

# **SCAT wind retrieval simulator & FoM**

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# Outline

- Objective
- SCAT Baseline and backup concepts
- End-to-End Performance Study
  - Input winds
  - Observation geometry
  - Instrumental and geophysical noise
  - Wind retrieval simulator
  - Figure of Merit
- Test cases
- Conclusions

# Objective

- KNMI is responsible for the End-to-End SCAT Performance Study for Post-EPS (2019):
  - *To assess the wind retrieval performance*
  - *To support parameter optimization*

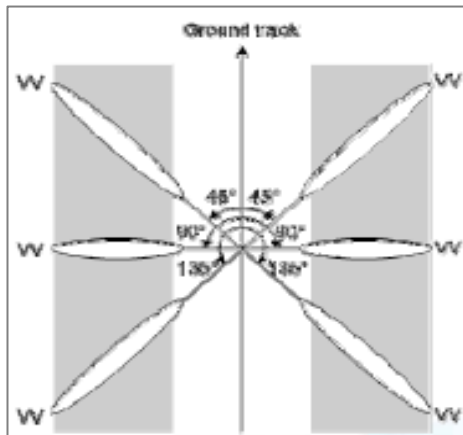
# Post EPS scatterometer (SCA)

## [baseline requirements and options]

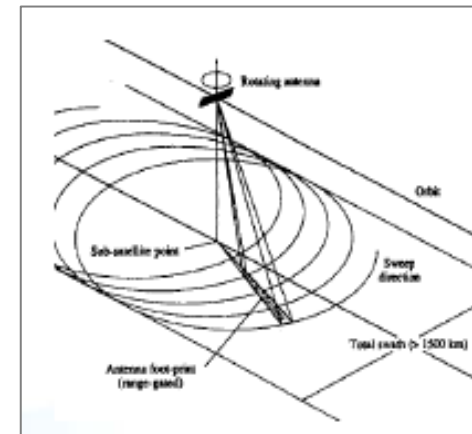
- Spatial resolution (25 km)
- Dynamic range (4-25 m/s)
- Radiometric resolution (~3-10% at 4 m/s)
- Swath coverage (95% in 48 hours for incidences between 20 and 60°)

