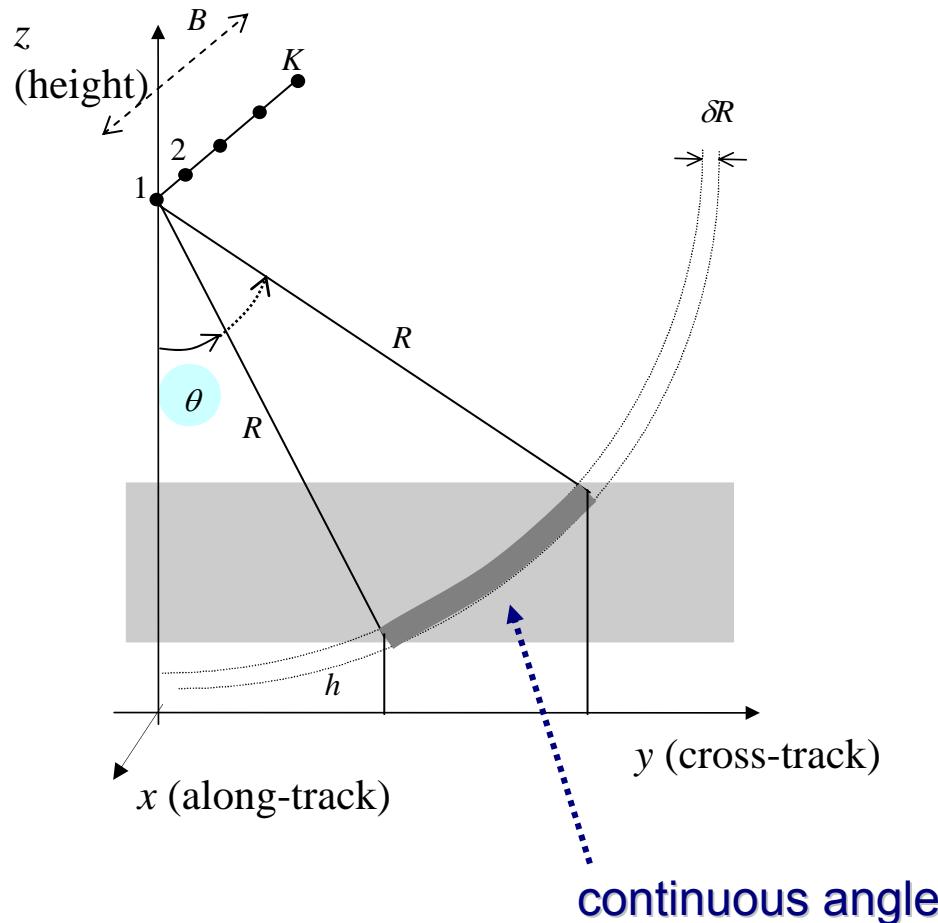




Radar Tomography: full 3-D

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Layover: the illuminated 3-D reflectivity $a(x,y,z)$ has been assumed to be concentrated to a 2-D surface



Aim of XTI-SAR:
extract τ_i and θ_i from data
from 2.5-D to 3-D

SAR-Tomography (full 3-D):
when $a(\theta)$ is continuous w.r.t. θ , not discrete

“TAC” radar!

Radar d'immagine (Tomografici, Full 3D)



E-SAR DLR
L-band, VV

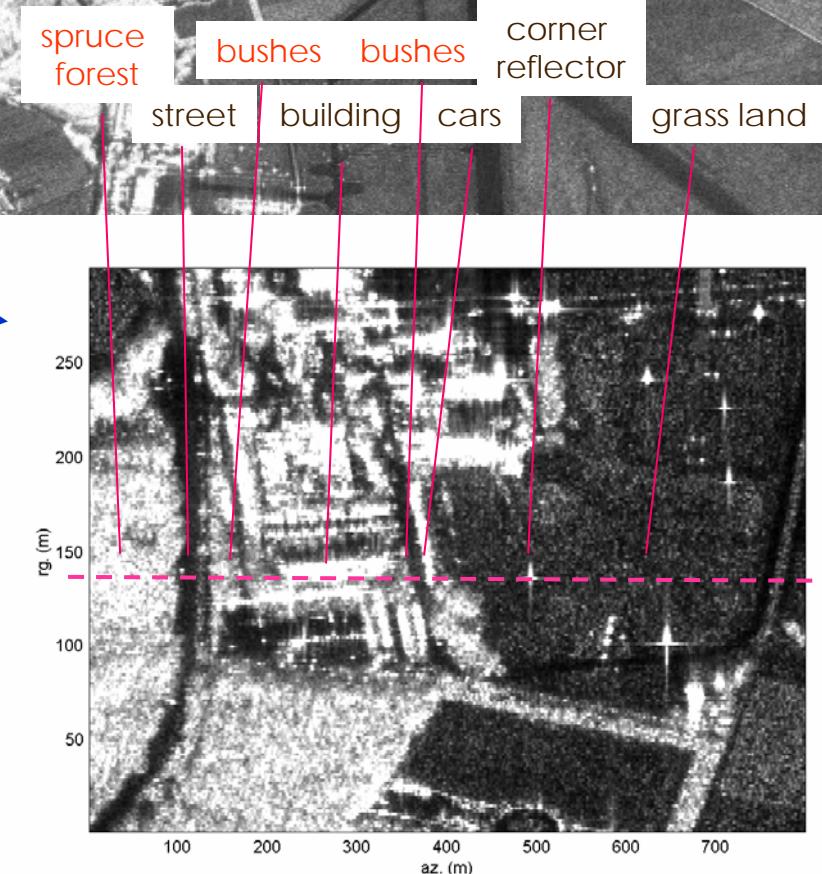
DGPS+INS

K=14 tracks
overall baseline: L=240 m

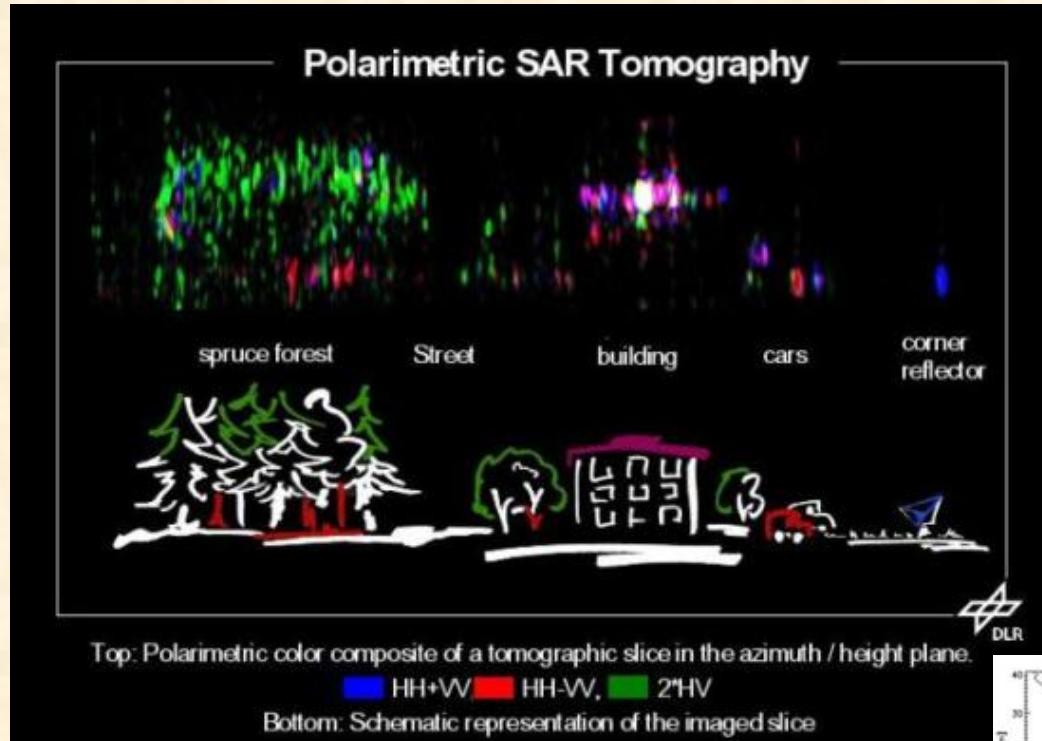
site: Oberpfaffenhofen Germany

Fourier height resolution: 3 m
unambiguous height range: 35 m

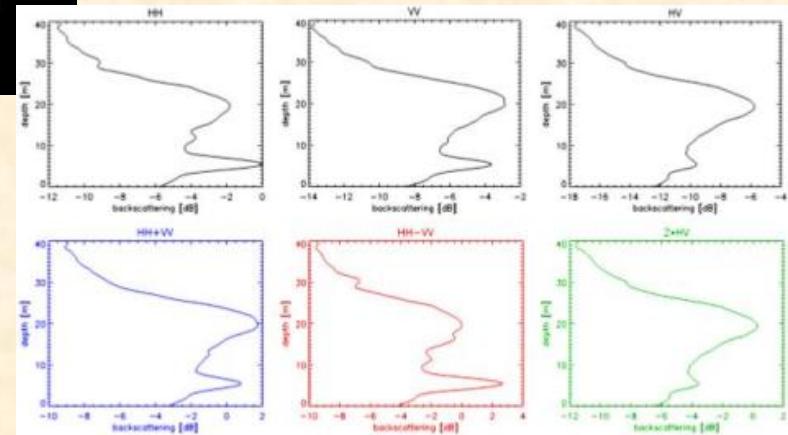
MB phase calibration:
corner reflector



Radar d'immagine (Tomografici, Full 3D)



