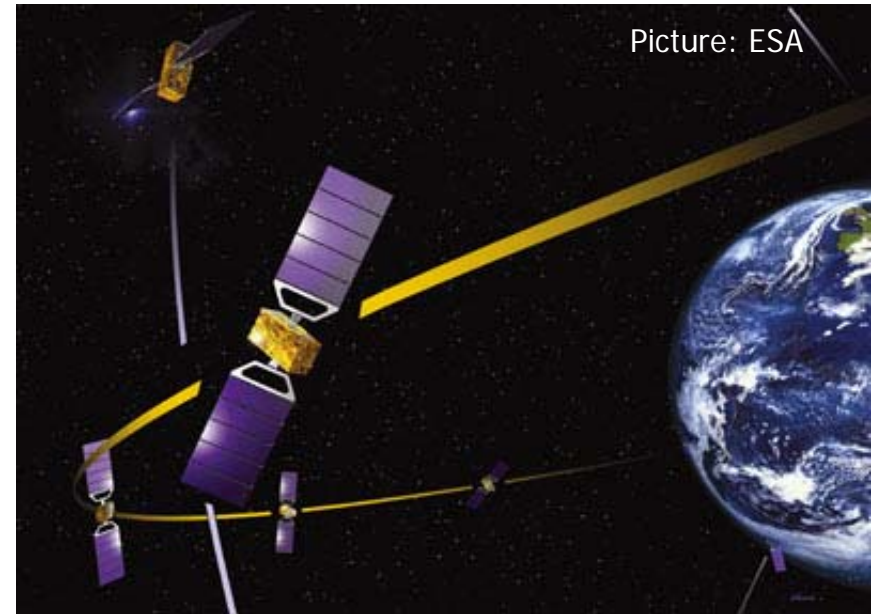


# Satellite Navigation

## Integrity and integer ambiguity resolution



**AE4E08**

**Sandra Verhagen**

**Course 2010 – 2011, lecture 12**

# Today's topics

- Integrity and RAIM
- Integer Ambiguity Resolution
  
- Study Section 7.4 – 7.6 (not LMS algorithm in 7.5.3)

# Integrity: performance measures

**Integrity** = ability of a system to provide timely warnings to users when the system should not be used

- **HPE/VPE** : **Horizontal/Vertical Position Error**  
(not known; due to measurement noise and biases)
- **HPL/VPL** : **Horizontal/Vertical Protection Level**  
horizontal/vertical position is assured to be within region defined by HPL/VPL
- **HAL/VAL** : **Horizontal/Vertical Alarm Limit**  
position error that should result in an alarm being raised
- **TTA** : **Time To Alarm**  
time between occurrence of integrity event (position error too large) and alarm being raised

# Integrity: performance measures

- HPE/VPE : Horizontal/Vertical Position Error  
(not known; due to measurement noise and biases)
- HPL/VPL : Horizontal/Vertical Protection Level  
horizontal/vertical position is assured to be within region defined by HPL/VPL (can be calculated)
- HAL/VAL : Horizontal/Vertical Alarm Limit  
position error that should result in an alarm being raised