MetOp orbit → Sun Sync  
with 820 km altitude

### I - Fixed beam (ASCAT type)

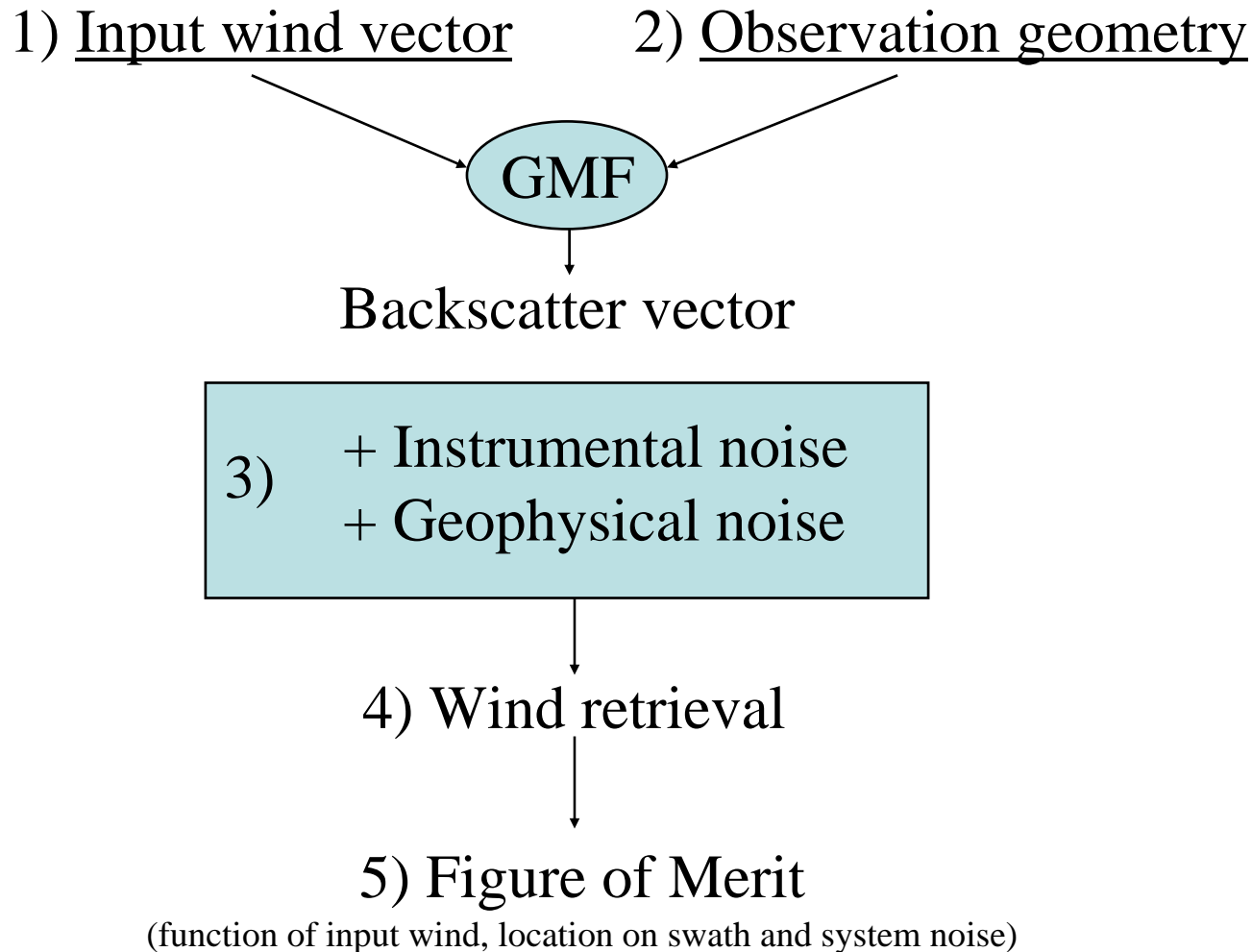


### II - Rotating beam (RFSCAT type)

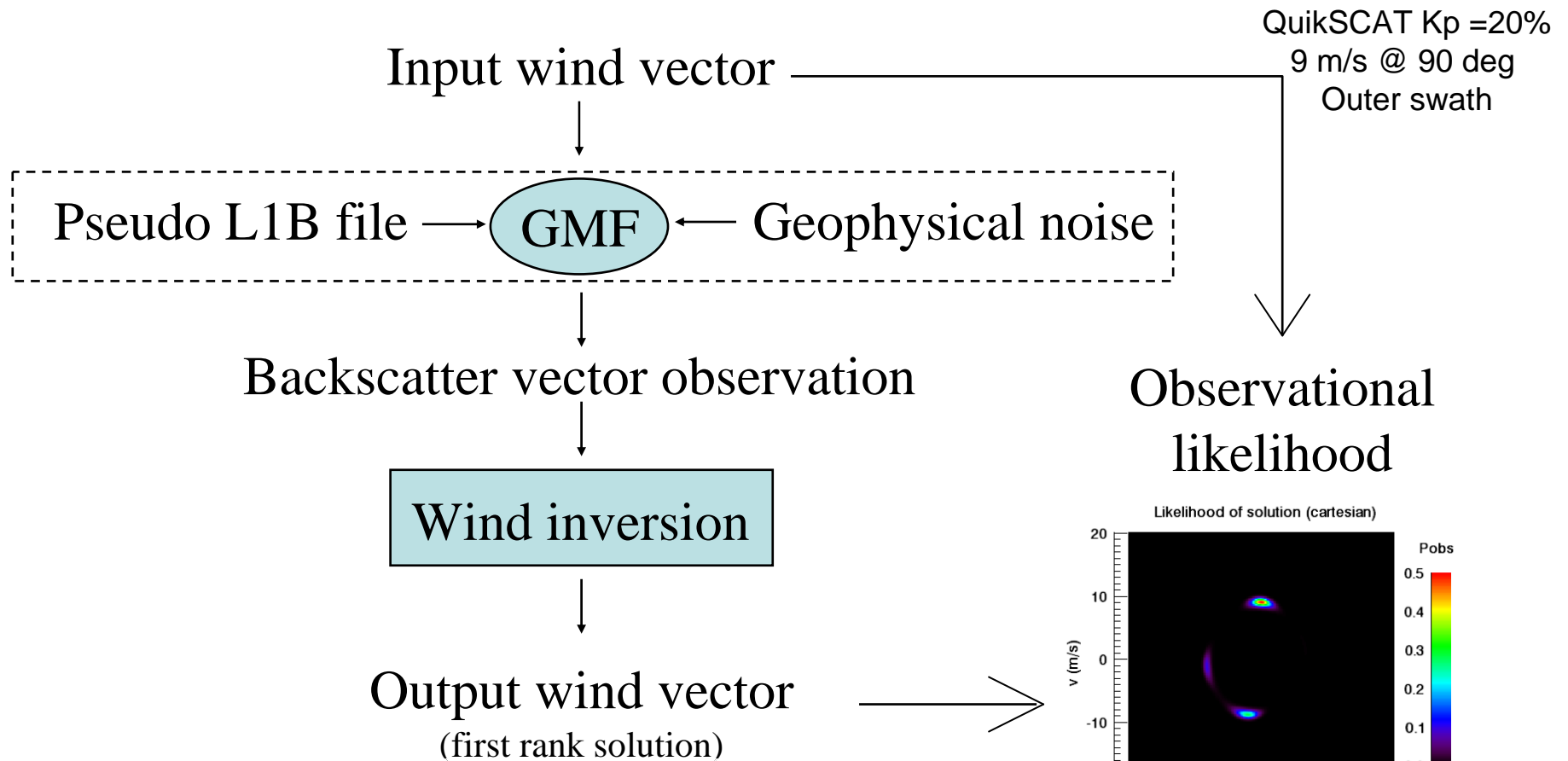


Discarded: Ku-band, pencil beam, extended nadir coverage for ASCAT type

# End-to-End Performance Study

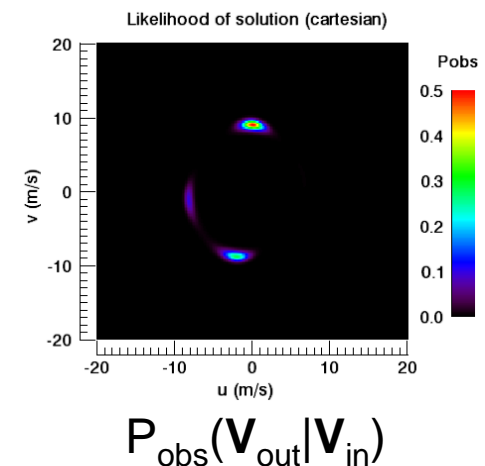


# SCA wind retrieval simulator (F90)



$$MLE(w, \phi) = \frac{1}{\langle MLE \rangle_{i=1, \dots, N}} \sum \frac{|\sigma_i^0 - \sigma_{GMF, j}^0(w, \phi)|^2}{K_p^2 (\sigma_i^0)^2}$$

CFOSAT, Oct 2009



# Specify complete SCA arrangement:

## 1) Antenna configuration

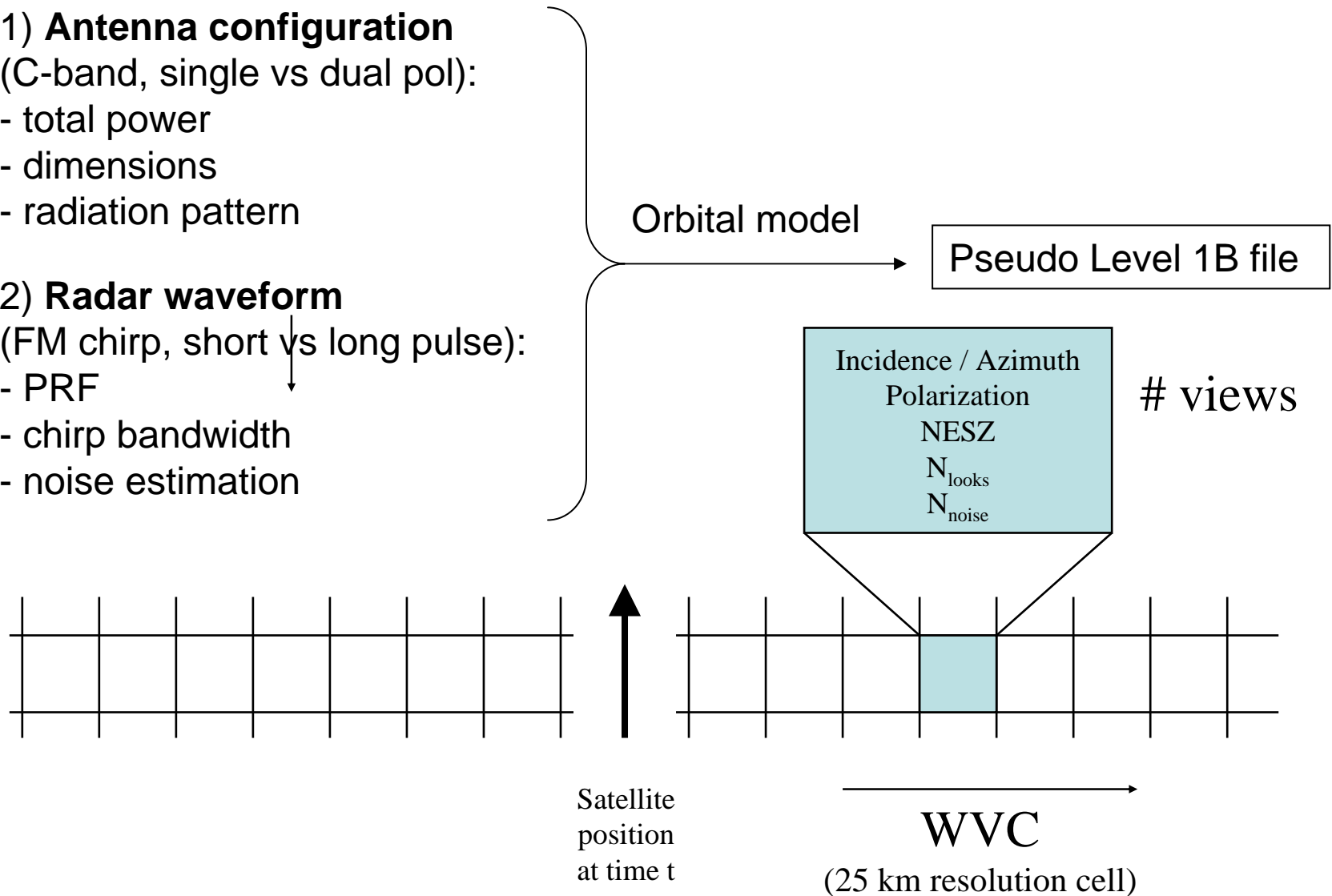
(C-band, single vs dual pol):

- total power
- dimensions
- radiation pattern

## 2) Radar waveform

(FM chirp, short vs long pulse):