Stima spettrale spaziale
non-parametrica



Current Problems of Radar Tomography

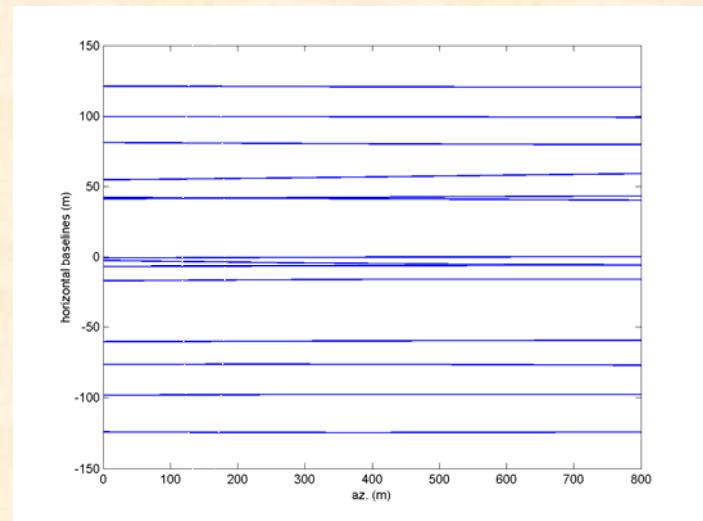
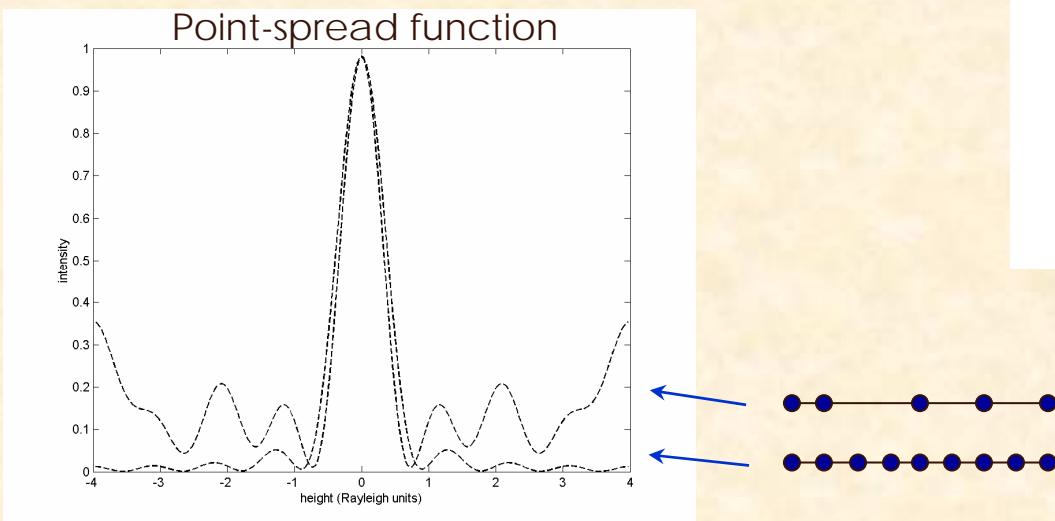
Wide field of potential applications, but:

Tradeoff ambiguity distance/z-resolution for given number K of tracks

→ Limited z-resolution, many tracks required

Typically irregular track distribution:

→ **Anomalous sidelobes** in the PSF !
(quasi-grating lobes)



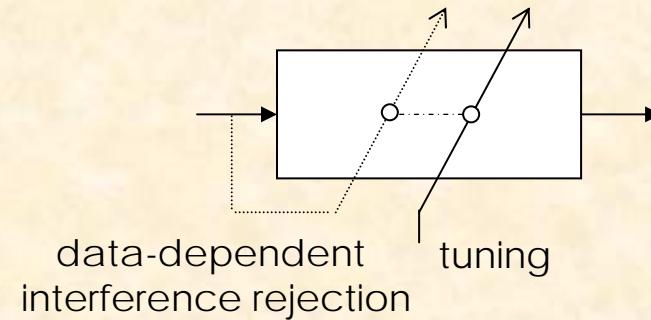
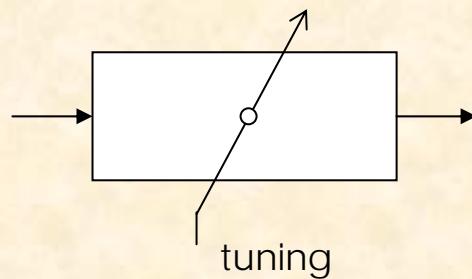
E-SAR, DLR

Stima spettrale non-parametrica *adattiva*

Lobi laterali → leakage → masking

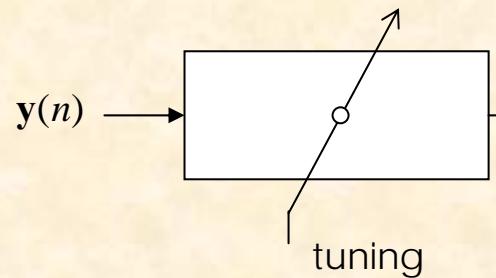


Stima spettrale non-parametrica adattiva



leakage suppression

Il filtro di Capon

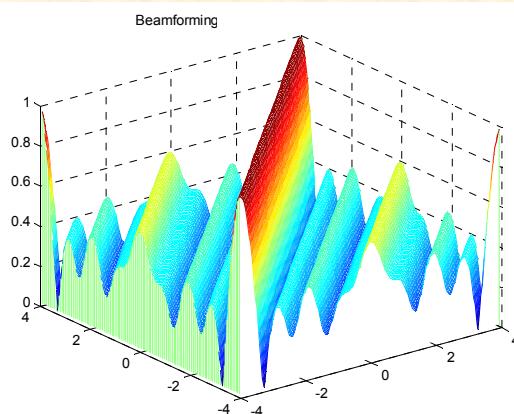


$$\mathbf{h} = \frac{\mathbf{a}(\omega)}{K}$$

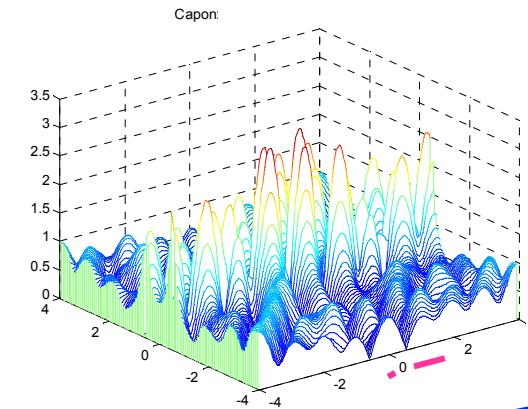
$$\hat{P}_B(\omega) = \frac{\mathbf{a}^H(\omega) \hat{\mathbf{R}}_y \mathbf{a}(\omega)}{K^2}$$

data-dependent
interference rejection

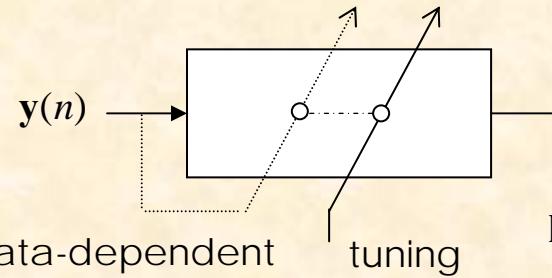
frequency response



N=32 looks SNR=9, 12 dB



reduced leakage by adaptively setting nulls !



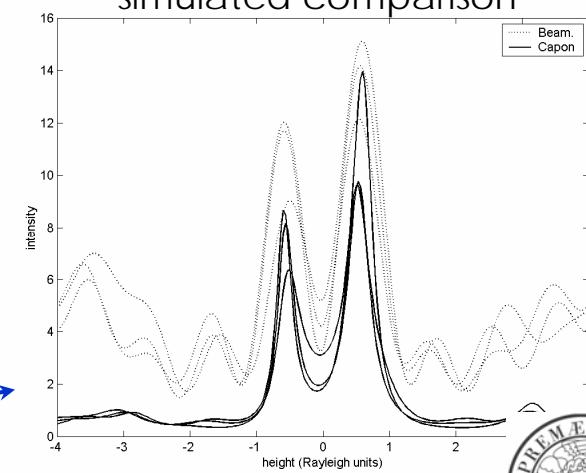
$\mathbf{y}(n)$

Stima MVU
Filtro sbiancante ad. + Period.

$$\mathbf{h} = \frac{\mathbf{R}_y^{-1} \mathbf{a}(\omega)}{\mathbf{a}^H(\omega) \mathbf{R}_y^{-1} \mathbf{a}(\omega)}$$

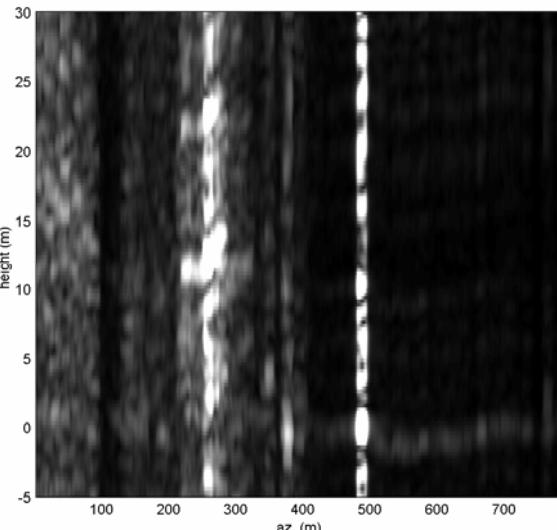
$$\hat{P}_C(\omega) = \frac{1}{\mathbf{a}^H(\omega) \hat{\mathbf{R}}_y^{-1} \mathbf{a}(\omega)}$$

simulated comparison

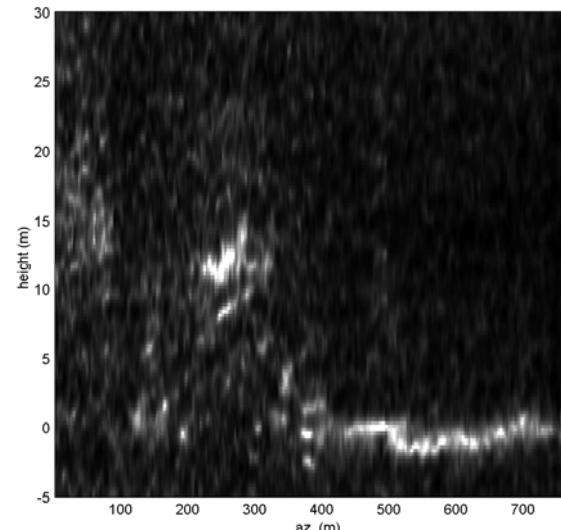


Radar d'immagine (Tomografici, Full 3D)