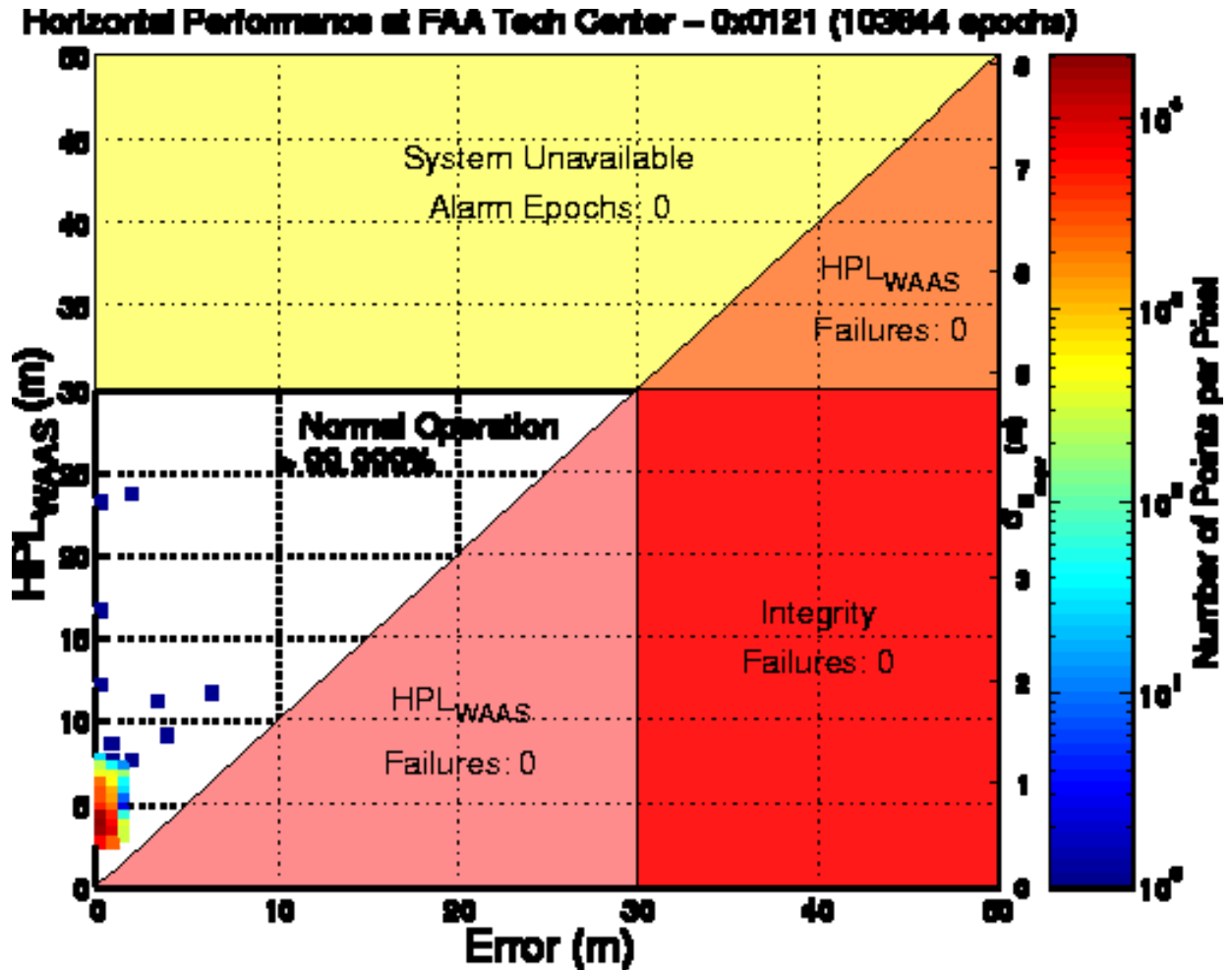
required:  $P(XPE > XPL) < \text{integrity risk}$



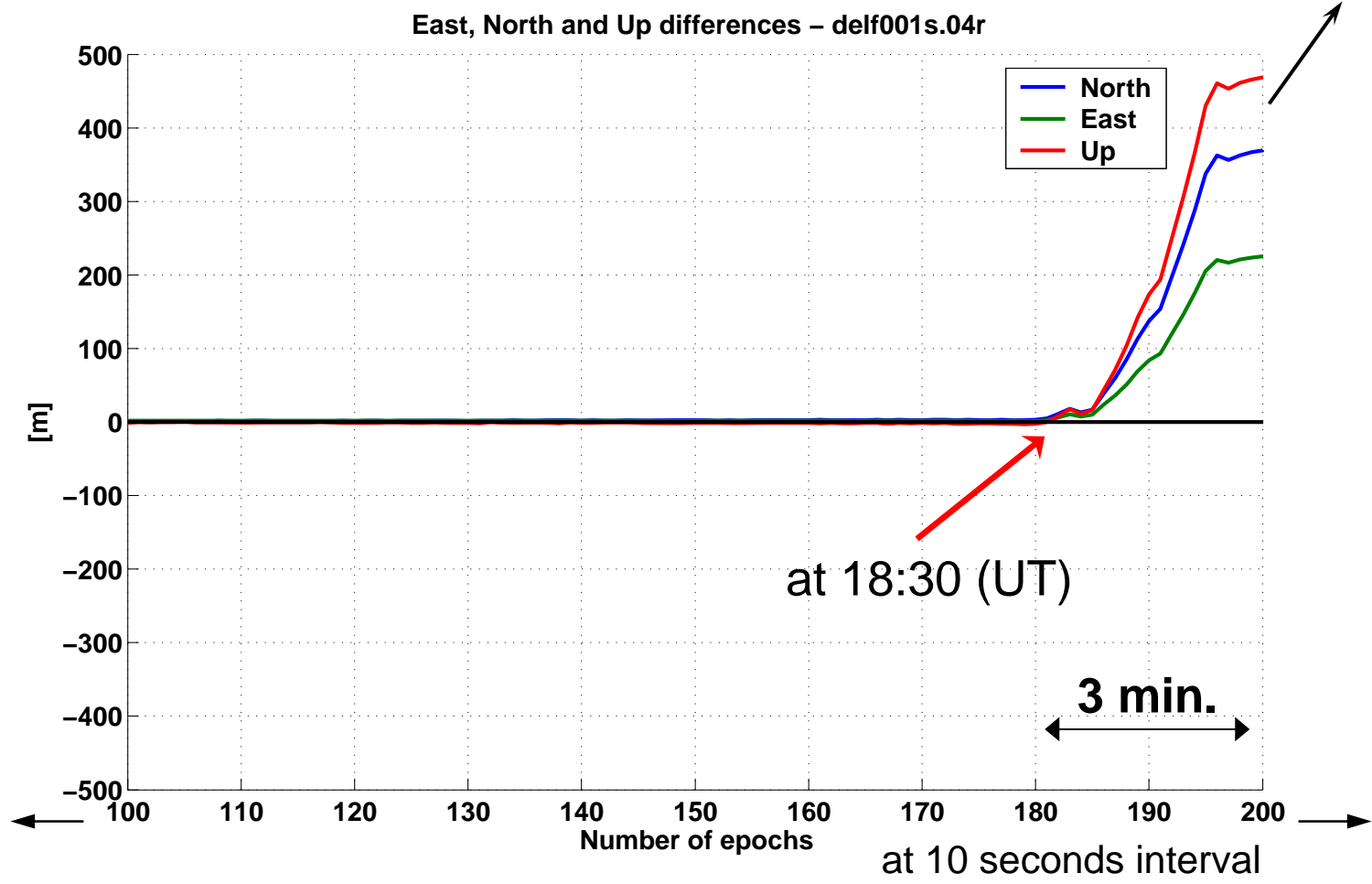
$XPL > XAL \rightarrow \text{alarm, system unavailable}$