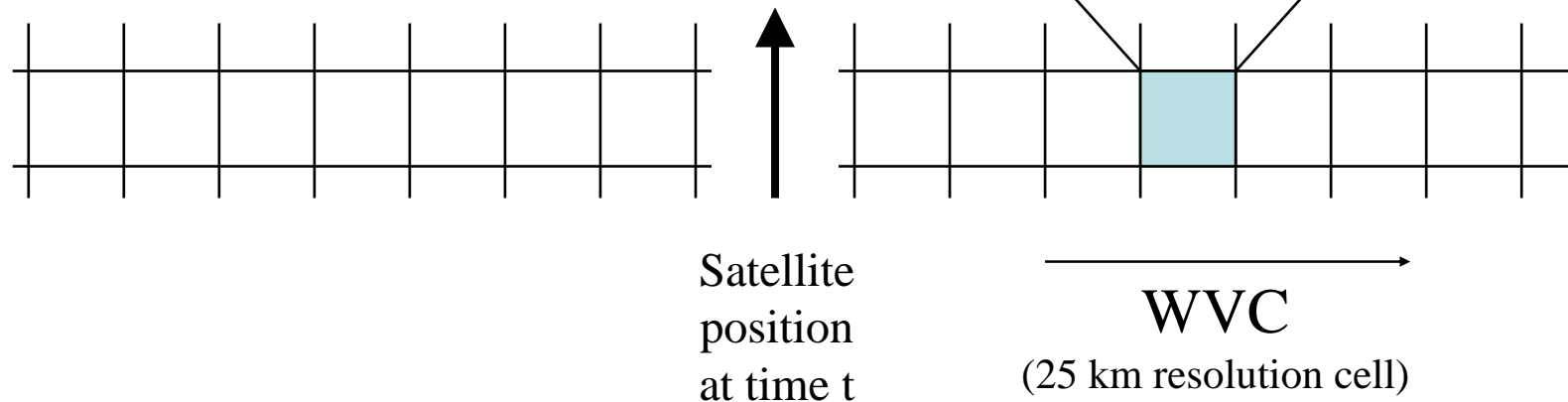
- PRF
- chirp bandwidth
- noise estimation



## 2) Observation geometry

- Orbital model
- Antenna/Pulse design
- Spatial ground filter

↓  
Pseudo L1B file  
(simulated swath coverage)





# Radiometric resolution (NESZ and Kp)

1) NESZ (Noise Equivalent Sigma Zero) for a single look:

$$NESZ = \frac{\sigma^0}{SNR} = \frac{k_B (T_0 + T_{eq})}{\frac{\lambda^2}{(4\pi)^3} \left( \frac{P_t G_{TX} G_{RX}}{R^4 \cdot L_{prop}} \right)} \frac{B_{look}}{A_{look}} \quad A_{look} = \Delta_{range} \Delta_{azimuth}$$

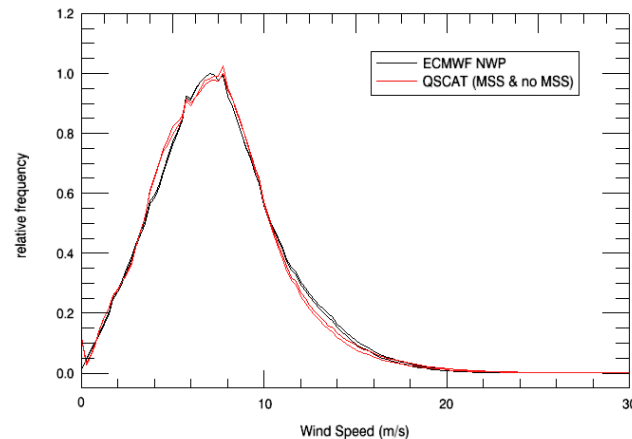
2) Number of looks per node:  $N_{looks} = \frac{\Delta x \Delta y}{A_{look}}$  (reduce speckle)

3) Number of noise samples:  $N_{noise} = f_s T_{noise}$  (noise estimation)

Radiometric resolution:  $K_p^2 = \frac{\text{var}\{\sigma^0\}}{\langle \sigma^0 \rangle^2} = \frac{1}{N_{looks}} \left( 1 + \frac{1}{SNR} \right)^2 + \frac{1}{N_{noise}} \left( \frac{1}{SNR} \right)^2$

# 1) Input wind vector

- Climatology distribution of wind inputs:



Weibull distribution in wind speeds (with maximum around 8 m/s) and uniform distribution in wind direction.

### 3) Instrumental & geophysical noise

- Instrumental (radiometric) noise

$$k_p^2 = \frac{\text{var}\{\sigma^0\}}{(\sigma^0)^2} = \frac{1}{N_{looks}} \left(1 + \frac{1}{SNR}\right)^2 + \frac{1}{N_{noise} SNR^2}$$

- Geophysical noise

$$k_g^2 = \frac{\text{var}\{\sigma^0\}}{(\sigma^0)^2} = 0.064(|\vec{v}| - 16)^2$$

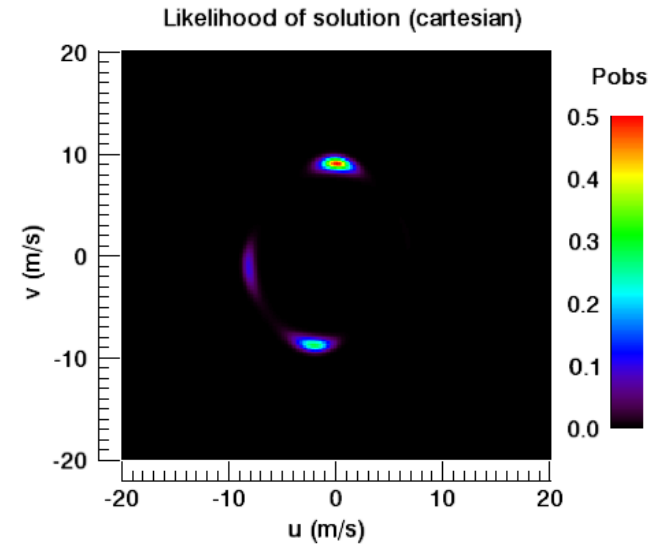
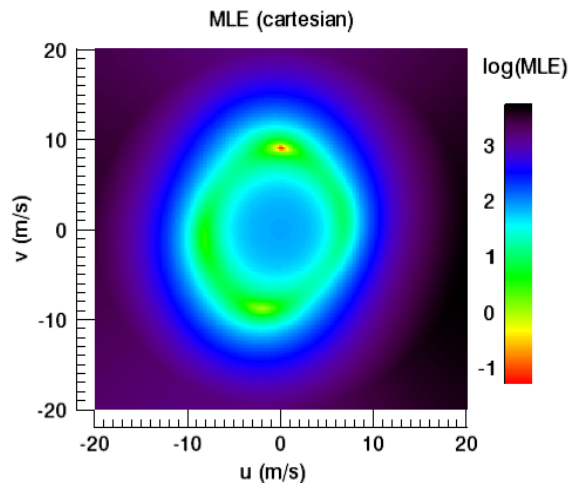
Simulated observations

→

$$\sigma^0 = \sigma_{GMF}^0 (1 + \sqrt{k_p^2 + k_g^2} \cdot N[0;1])$$

# 4) Wind retrieval – MLE and likelihood

$$MLE(w, \phi) = \frac{1}{\langle MLE \rangle} \sum_{i=1, \dots, N} \frac{|\sigma_i^0 - \sigma_{GMF,i}^0(w, \phi)|^2}{\text{var}\{\sigma_i^0\}}$$