spatial Periodogram



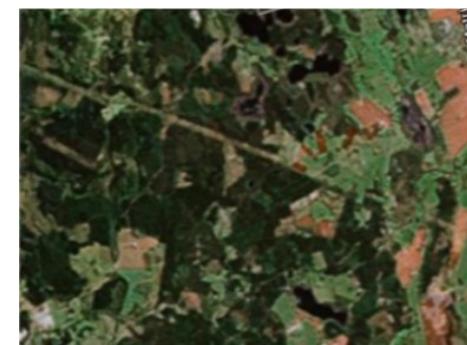
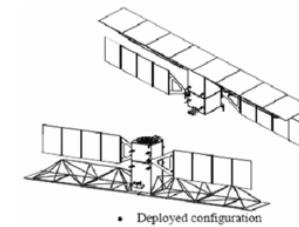
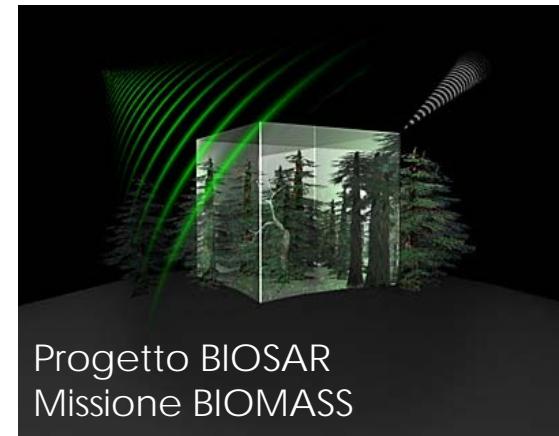
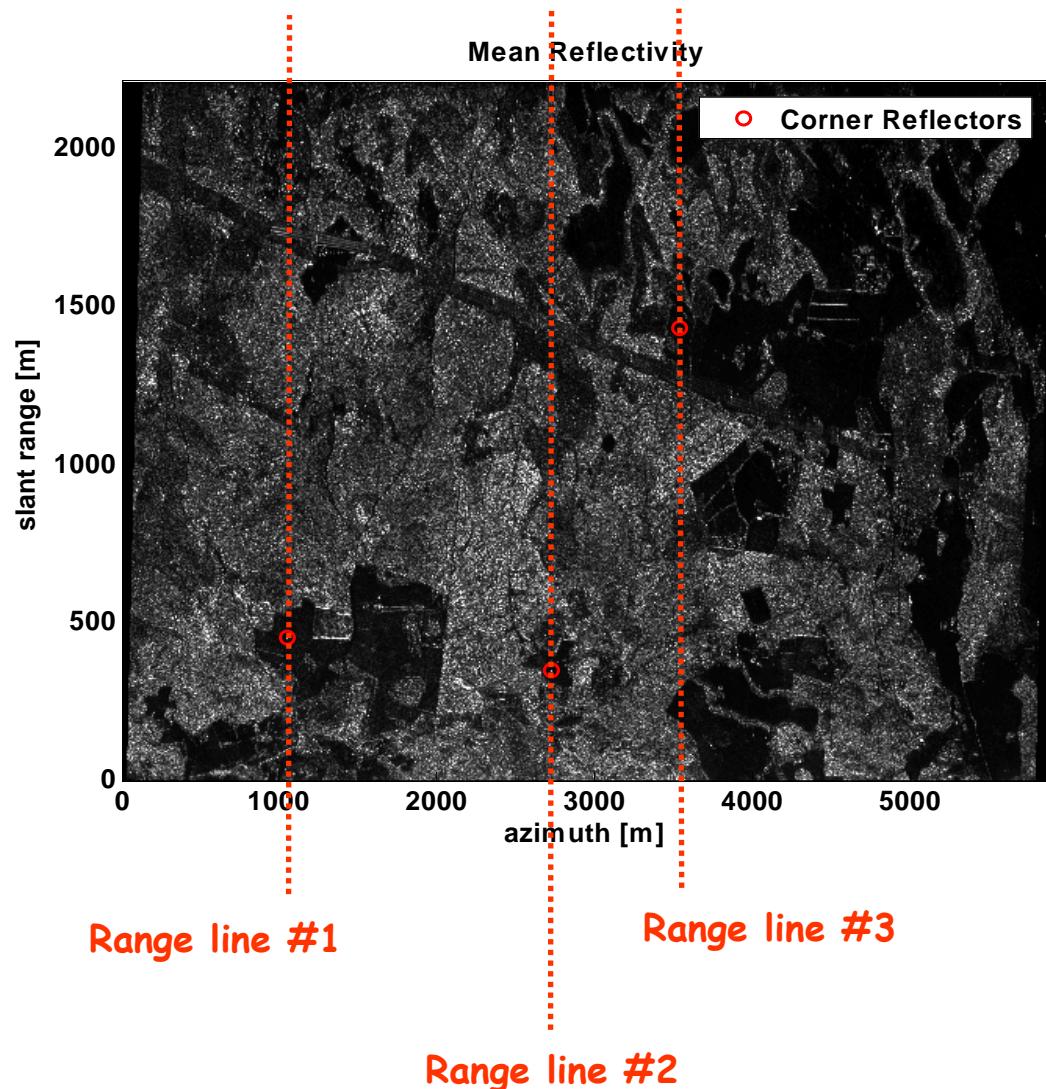
Capon



K=14 tracks

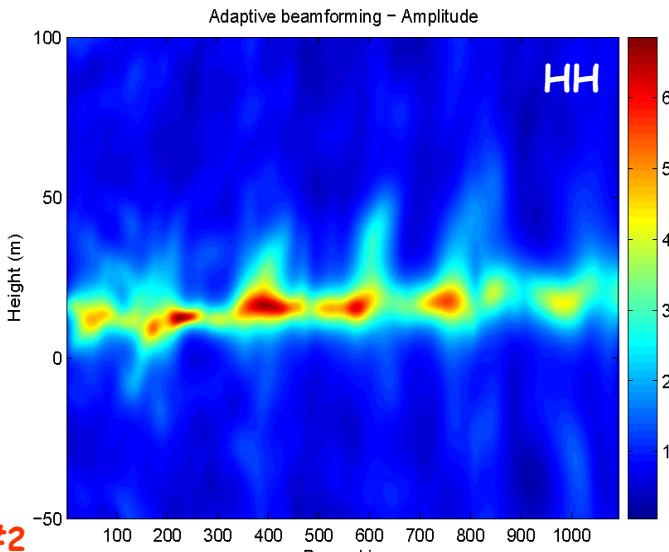
N=13 obs.



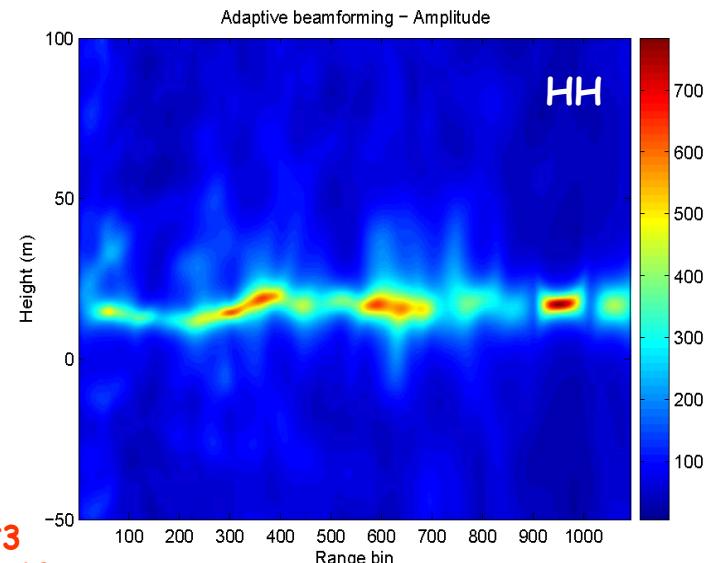
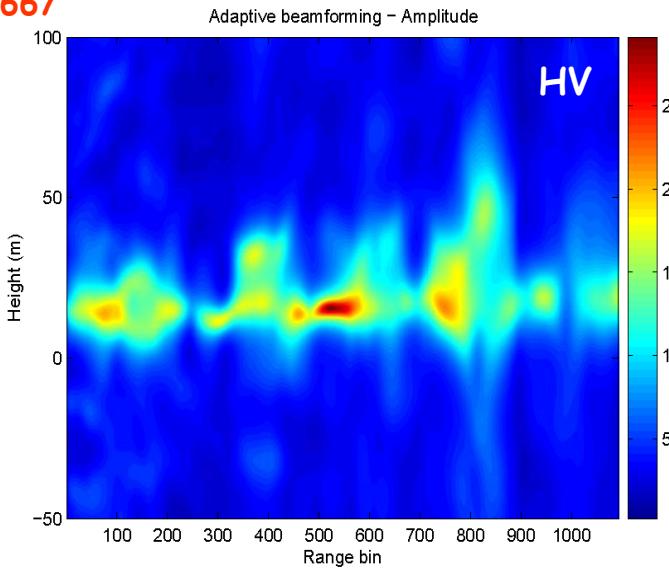




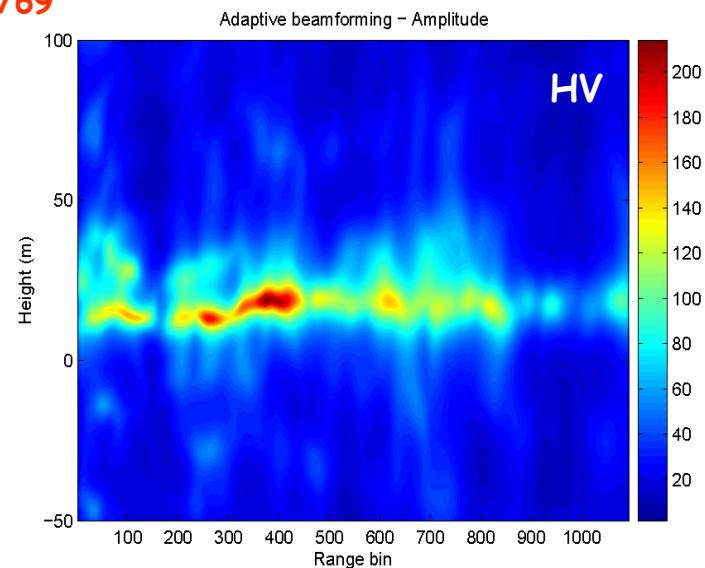
P-band data analysis



Range line #2
abs. az. bin 3667
(Capon)



Range line #3
abs. az. bin 4769
(Capon)





Differential Imaging

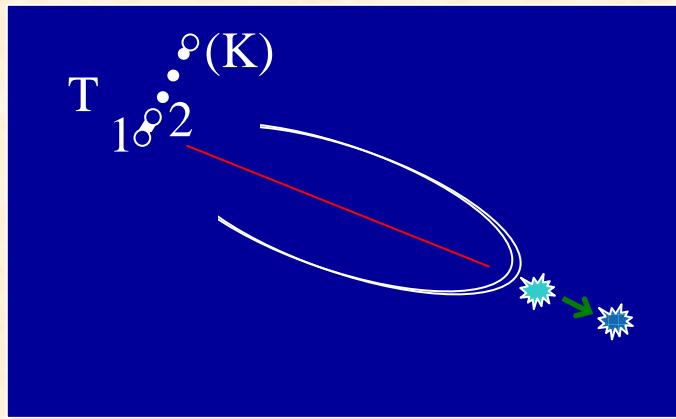
3D Tomo-SAR: multi-baseline

→ layover solution, h-backscatter profiling;
recently introduced

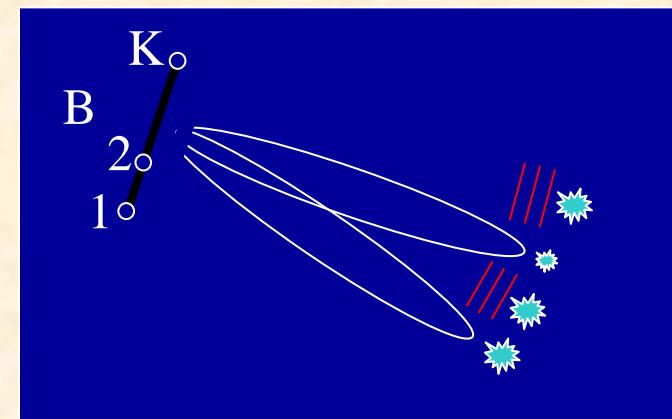
D-InSAR: *multiple-pass*

→ *displacements (single h-estimation);
operational technique*

"micro-Doppler" technique
(phase time changes: **temporal** frequency)