# Stanford plots

<http://waas.stanford.edu/metrics.html>

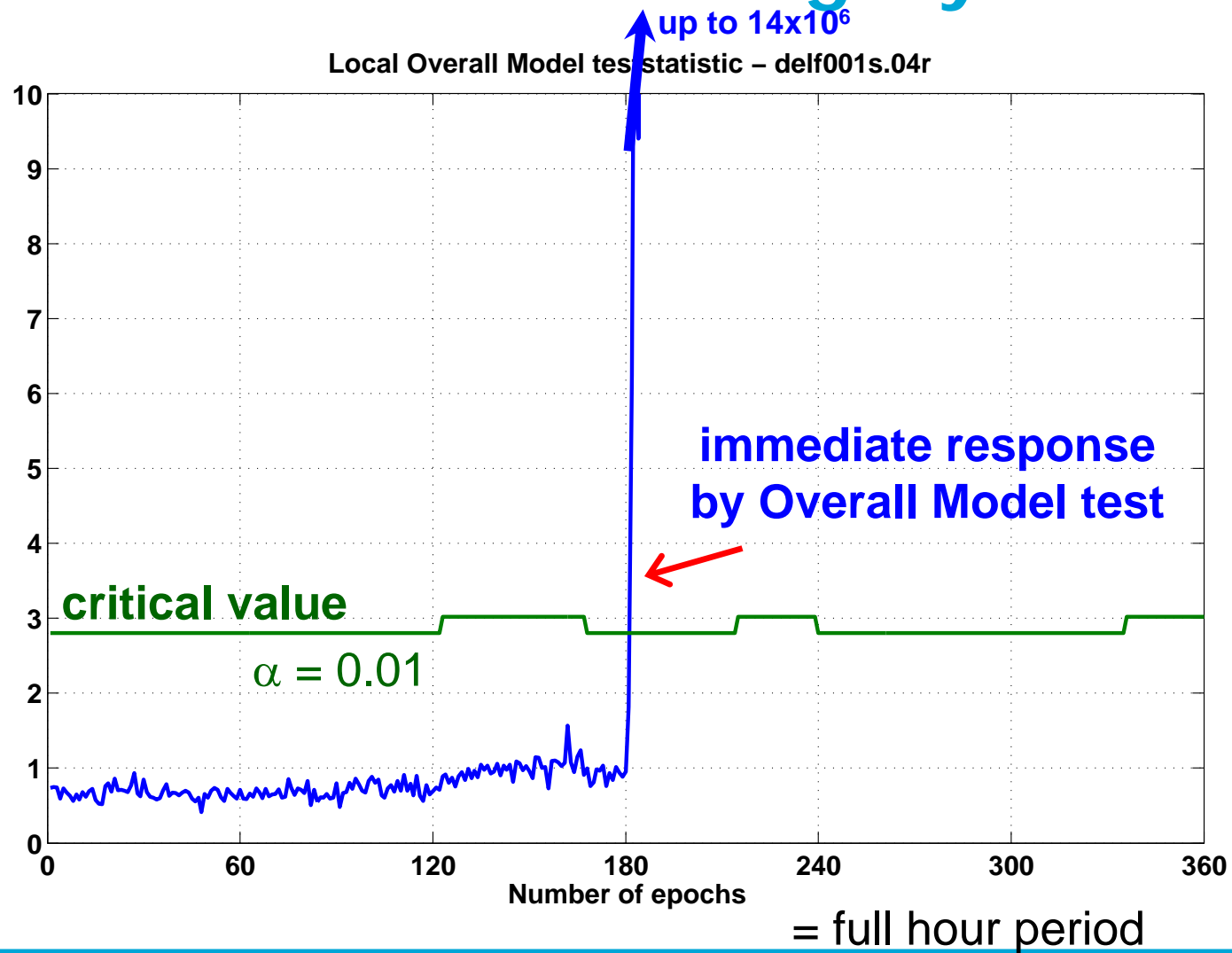


# GPS PRN 23 Anomaly, 1 Jan, 2004



Not noticed by US for 3 hours  
Picked up by EGNOS  
Alternative: check @receiver

# Receiver Autonomous Integrity monitoring



# RAIM - Overall model test

RAIM: detect and correct for errors in GPS data  
@receiver

Overall model test: does  $H_o$  provide good model?

Model:  $\underline{y} = A\underline{x} + \underline{e}$

Measurements:  $y \rightarrow \hat{x} = \left( A^T Q_{yy}^{-1} A \right)^{-1} A^T Q_{yy}^{-1} y$

$$\hat{y} = A\hat{x}$$

Residuals:  $\hat{e} = y - \hat{y}$