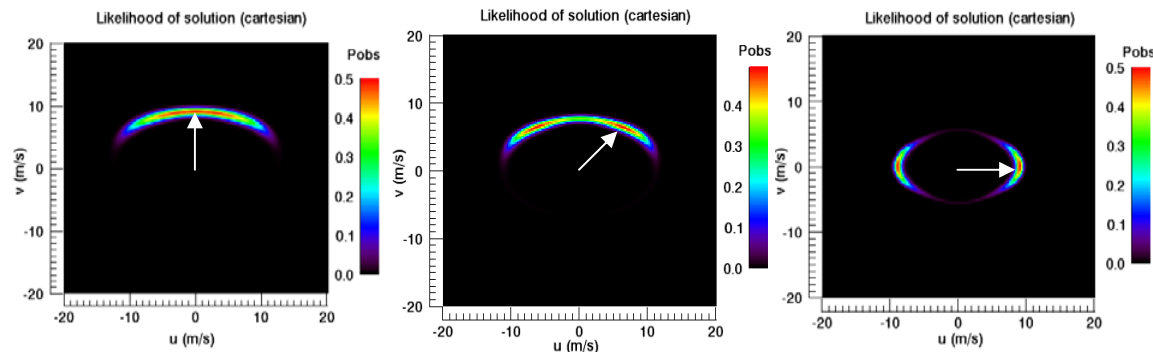


$$P_{obs}(\sigma^0; u, v) = \chi_{N-2}^2 [MLE(u, v)]$$

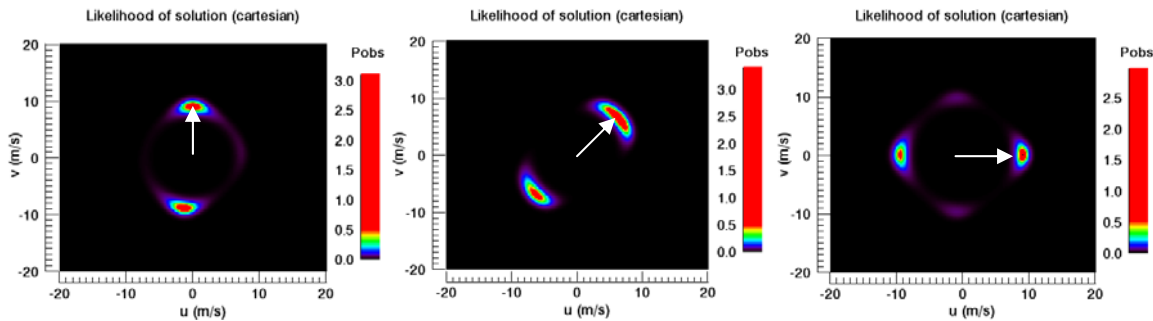
- Multiple ambiguous solutions and broad maxima are limiting factors in wind retrieval performance.

# 4) Wind retrieval – worst cases

- Rotating beam (SeaWinds WVC=0)

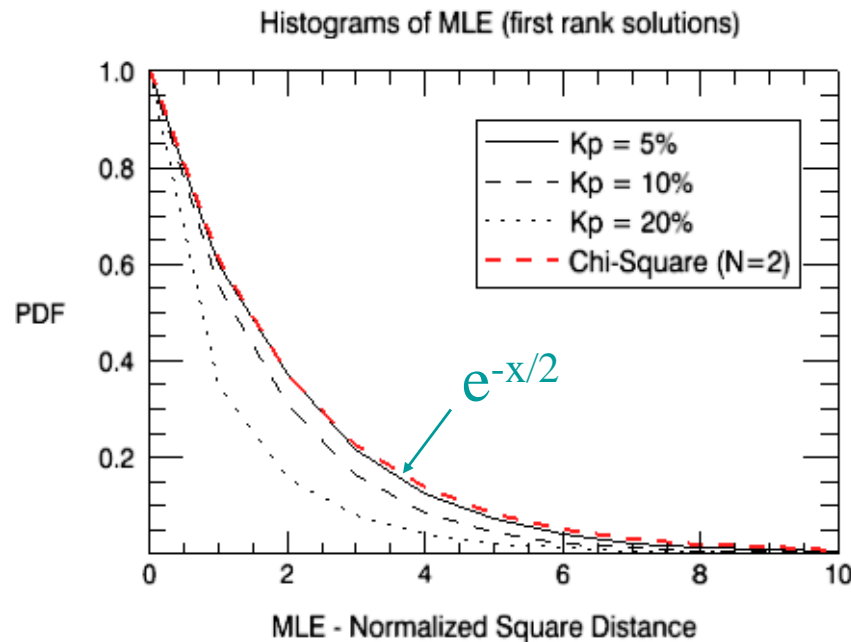


- Fixed antennas (ASCAT WVC = 0)



# SCAT simulator verification tests

## 1) Verifying MLE statistics (“chi-square assumption”)

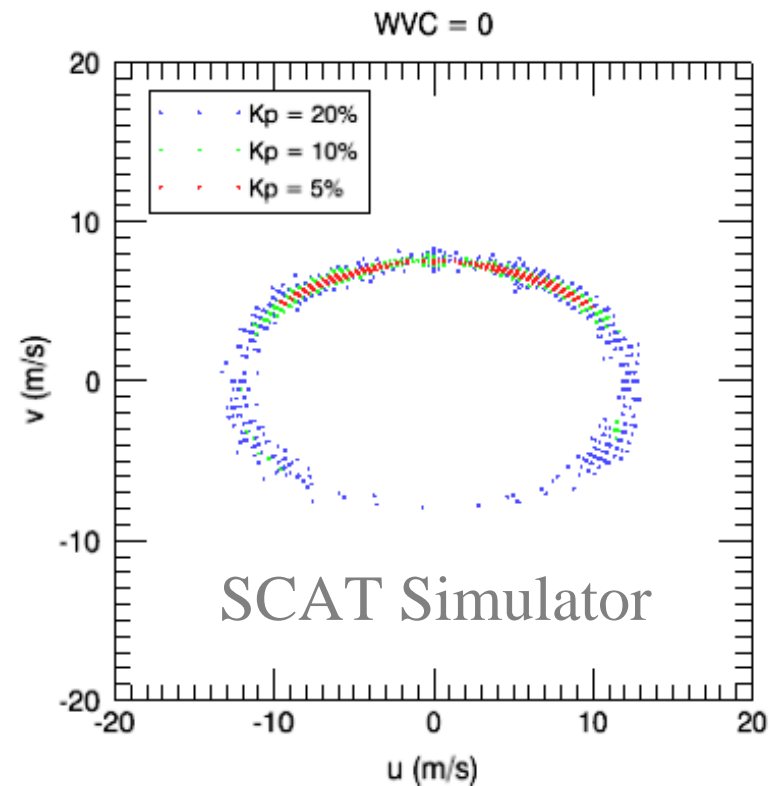
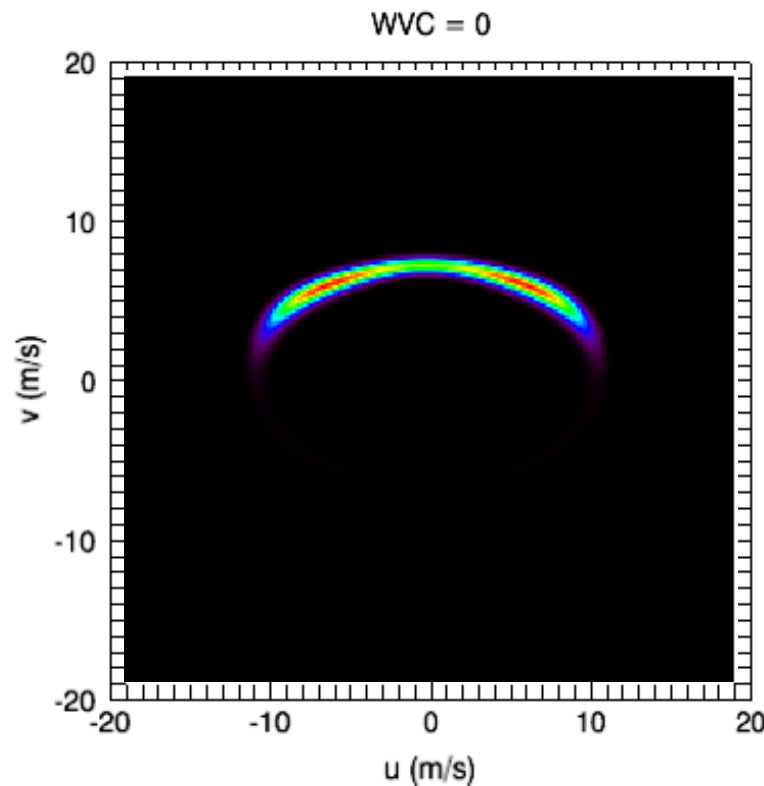


- Simulator MLE statistics follow  $\chi^2$  assumption at low noise levels.
- At higher noise levels GMF loses its 2D surface quality and statistics are modified accordingly!