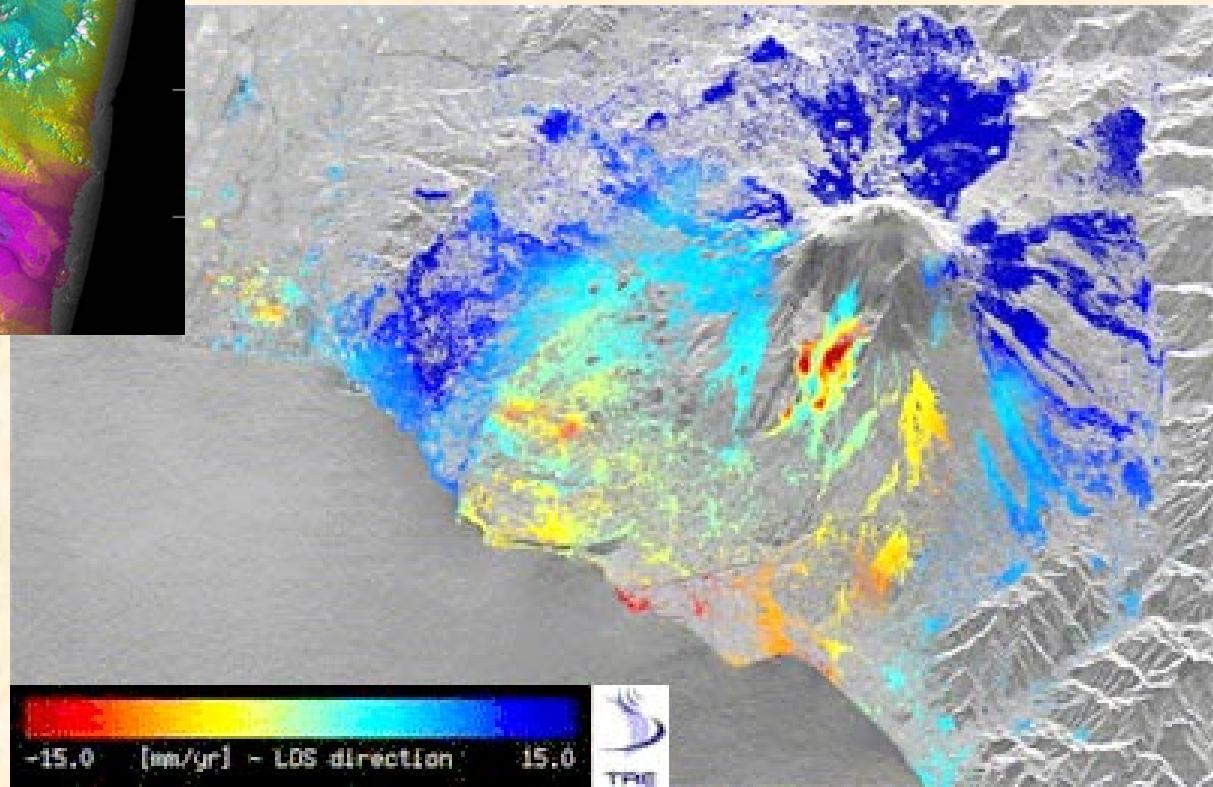
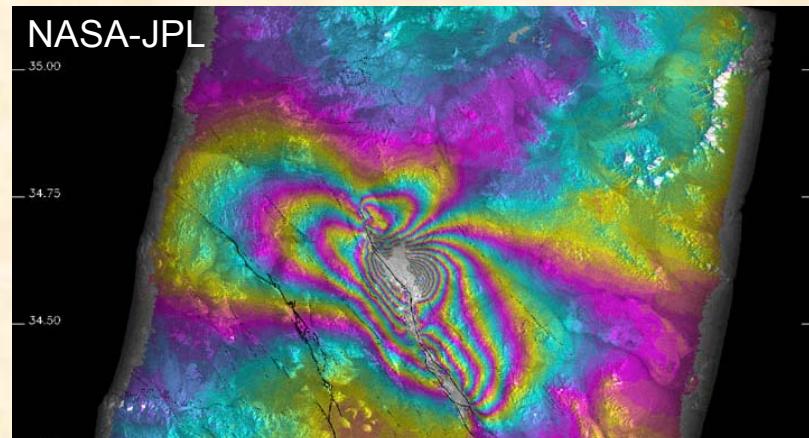


Beamforming technique
(**spatial** frequencies)



Radar d'immagine (Tecniche Differenziali)

Applicazioni: analisi subsidenze, fronti di frana, faglie, terremoti, ...



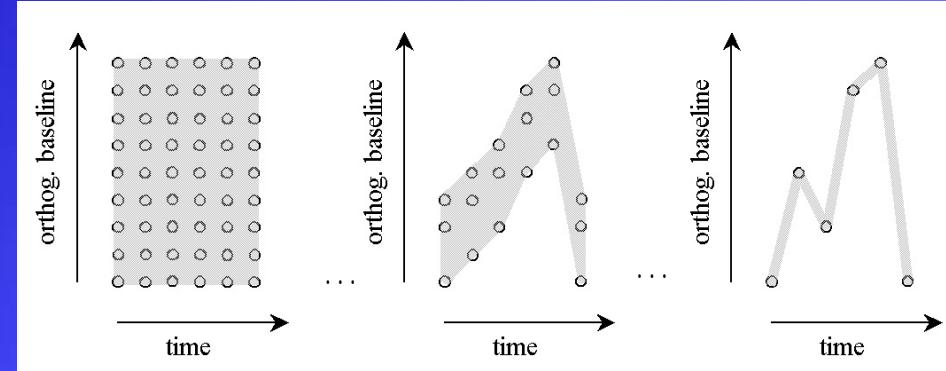
<http://eopi.esa.int/esa/esa?type=upload&ts=1071071849723&table=result&cmd=image&id=128>

“Differential Tomography”!



2D support in 2D Baseline-Time plane needed

Sparse 2D:
two-/multi-antenna
/sat. cluster
& multi-pass



“1D” Curvilinear:
plain multi-baseline
by multi-pass,
airborne/sat.

Processing Framework

after registration, compensation of phase artifacts, deramping:

$$\mathbf{A}(\omega_s, \omega_t) = \begin{bmatrix} 1 & \dots & e^{j[\omega_s B_{\perp}(1, N_p) + \omega_t t(N_p)]} \\ \vdots & \ddots & \vdots \\ e^{j\omega_s \dots} & \dots & \dots & \dots \\ & \dots & \dots & t(N_p) \end{bmatrix}$$

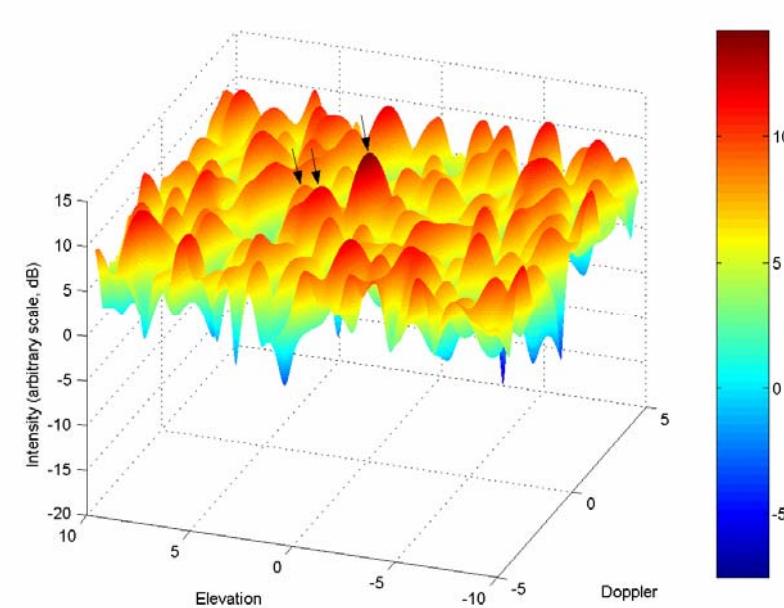
Bidimensional Baseline-Time
Spectral Estimation !
 $\mathbf{a}(\omega_s, \omega_t)$ → z (3D) + Doppler resolution cell

Challenge: **Sparse Sampling** of 2D baseline-time support -
2D z-Doppler sidelobe problems to be attacked ! ←

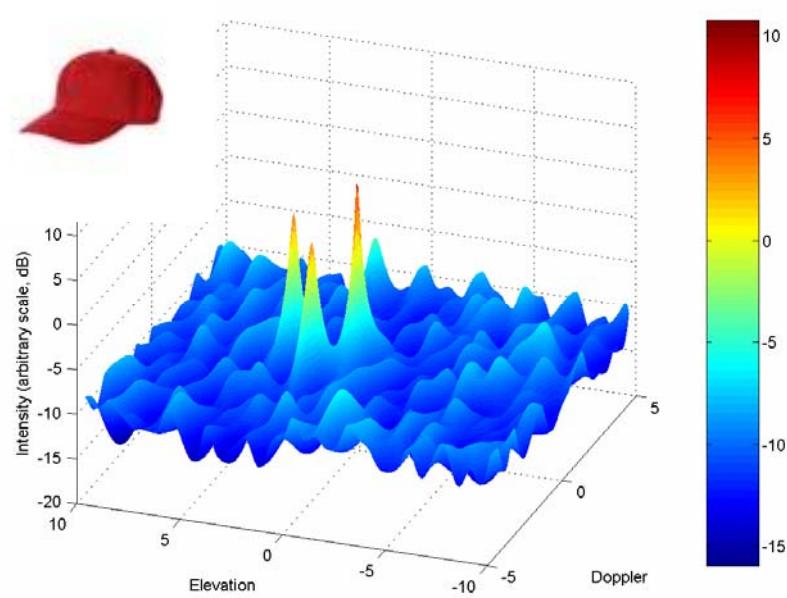
“4D Imaging”! (3D + T)

Simulated Bonn Data Set
ERS-1
10 passes
time span: 27 days
overall baseline: 1418 m

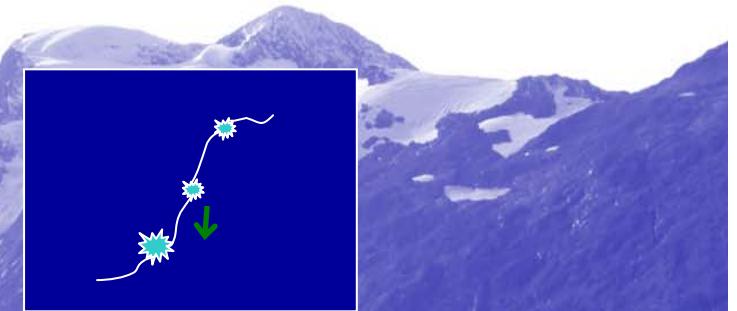
SNR=15, 12, 9 dB
negligible baseline
and temporal decorrelation;
monotonic motion jitter.