“Mismatch”:

$$\underline{T}_{q=m-n} = \underline{\hat{e}}_o^T Q_{yy}^{-1} \underline{\hat{e}}_o$$



# RAIM - Overall model test

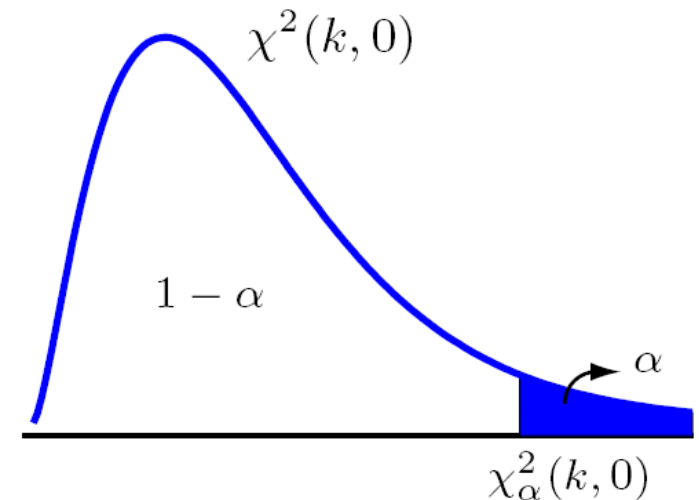
RAIM: detect and correct for errors in GPS data  
@receiver

Overall model test: does  $H_o$  provide good model?

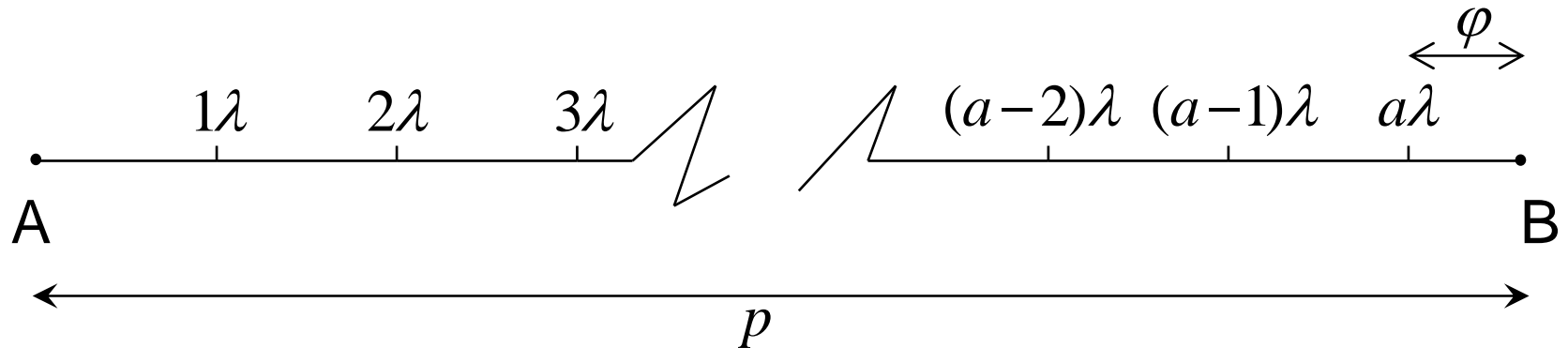
$$\underline{T}_{q=m-n} = \hat{\underline{e}}_o^T Q_{yy}^{-1} \hat{\underline{e}}_o$$

$$H_o : \underline{T}_q \sim \chi^2(m - n, 0)$$

$$H_a : \underline{y} \in \mathbb{R}^m$$



# A Simple Ambiguity Resolution Example



$\rho$  : unknown distance AB

$p$  : measured distance AB ( $\sigma_p = 20$  cm)