QSCAT (N = 4 views)

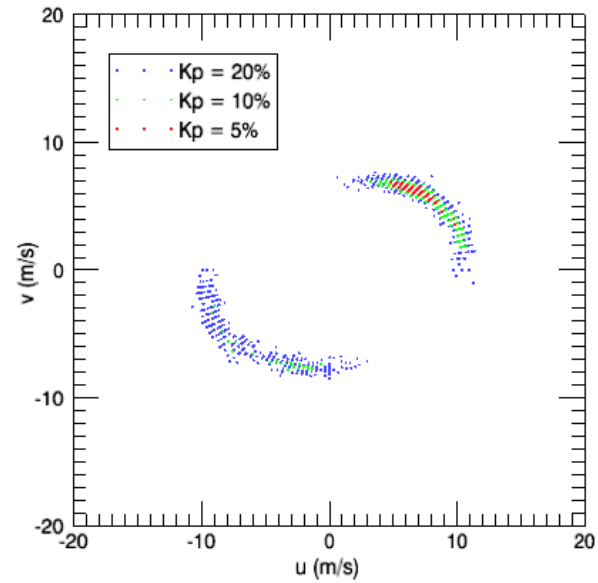
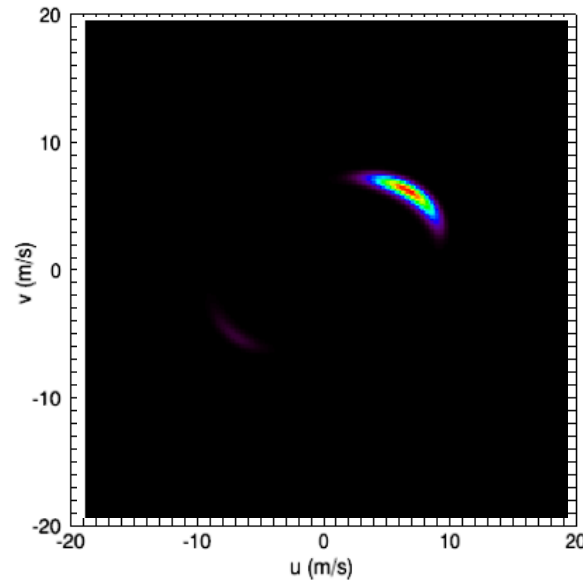
# SCAT simulator verification tests

## 2) Verifying observational likelihoods

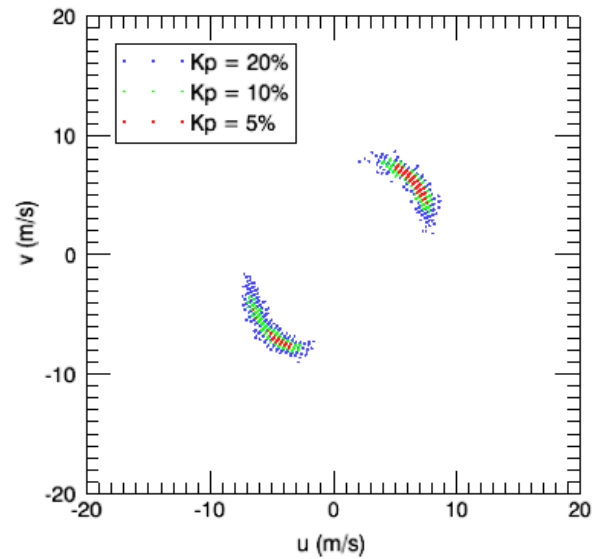
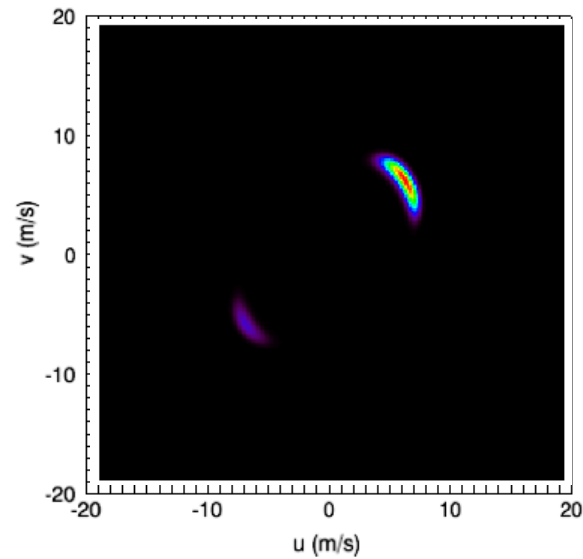


(True wind input is 9 m/s @ 45 deg, 1000 simulator runs)

QSCAT  
WVC = 13



QSCAT  
WVC = 26



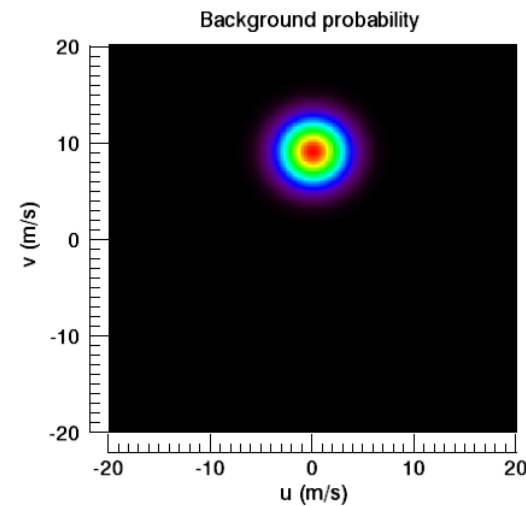
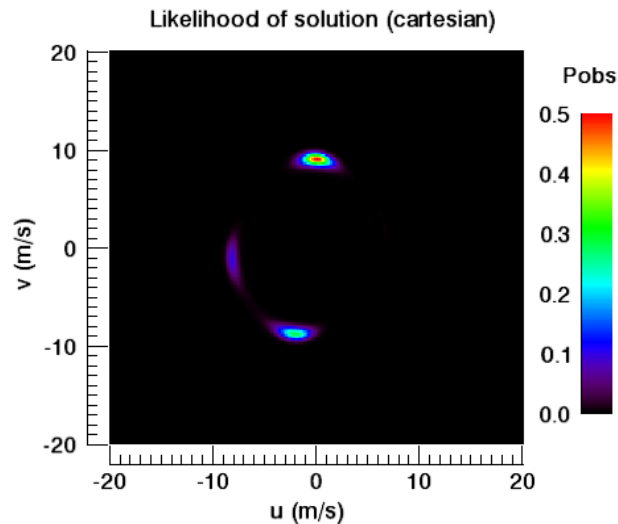
- SCAT Simulator is more realistic “than expected”.  
CFOSAT, Oct 2009



# 4) Wind retrieval – background info

- Combined observation and NWP background cost functions:

$$J = -2\ln(\text{Probability}) = J_{obs} + J_{bg} = -2\ln\left(\chi_{N-2}^2 [MLE(\vec{v})]\right) - 2\ln\left(N\left[\vec{v} - \vec{v}_{bg}; \sigma_{bg}\right]\right)$$



$$p_{obs}(u, v) = C \chi_{N-2}^2 [MLE(u, v)]$$

$$p_{bg}(u, v) = N\left[\vec{v} - \vec{v}_{bg}; \sigma_{bg}\right]$$

Best wind estimate  $\longrightarrow$  Maximum likelihood

# 5) Figure of Merit - definition

- Enable the comparison of different SCAT configurations

1) Vector RMS error:  $FoM = \frac{RMS_{obs}}{RMS_{bg}} \subset [0, 1]$

$$RMS_{obs} = \left( \int |\vec{v} - \vec{v}_{true}|^2 p_{obs}(u, v) p_{bg}(u, v) dudv \right)^{1/2}$$