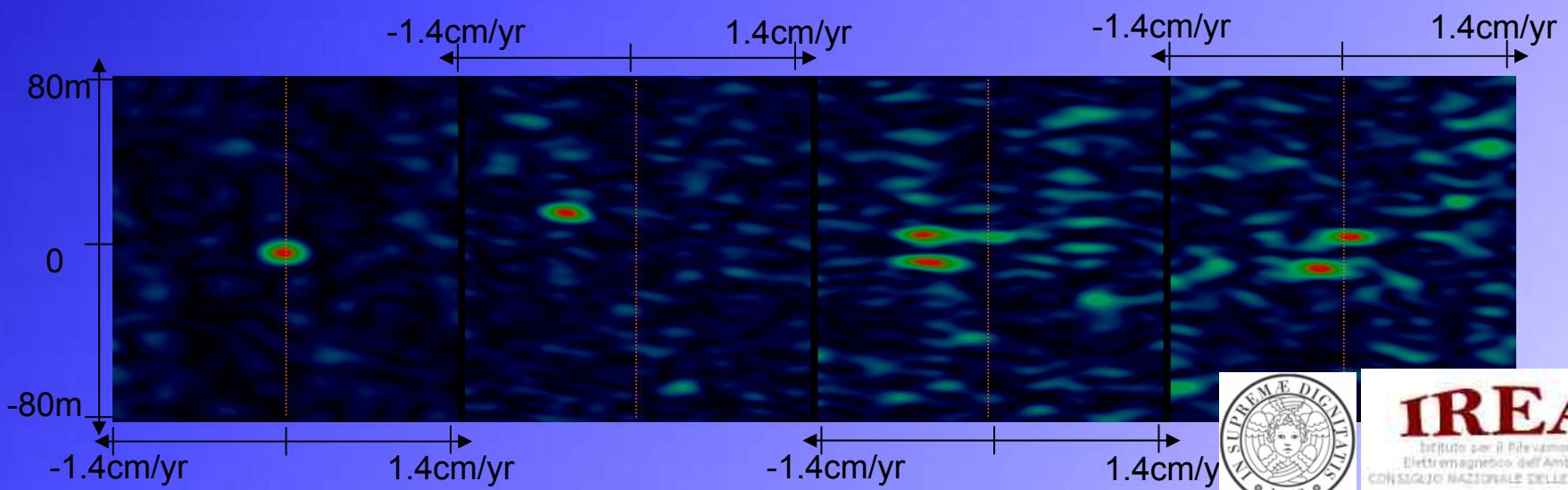
portion of Diff. + Tomo image



portion of Diff. + Tomo image



Results on real data of the ERS 1/2



IREA
Istituto per il Rilevamento
Elettromagnetico dell'Ambiente
CONSIGLIO NAZIONALE DELLE RICERCHE

Results on real data of the ERS 1/2



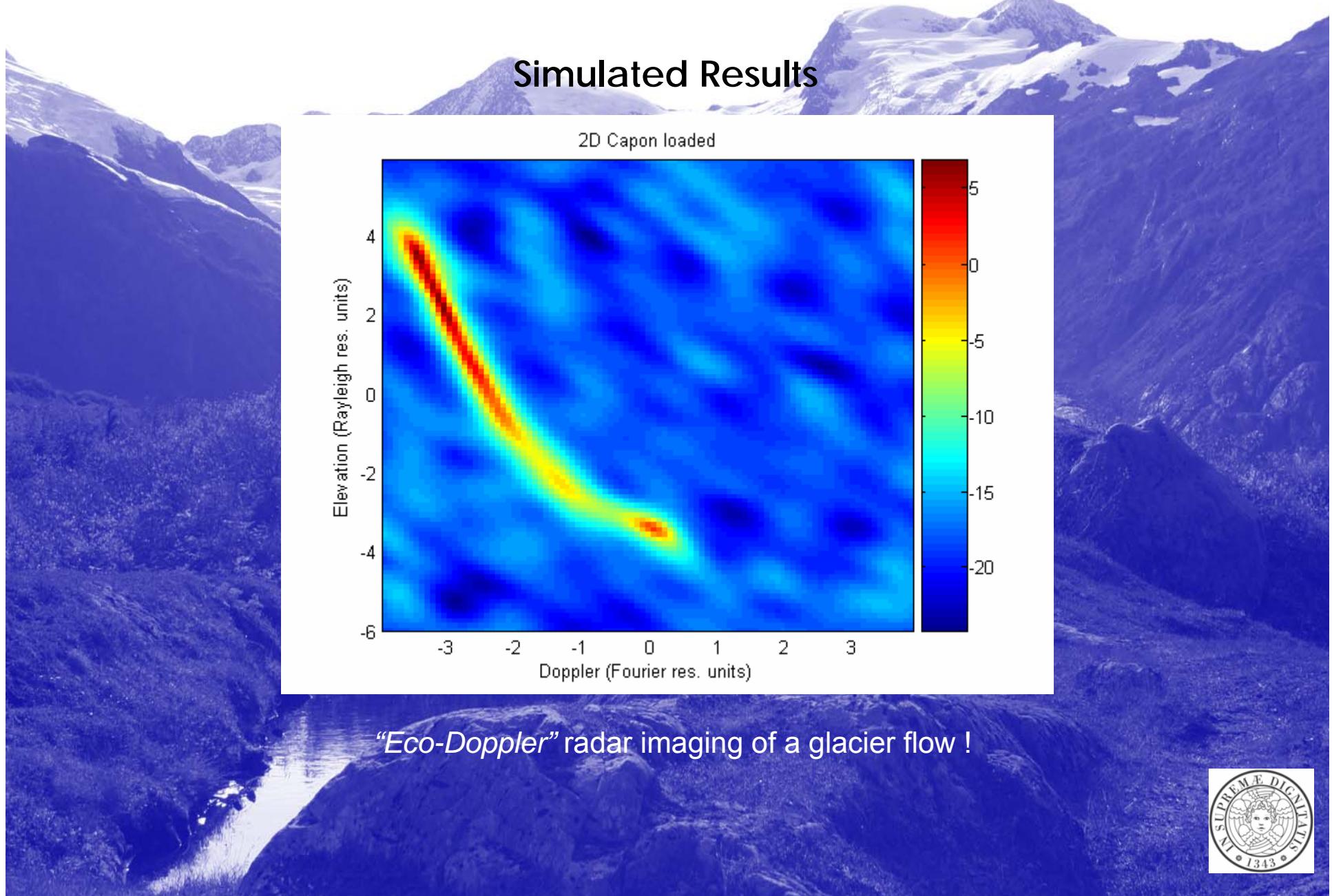
- Single scatterers -



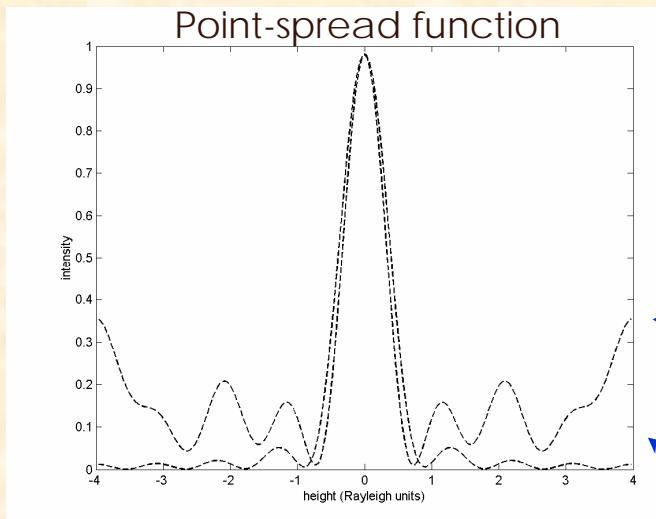
- Double scatterers -



“full 4D Imaging”! (continuous 3D + T)



Interpolation of non-uniformly sampled data

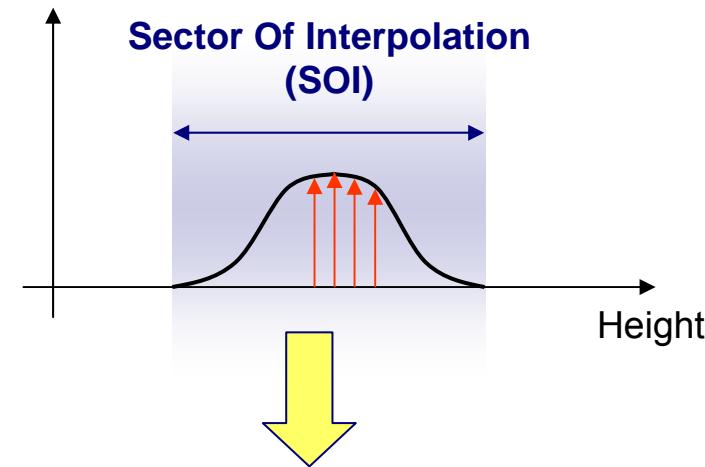


A-priori information:
spectral support
(Sector of Interpolation)

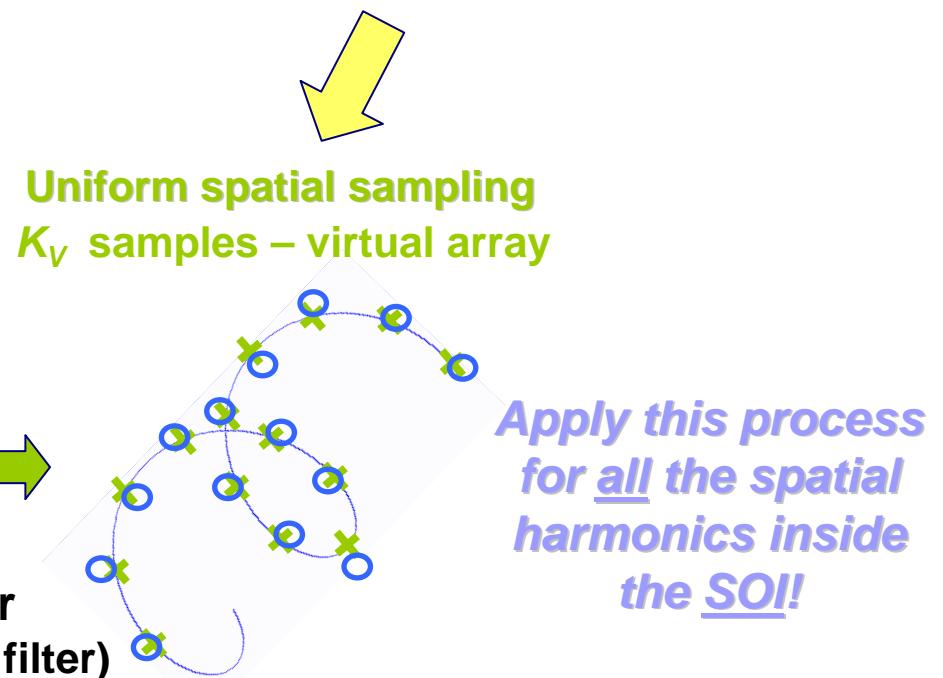
Problema di “predizione”