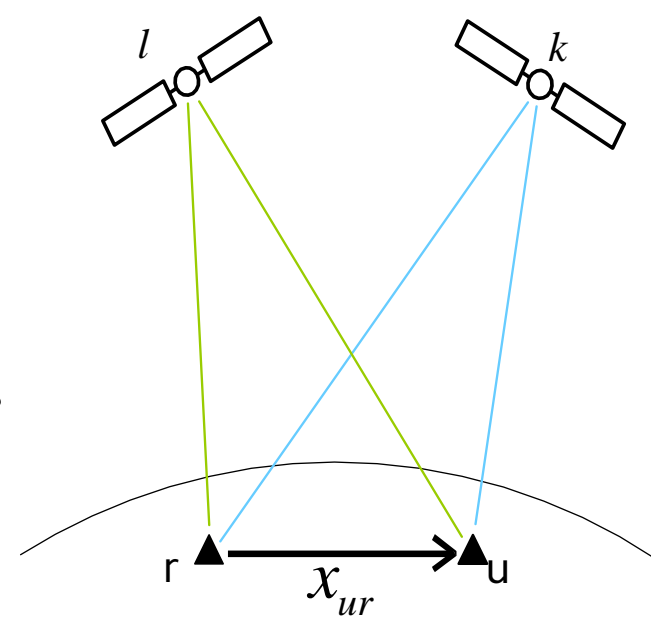
$\varphi$  : measured fraction distance AB ( $\sigma_\varphi = 2$  mm)

$a$  : number of times  $\lambda$  (= known) fits into  $\rho$  ( $a =$  unknown integer)

Model: 
$$E \begin{pmatrix} \varphi \\ p \end{pmatrix} = \begin{pmatrix} 1 & -\lambda \\ 1 & 0 \end{pmatrix} \begin{pmatrix} \rho \\ a \end{pmatrix}, \quad D \begin{pmatrix} \varphi \\ p \end{pmatrix} = \begin{pmatrix} \sigma_\varphi^2 & 0 \\ 0 & \sigma_p^2 \end{pmatrix}, \quad a = \text{integer}$$

# Relative Positioning: Double Differencing

- elimination of **receiver clock errors**
- elimination of **initial receiver phase offsets**
- **DD phase ambiguity** is an **integer** number!



DD code observation:

$$\rho_{ur,i}^{(kl)} = \left(-\mathbf{1}_r^{(kl)}\right)^T \mathbf{x}_{ur} + \mu_i I_{ur}^{(kl)} + T_{ur}^{(kl)} + \varepsilon_{\rho_i,ur}^{(kl)}$$

DD phase observation:

$$\Phi_{ur,i}^{(kl)} = \left(-\mathbf{1}_r^{(kl)}\right)^T \mathbf{x}_{ur} - \mu_i I_{ur}^{(kl)} + T_{ur}^{(kl)} + \lambda_i N_{ur,i}^{(kl)} + \varepsilon_{\Phi_i,ur}^{(kl)}$$