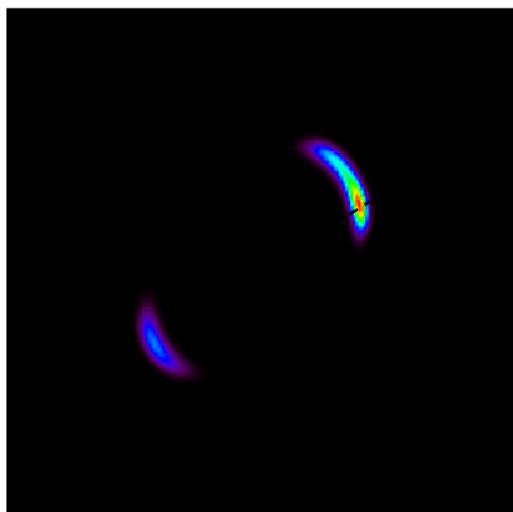
$$RMS_{bg} = \left( \int |\vec{v} - \vec{v}_{true}|^2 p_{bg}(u, v) dudv \right)^{1/2} = \sqrt{2} \sigma_{bg}$$

2) Ambiguities:

$$FoM_{amb} = \frac{\int p_{obs}(u, v)(1 - p_{bg}(u, v)) dudv}{\int p_{obs}(u, v) p_{bg}(u, v) dudv} \subset [0, \infty]$$

# Systematic biases

Determined by skewness in observational likelihood:



skewness = mode - mean  
(in wind speed and direction)

$$skew_{speed} = v_{true} - \int v \cdot p_{obs}(v, \phi_{true}) p_{bg}(v, \phi_{true}) dv$$

$$skew_{dir} = \phi_{true} - \int \phi \cdot p_{obs}(v_{true}, \phi) p_{bg}(v_{true}, \phi) d\phi$$

- Example: True wind input with 9 m/s @ 30 deg features small wind speed bias but output wind direction seems drawn to 45 deg solution!

# Wind retrieval performance

$P_{obs}(\mathbf{V}_{out}|\mathbf{V}_{in})$

NWP background ( $5 \text{ m}^2/\text{s}^2$ )

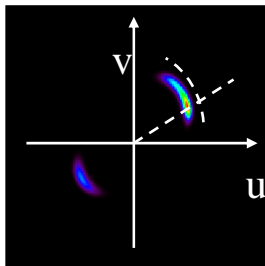
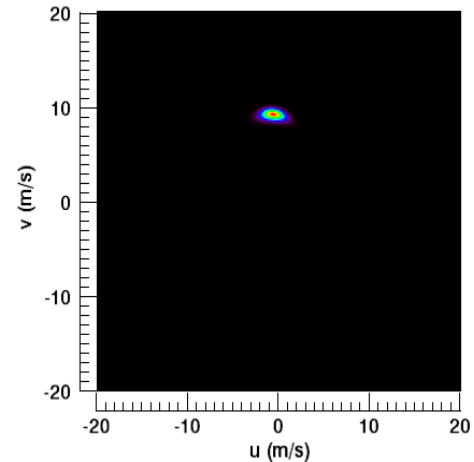
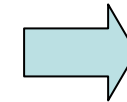
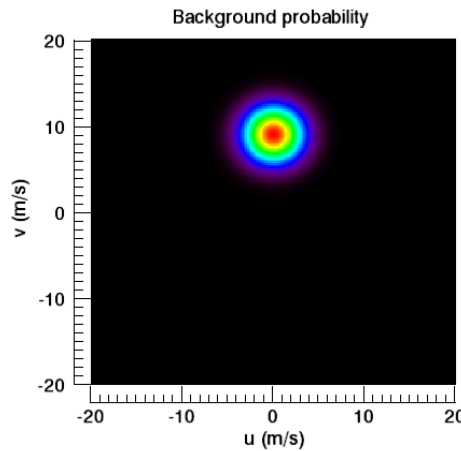
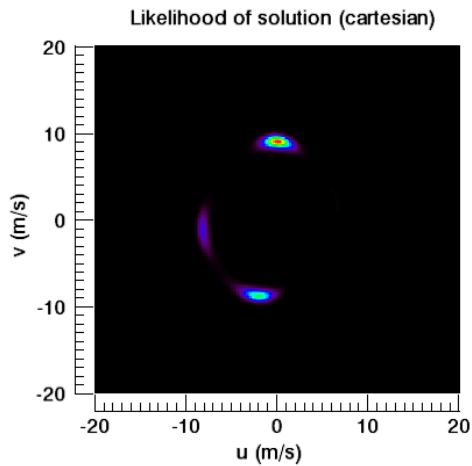


Figure of Merit

- 1) Wind Vector RMS error
- 2) Ambiguity susceptibility
- 3) Wind biases (skewness)

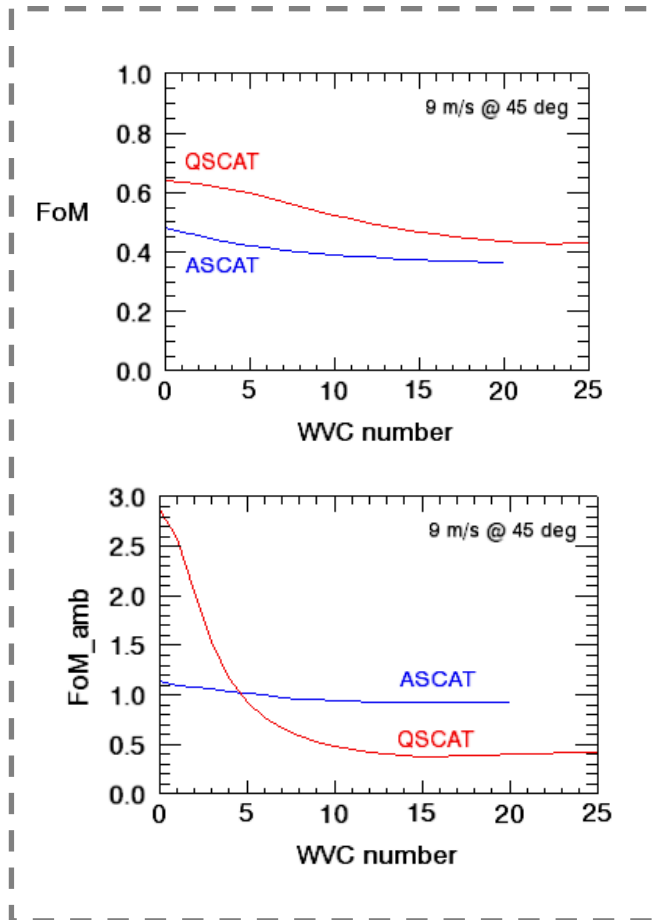
For instance:

$$RMS_{obs} = \left( \int |\vec{v} - \vec{v}_{true}|^2 p_{obs}(\vec{v}) p_{bg}(\vec{v}) d^2v \right)^{1/2}$$