The new interpolated approach



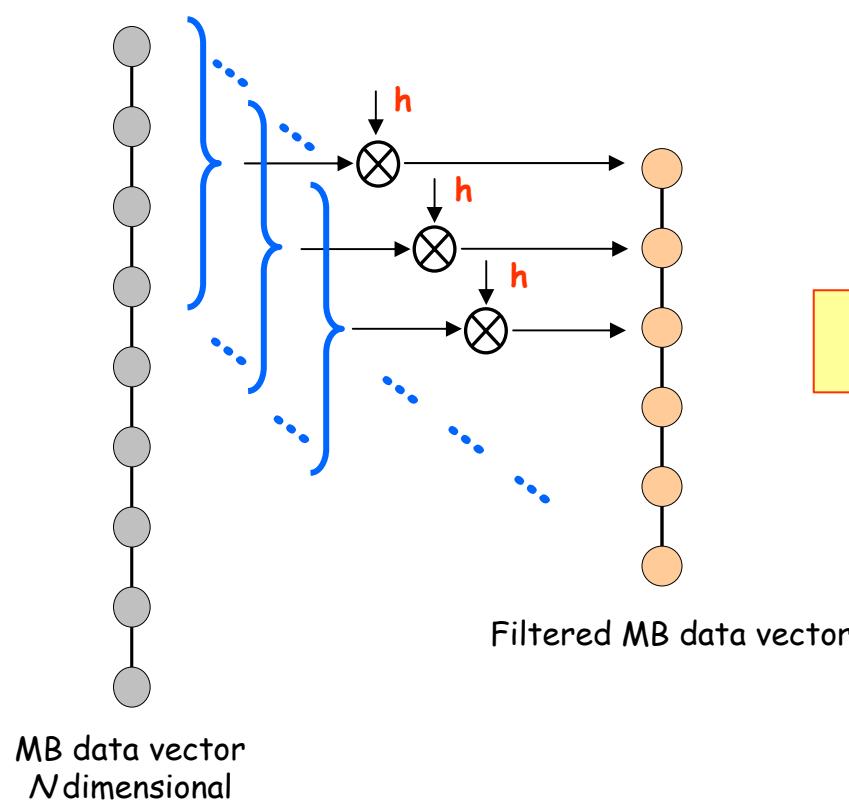
IDEA:
Reconstruct K_V uniform samples
of the spatial harmonic of interest
from the actual non-uniform samples



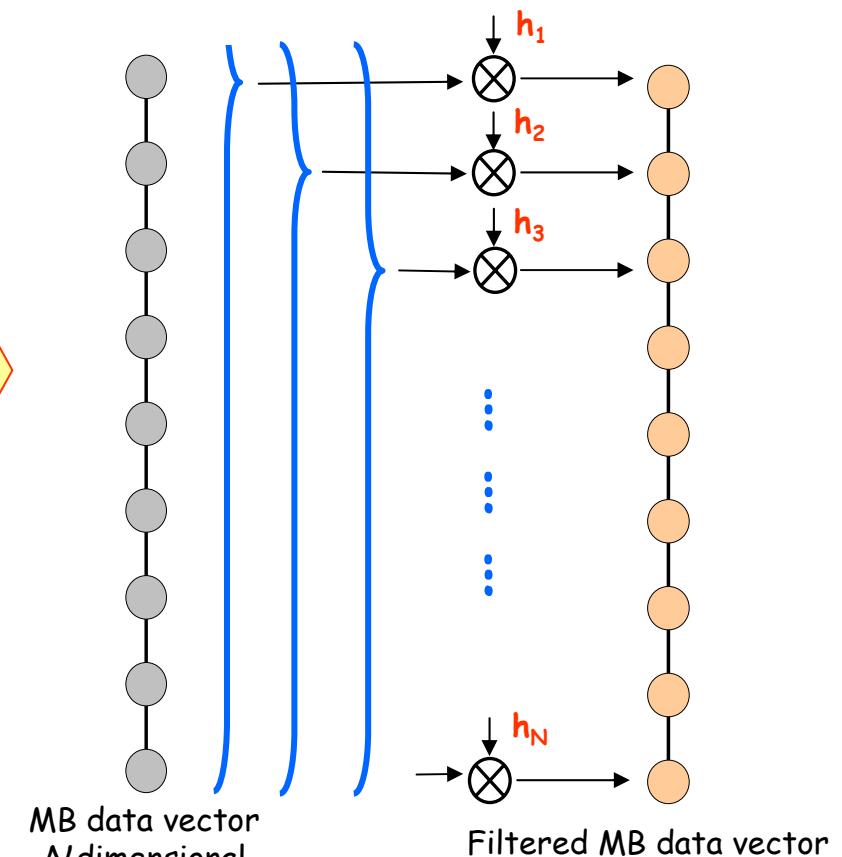
Linear Interpolator
(non-stationary “matrix” filter)

Matrix filter principle

Moving window filter - One coefficient vector h
 Linear, spatially stationary



Matrix filter - Coefficient vectors h_1, \dots, h_N
 Linear, spatially non-stationary




 inner product

- Arbitrary input/output MB spacing allowed

Array interpolation

Deterministic approach – IA

The output vector of the virtual array can be obtained by means of a $K_V \times K$ transformation matrix by solving the least squares (LS) problem:

$$\mathbf{H}_F = \arg \min_{\mathbf{H}} \|\mathbf{A}_V - \mathbf{H}\mathbf{A}\|_F^2$$

where

$$\begin{aligned}\mathbf{A}_V &= [\mathbf{a}_V(\phi_1) \quad \mathbf{a}_V(\phi_2) \quad \cdots \quad \mathbf{a}_V(\phi_s)] \\ \mathbf{A} &= [\mathbf{a}(\phi_1) \quad \mathbf{a}(\phi_2) \quad \cdots \quad \mathbf{a}(\phi_s)]\end{aligned}$$

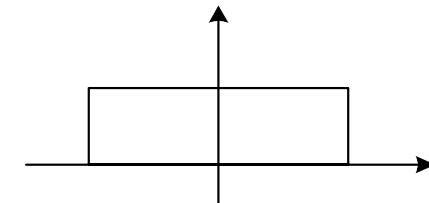
$[\phi_1 \phi_2 \dots \phi_s]$ are s phase samples inside the discretized SOI.

Statistical approach – MSE-IA

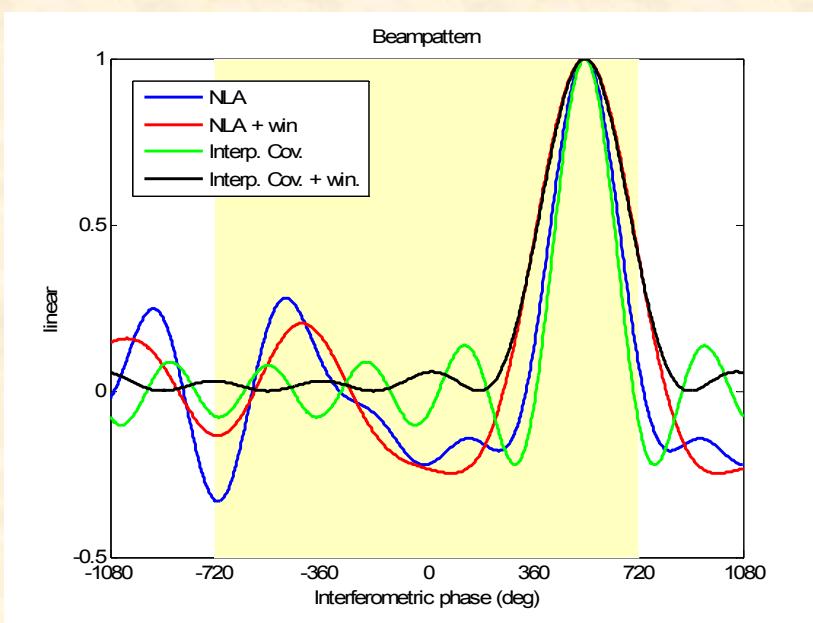
The signal vector χ_{ULA} can be estimated minimizing the mean square error (MSE):

$$\mathbf{H}_M = \arg \min_{\mathbf{H}} E\left\{\left(\chi_{ULA} - \mathbf{H}\mathbf{y}\right)^2\right\}$$

This minimization (*Wiener-Hopf equations*) involves the knowledge of the spatial power spectral density, which can be assumed flat over the SOI in the InSAR application.