↑  
relative receiver  
position

↑  
integer DD ambiguity

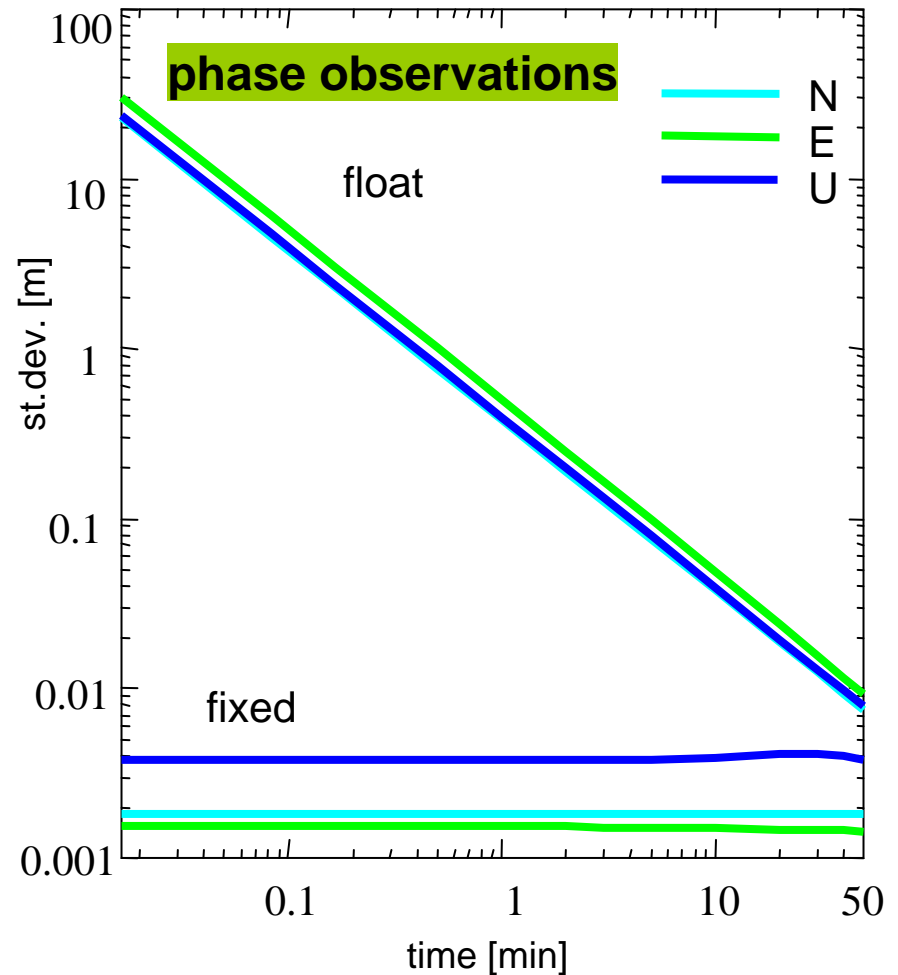
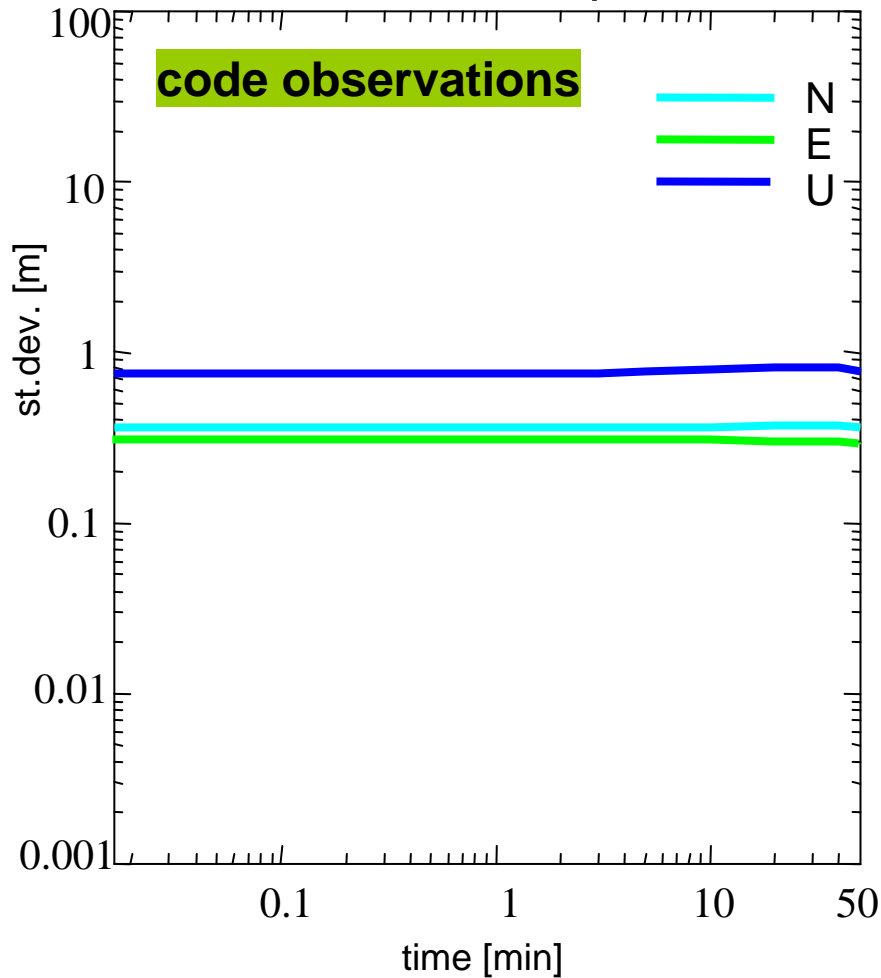
# Resolution of the DD ambiguities