CFO SAT, Oct 2009

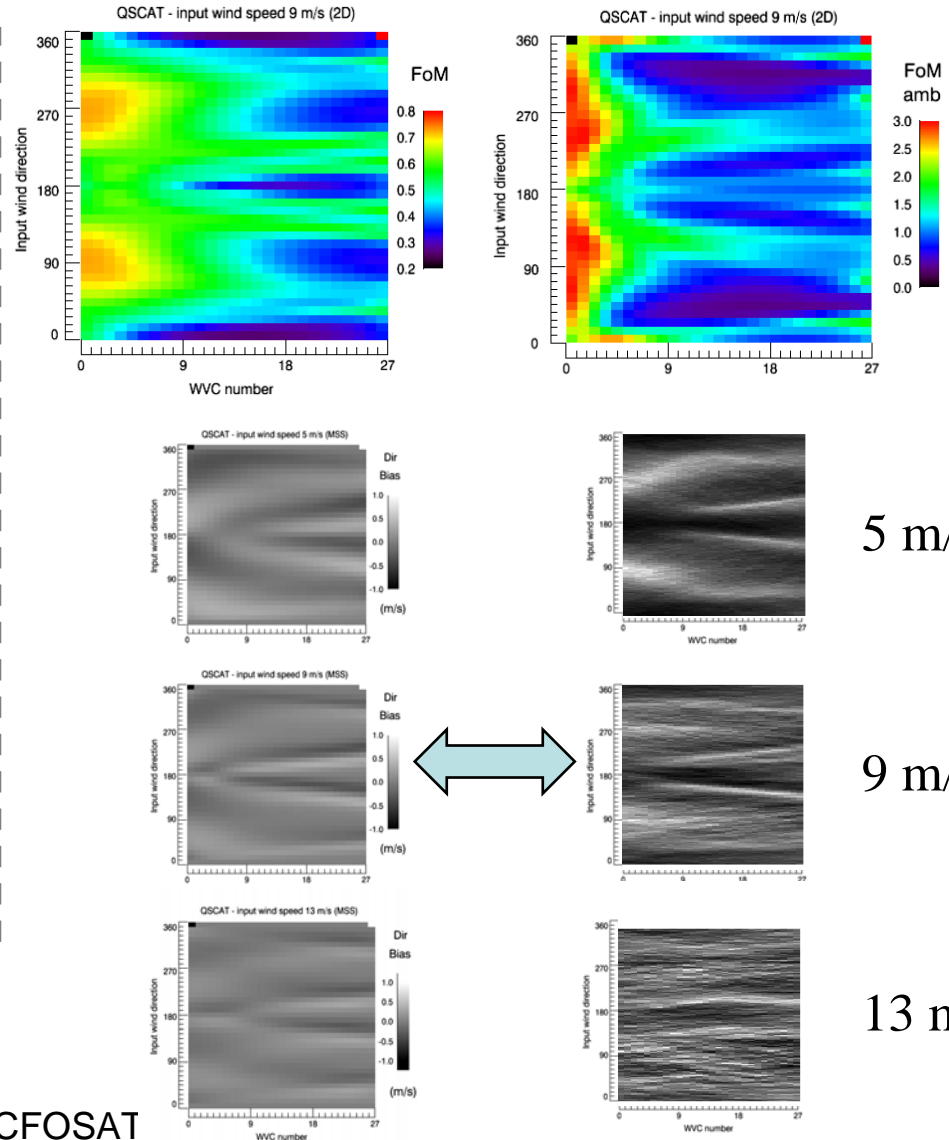
# Wind retrieval performance

9 m/s

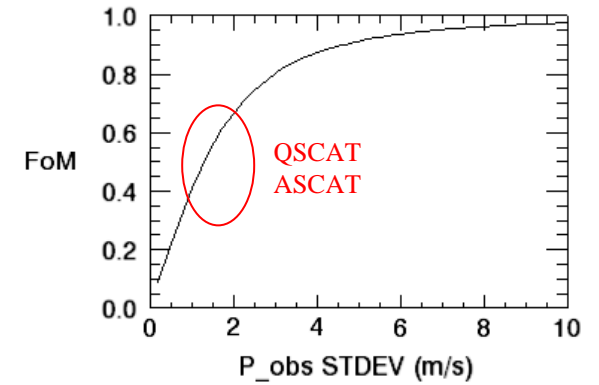


examples

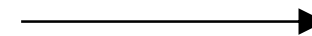
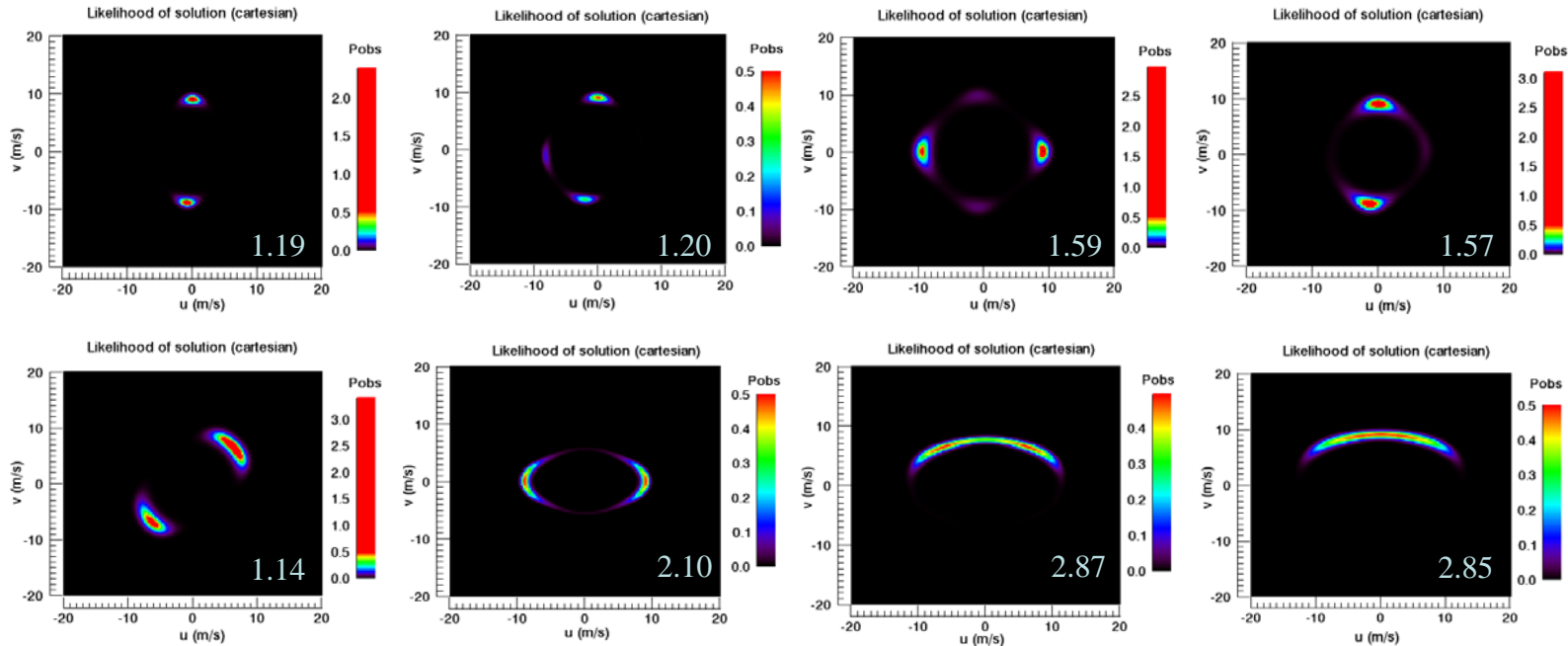
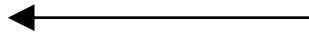
CFOSAT