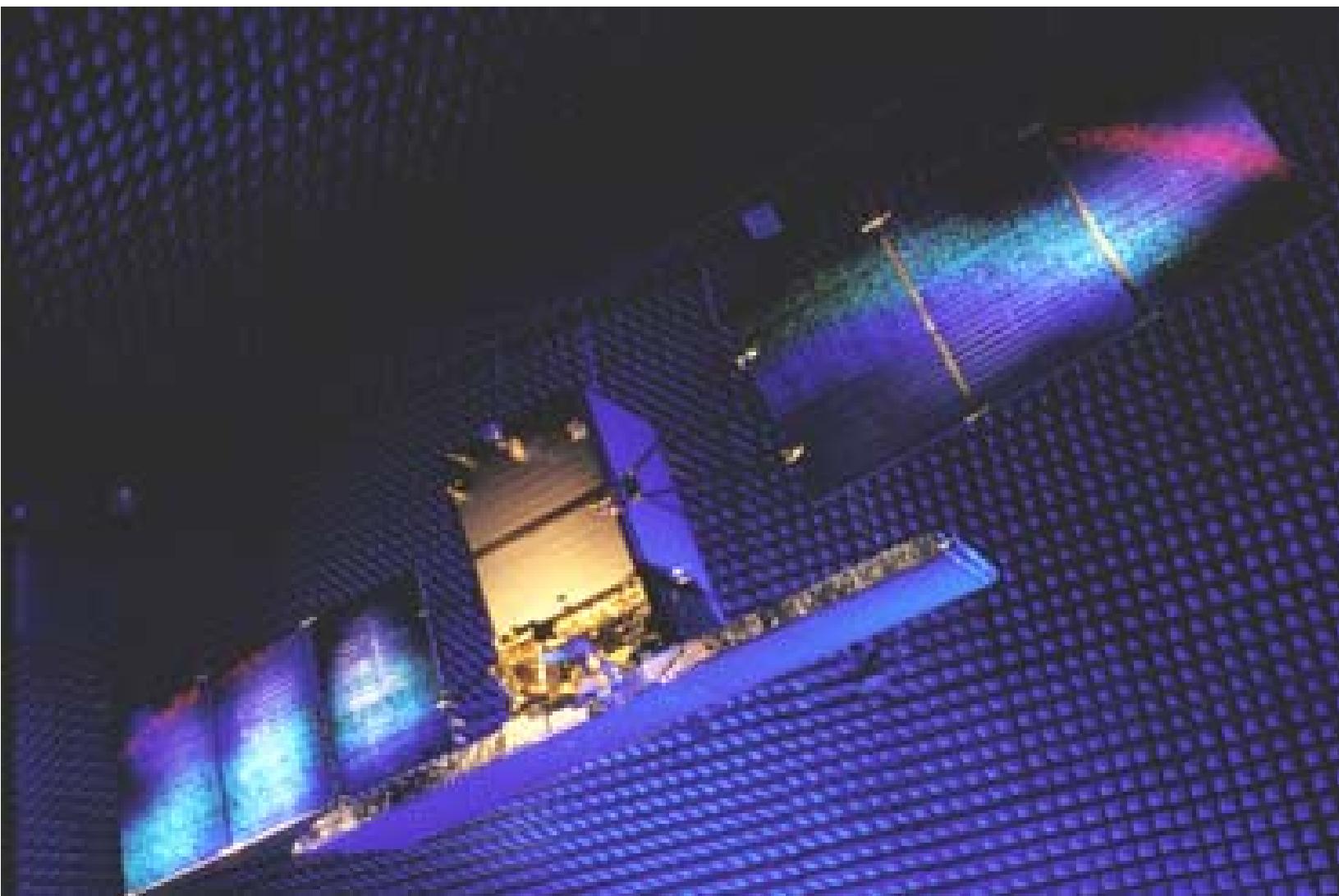


Costa Rica and Amazon Rainforest

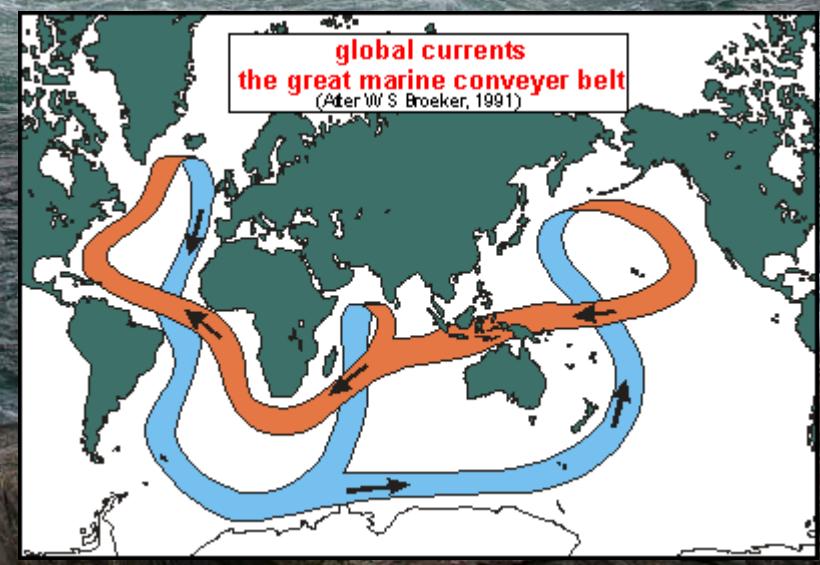


Jet Propulsion Laboratory
California Institute of Technology





Il satellite italiano COSMO-SkyMed



Feasibility Study of Along-track SAR Interferometry with the COSMO-SkyMed Satellite System

Dept. of Ingegneria dell'Informazione, University of Pisa, via Caruso 14 - Pisa, Italy

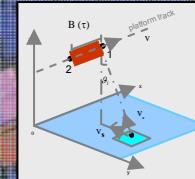


The Italian COSMO-SkyMed system

- 4-satellite constellation
- mapping of Mediterranean latitudes
- Hi-res X-band SAR
- Short revisit time

Along-Track Interferometry

- ocean surface current velocity
- ocean coherence time
- sea waveheight spectrum and MTI



Spaceborne ATI

- NASA SRTM (along-track boom component)
- TerraSAR-X split-antenna
- RADARSAT-II (split-antenna MTI)
- Satellite formations (distributed interferometers)

Performance study of *split-antenna ATI* modes for COSMO-SkyMed SAR:

Statistical data model

multilook data vector (up to 5 rx channels):
 $\mathbf{y}(n) = \sqrt{\sigma^2} \mathbf{A}(\phi) \mathbf{x}(n) + \mathbf{v}(n)$
 Multi Baseline ATI

speckle autocorrelation function
 $[\mathbf{C}_x]_{i,j} = \exp\left\{-\frac{(l-i)\tau}{(K-1)\tau_c}\right\}$

data covariance matrix

$$\mathbf{R} = E\{\mathbf{y}(n)\mathbf{y}^H(n)\} = \sigma^2 \mathbf{A}(\phi) \mathbf{C}_x \mathbf{A}(\phi)^H + \sigma_v^2 \mathbf{I}$$

Cramér-Rao Lower Bound accuracy analysis

unknown and nuisance parameters $\chi = [\phi \ \tau_c \ \sigma^2 \ \sigma_v^2]^T$

$$CRLB(\chi_i) = [\mathbf{J}_{FIM}]_{i,i}$$

$$\text{Fisher information} \quad [\mathbf{J}_{FIM}]_{i,j} = N \operatorname{tr} \left\{ \mathbf{R}^{-1} \frac{\partial \mathbf{R}}{\partial \chi_i} \mathbf{R}^{-1} \frac{\partial \mathbf{R}}{\partial \chi_j} \right\}$$

surface current velocity

$$v_s = \frac{\lambda v}{4\pi B \sin \theta_i} \varphi + v_B \quad \text{Bragg velocity}$$

accuracy of conventional 2-ch. ATI

$$CRLB^{1/2}(v_s) = \frac{\lambda v}{4\pi B \sin \theta_i} \sqrt{\frac{1}{\sqrt{2N}} \frac{1 - (\rho_s \rho_t)^2}{\rho_s \rho_t}}$$

System and physical models

SAR equation for extended targets, split antenna

Moore and Fung ocean scattering model

$$\sigma^0(\phi_{ac}, \theta_i, u_w) = [A + B \cos \phi_{ac} + C \cos 2\phi_{ac}] e^{-(\frac{r_s - r_b}{\eta})/g_0} \left(\frac{u_w}{\bar{u}_w}\right)^r$$

coherence time model (Fraser and Camps)

$$\tau_c = 3 \frac{\lambda}{u_w} \operatorname{erf}^{-1/2} \left(2 \cdot 7 \frac{r_s}{u_w} \right)$$

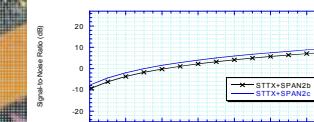
Platform velocity	7548 m/s
Slant range distance	732 Km
Antenna length	5.6 m
Carrier frequency	9.58 GHz
Incidence angle	25°-50°
TX peak power	5 KW
Bandwidth	99.3 MHz

A few split mode configurations:	# rx panels per ch.	# ch.	B [m]
STTX+SPAN2a	1	2	2.24
STTX+SPAN2b	2	2	1.68
STTX+SPAN2c	3	2	1.12
STTX+SPAN3d	3	3	1.12

Investigated trade-offs

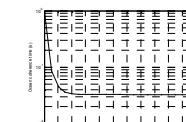
- baseline length
- sub-aperture gain
- channel number
- speckle decorrelation azimuth ambiguity

Signal level



cross-wind, mid-range, $r_s=2.7$ m

Coherence time



up-and-down-wind

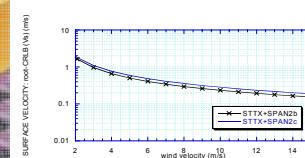
cross-wind

Conv. and MB coherence time estimation and quantization effects also investigated

Rankings

1° STTX+SPAN4/SPAN5
2° STTX+SPAN2b
3° STTX+SPAN2a/SPAN3a
4° STTX+SPAN2c/3d

Velocity accuracy



down-wind, $N=1000$ looks

Resolution

	Accuracy=10 cm/s, mid-range	Accuracy=25 cm/s, mid-range	Accuracy=25 cm/s, far-range
STTX+SPAN2b	$N=189035$ ATI cell = 1271 m	$N=30245$ ATI cell = 508 m	$N=50592$ ATI cell = 657 m
STTX+SPAN2c	$N=235683$ ATI cell = 1419 m	$N=37709$ ATI cell = 568 m	$N=58602$ ATI cell = 708 m



Lancio del satellite italiano COSMO-SkyMed (COSMO-1, Giugno 2007)

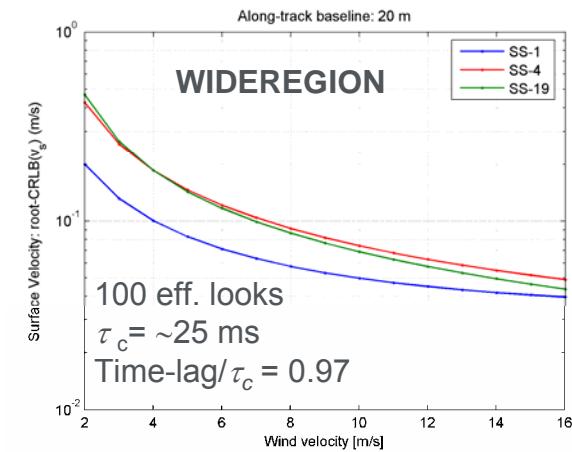
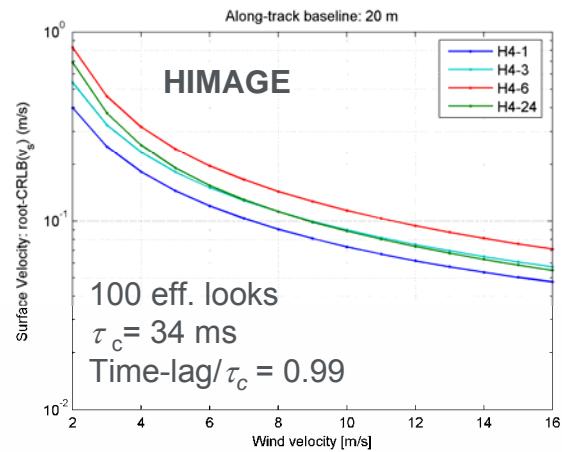
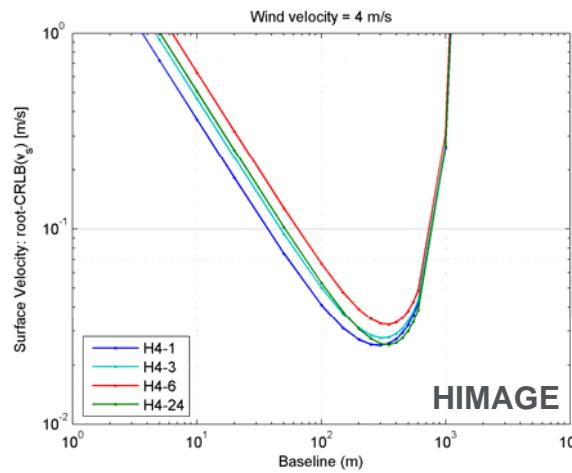
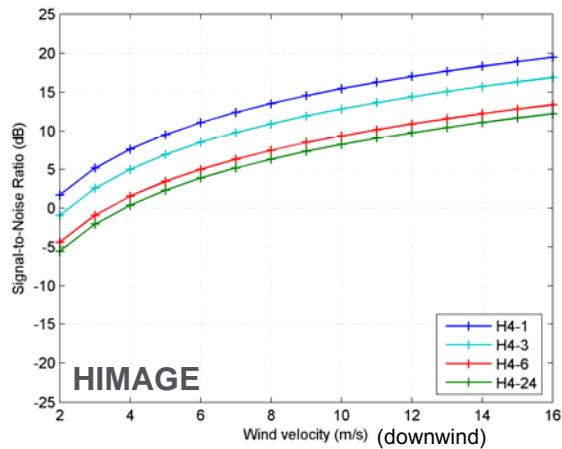
COSMO-SkyMed/BISSAT cluster