- code observation: **dm** precision
- phase observation: **mm** precision,
  - but: **receiver-satellite geometry** has to change considerably (long observation time) to solve position with mm-cm accuracy
  - if DD ambiguities are resolved to **integers within a short time (or instantaneously)**, positions (and other parameters) can be solved with mm-cm accuracy



# Precision of relative GPS positioning

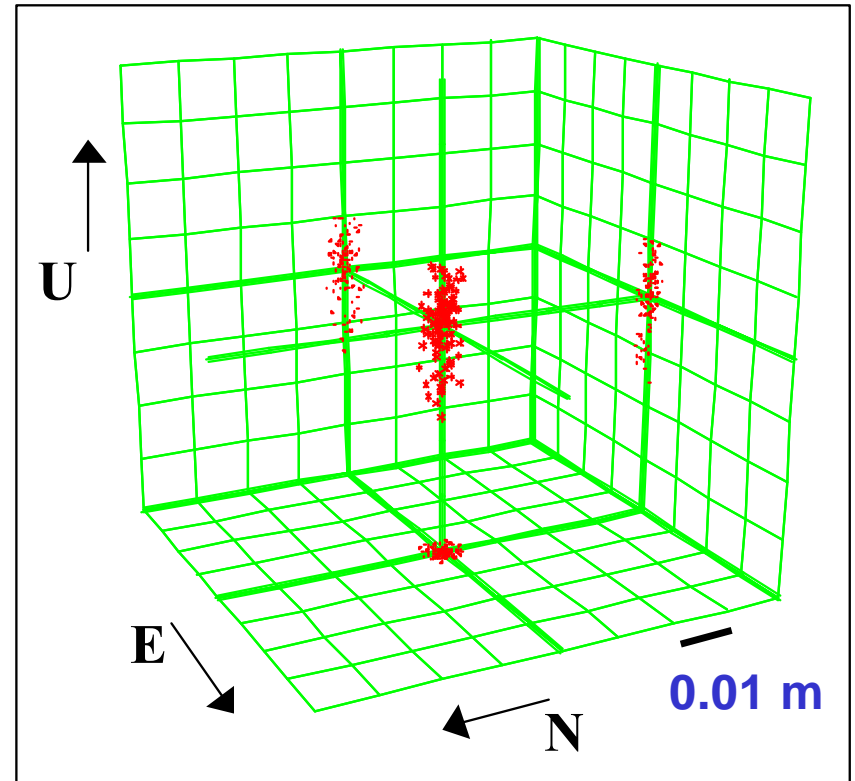
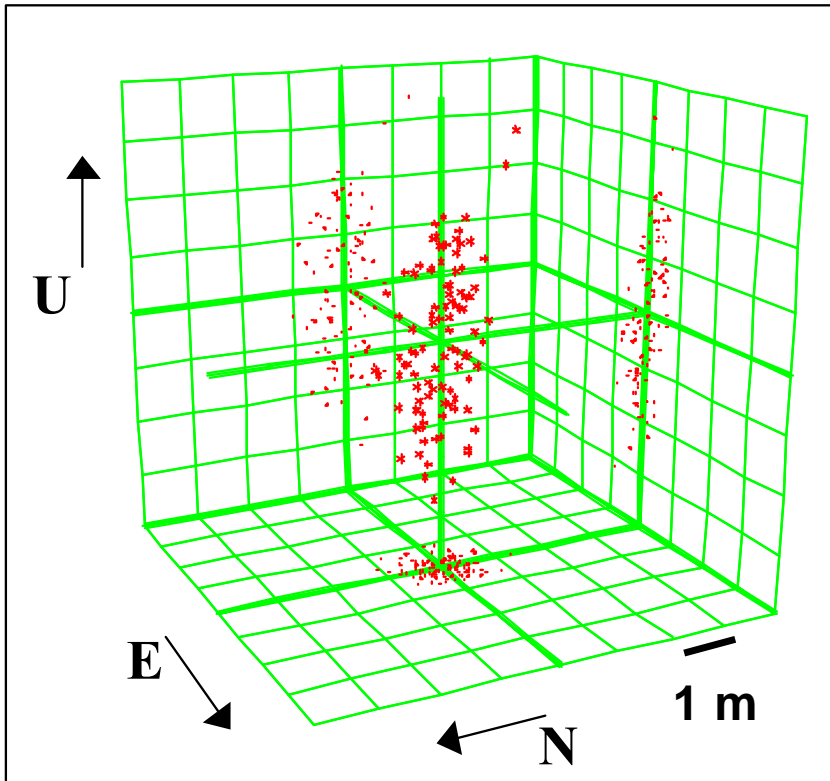
two epochs of data: varying time span



# Precision code vs. phase observations

code observations

phase observations



- both RELATIVE positioning
- phase: provided that the integer ambiguity is KNOWN