# FoM – Test cases

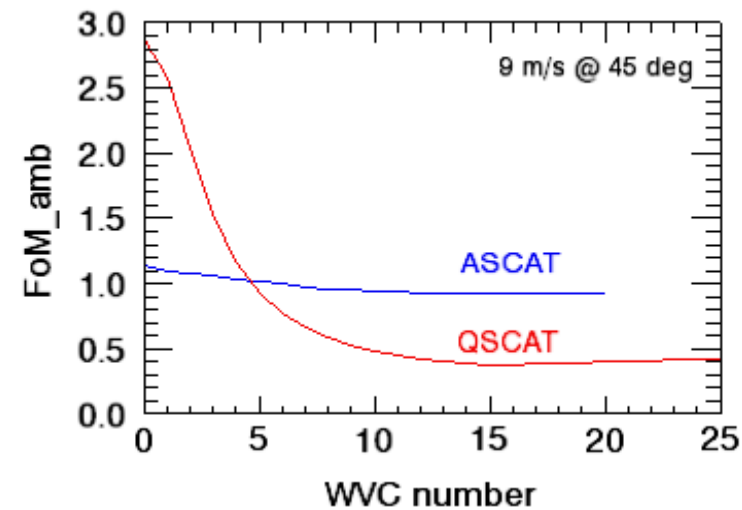
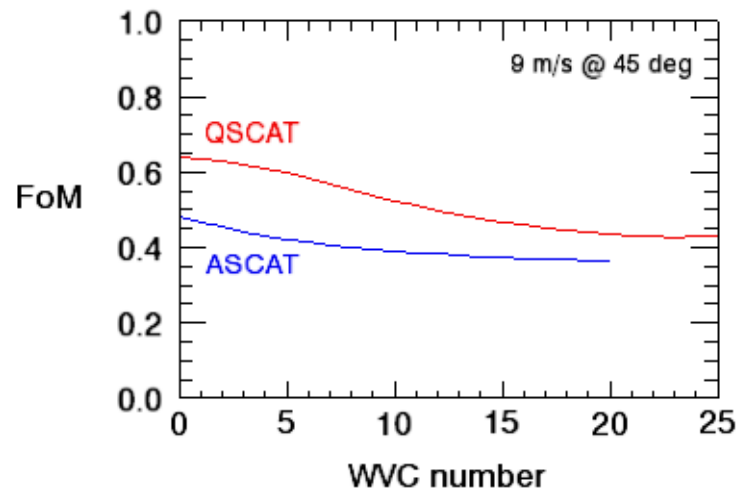


better



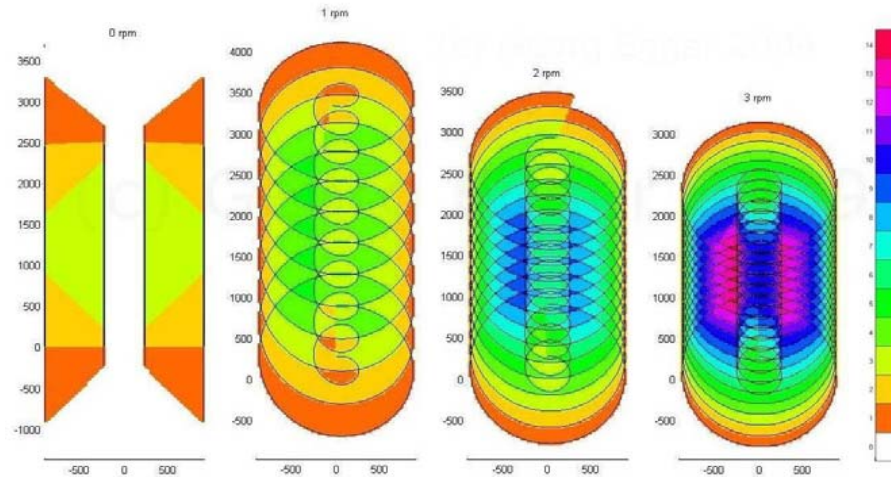
worse

# FoM – Test cases



- For this wind input, ASCAT gives better background constrained information, although with more ambiguity.

# Preliminary SCA assessment

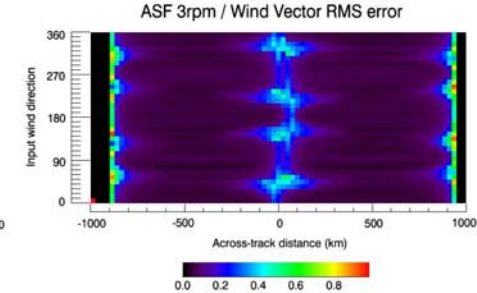
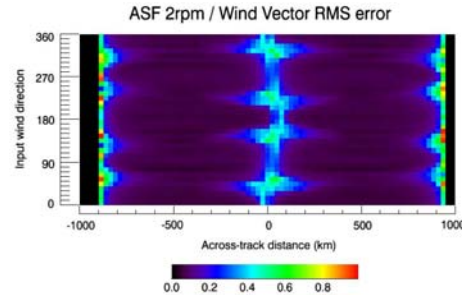
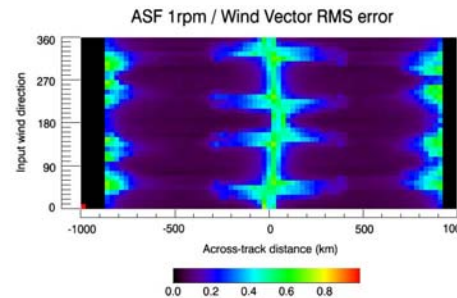
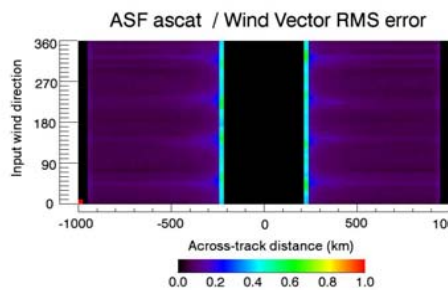


ASCAT  
type

RFSCAT  
(1rpm)

RFSCAT  
(2rpm)

RFSCAT  
(3rpm)

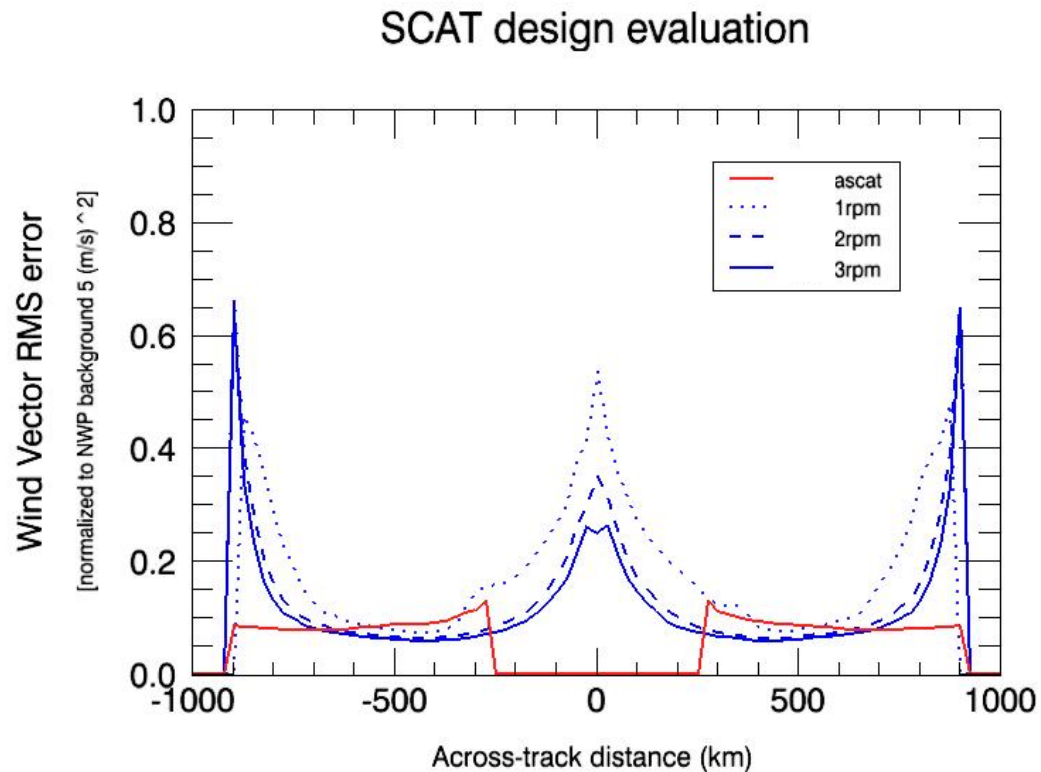


Wind Vector RMS error across swath  
CFOSAT, Oct 2009



# Conclusion

- Post-EPS RFSCAT can equal ASCAT quality and close the nadir gap
- **Accomodation cost ?**



To consider: additional sensitivity studies, like the dual-polarization option.