ATI Sea Currents and Waves



TELESPAZIO
A Finmeccanica / Thales Company

- Cramér-Rao bound analysis on the expected precision on surface current velocity estimation
- SAR equation for extended target
- Physical models for ocean scattering and coherence time (τ_c)



- A rescaling of the number of looks is evaluated to obtain the desired precision ($\leq 0.1 \text{ m/s}$ at worst)
- @ $B_{ATI}=20 \text{ m}$, $\sim 0.1 \text{ m/s}$ absolute surface velocity accuracy is obtained with $B_{XTI} < 500 \text{ m}$ (availability of precise radar altimeter oceanographic data and DEM of a terrain zone in the image)

Possible extension: BISSAT with COSMO-SkyMed in split-antenna mode

- Dual baseline system
- Small baseline: enlargement of the velocity ambiguous range
- Long baseline: augmented interferometric sensitivity, eased absolute calibration

Example: 200 m total baseline, 3.36 m small baseline (STTX-SPAN2b): higher multilooking degree (50 instead of 15 in the single baseline case) to ensure the desired precision of 0.1 m/s

System model perturbations

Residual multibaseline array calibration or atmospheric compensation errors affect the data in all the practical applications

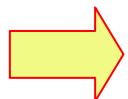
Phase error vector introduced by the residual miscalibration at each array phase center:

$$\mathbf{e} = [e^{j4\pi d_1} \quad e^{j4\pi d_2} \quad \dots \quad e^{j4\pi d_K}]^T$$

$d_1 \dots d_K$

Gaussian, zero mean, uncorrelated random variables, with variance σ^2

- **Positioning errors** of the array elements (normalized to wavelength), due to uncertainties in the platform positioning in the direction parallel to the line of sight (errors in the perpendicular direction can be neglected)
- **Residual one-way path delay** variations of wave propagation through the atmosphere (normalized to wavelength)



Total array response: $\mathbf{a}(h) = \mathbf{a}_P(h) \odot \mathbf{e}$

Element-by-element product



Hybrid Cramér-Rao Bound

Signal model:

$K = \# \text{ of array elements}$
 $N_S = \# \text{ of sources}$

$$\mathbf{y}(n) \approx \sum_{i=1}^{N_S} \sqrt{\tau_i} \mathbf{a}_P(\varphi_i) \odot \mathbf{e} \odot \mathbf{x}_i(n) + \mathbf{v}(n) \quad n = 1, 2, \dots, N$$

↑ *Steering vector* ↑ *Calibration error vector* ↑ *Speckle vector (multiplicative noise)*

Cramér-Rao bound

$$\xi = [\varphi_1 \ \cdots \ \varphi_{N_S} \ \tau_1 \ \cdots \ \tau_{N_S} \ b_1 \ \cdots \ b_{N_S} \ \sigma_v^2]^T \quad \text{Parameter vector}$$

$$[\mathbf{J}_{FIM}]_{m,n} = -\mathbb{E}_{\mathbf{y}} \left\{ \frac{\partial^2 \ln \left(\int \int \dots \int f(\mathbf{y}, \mathbf{e}; \xi) d\mathbf{d}_1 d\mathbf{d}_2 \dots d\mathbf{d}_K \right)}{\partial \xi_m \partial \xi_n} \right\} \rightarrow \text{Unfeasible !}$$

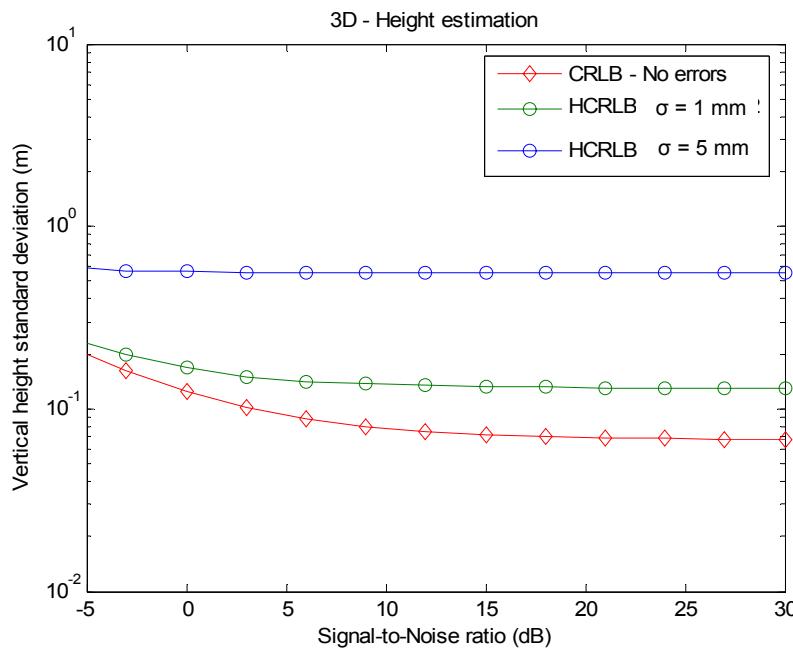
Hybrid Cramér-Rao bound

Good trade-off between easiness of calculation and tightness

$$\xi = [\varphi_1 \ \cdots \ \varphi_{N_S} \ \tau_1 \ \cdots \ \tau_{N_S} \ b_1 \ \cdots \ b_{N_S} \ \sigma_v^2 \ d_1 \ \cdots \ d_K]^T \quad \text{Nuisance parameters}$$

$$[\mathbf{J}_{H-FIM}]_{m,n} = -\mathbb{E}_{\mathbf{y}, \mathbf{d}} \left\{ \frac{\partial^2 \ln f(\mathbf{y}, \mathbf{e}; \xi)}{\partial \xi_m \partial \xi_n} \right\} = -\mathbb{E}_{\mathbf{d}} \left\{ \mathbb{E}_{\mathbf{y}|\mathbf{d}} \left\{ \frac{\partial^2 \ln f(\mathbf{y}, \mathbf{e}; \xi)}{\partial \xi_m \partial \xi_n} \right\} \right\} \rightarrow HCRB(\xi_k) = [\mathbf{J}_{H-FIM}]_{k,k}$$

System model perturbations: Hybrid CRLB



3D: height

Double scatterer distance in height: 2 r.u.

Notes:

- σ is standard deviation of residual atmospheric one-way path delay variations
- Variations uncorrelated from track to track
- Compensated I.o.s. velocity



Robust interpolator derivation

Again: $\mathbf{y}_V(n) = \mathbf{H}_I \mathbf{y}(n)$

Robust interpolator, $K_V \times K$ matrix

In the perfectly calibrated case: $\mathbf{H}_I \mathbf{A}_P = \mathbf{A}_{P,V}$ *Overdetermined equation system:
LS solution*

NLA array steering matrix:
The columns are the NLA st.
vectors calculated for a number of
spatial frequency in the SOI

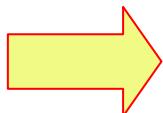
Virtual array steering matrix:
The columns are the virtual st.
vectors calculated for the same
spatial frequencies in the SOI

In the presence of calibration errors:

$$\mathbf{H}_I (\mathbf{A}_P + \Delta \mathbf{A}_P) = \mathbf{A}_{P,V} + \Delta \mathbf{A}_{P,V}$$

Calibration error

Interpolation error

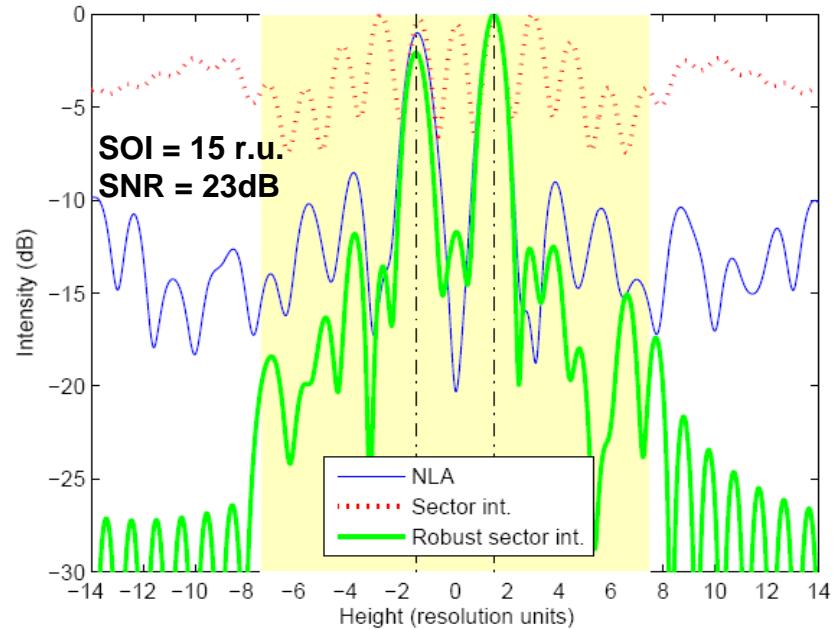
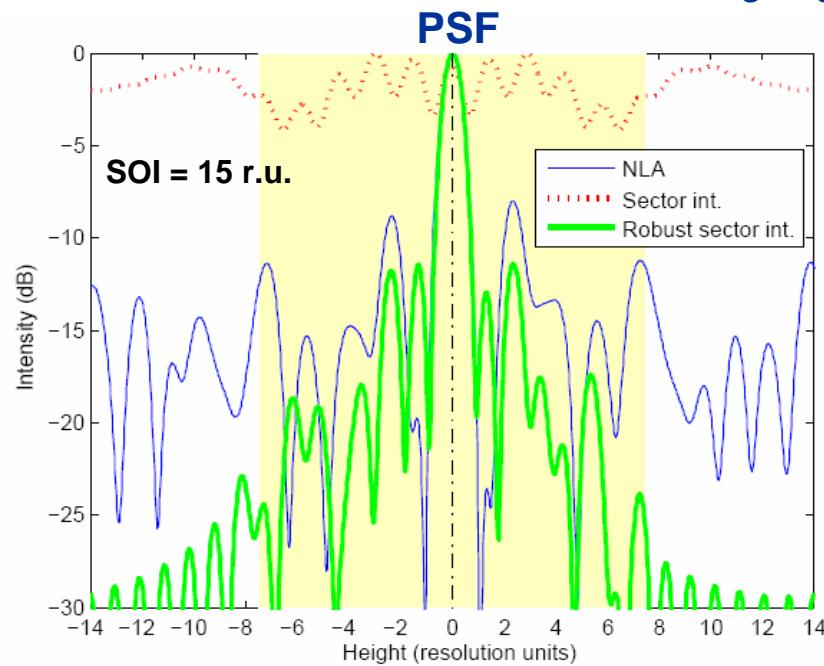


$$\boxed{\mathbf{H}_I = \arg \min_{\mathbf{H}} E \left\{ \|\Delta \mathbf{A}_{P,V}\|_F^2 \right\}}$$

*LS solution to a
perturbed equation system*

Results of the robust processing

$K = 24$ highly non uniform, $K_V = 29$
 $\sigma = 0.01$ (λ units)



- The basic sector interpolator degrades in the reconstruction of the tomographic profile
- The higher effectiveness of the robust method is apparent