# Ionosphere-fixed, -float, -weighted model

- **Ionosphere-fixed model:**
  - **no** differential ionospheric delay parameters
  - observations may be **corrected** a priori for ionosphere
  - for **short** baselines only
  - can already be based on **single-frequency** data
- **Ionosphere-float model:**
  - **estimation** of differential ionospheric delays
  - no a priori corrections
  - for **long** baselines
  - based on at least **dual-frequency** data
- **Ionosphere-weighted model:**
  - **ionosphere corrections from network RTK 'subtracted'**
  - for **medium** to **long** baselines

# GNSS model

In book:

$$\mathbf{y} = \mathbf{A}\mathbf{N} + \mathbf{G}\delta\mathbf{x} + \boldsymbol{\varepsilon}$$

Observation equations:

$$\mathbf{y} = \mathbf{A}\mathbf{a} + \mathbf{B}\mathbf{b} + \mathbf{e}, \quad \mathbf{a} \in \mathbb{Z}^n; \quad \mathbf{Q}_{yy}$$

$\mathbf{y}$  data vector

$\mathbf{a}$  ambiguities

$\mathbf{b}$  baseline coordinates & other unknowns

$\mathbf{Q}_{yy}$  variance-covariance matrix of data



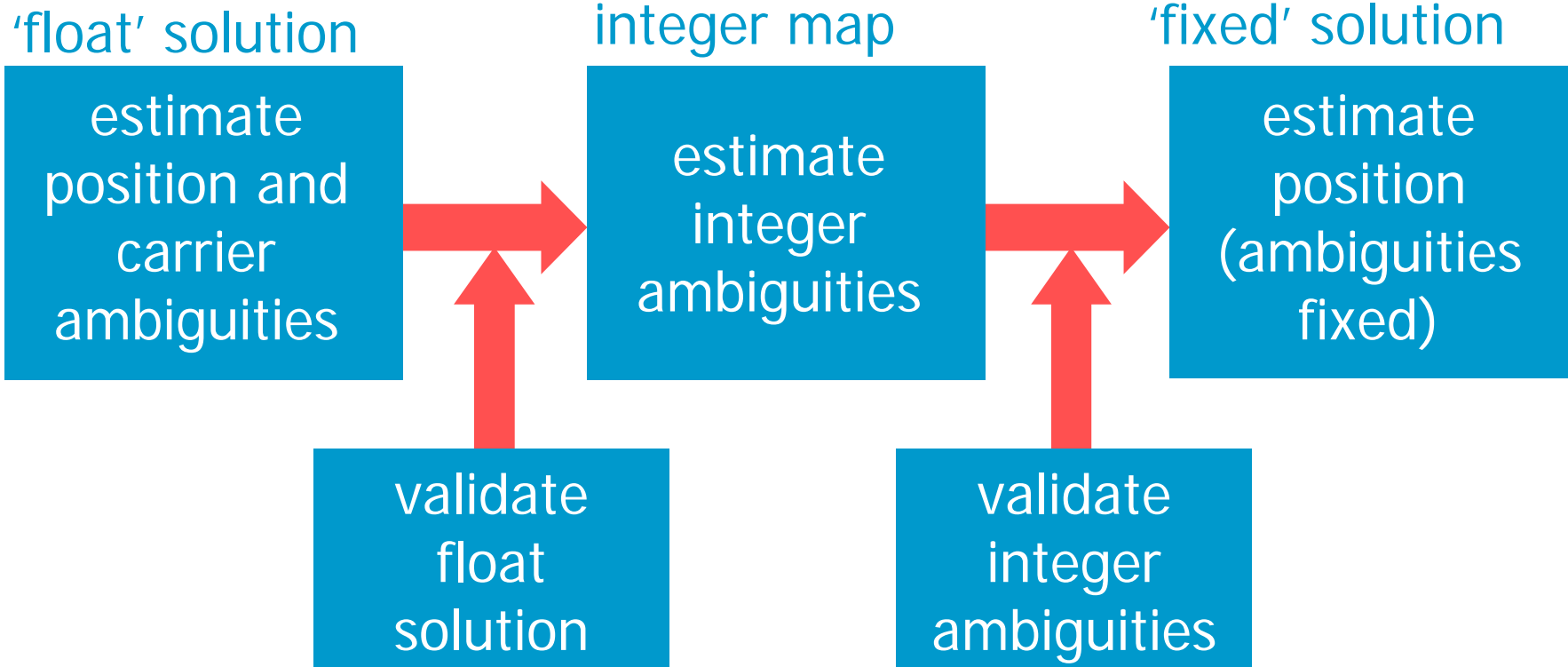
# **SUCCESSFUL** INTEGER AMBIGUITY RESOLUTION

is the **key** to

**FAST** and **PRECISE** GNSS parameter  
estimation

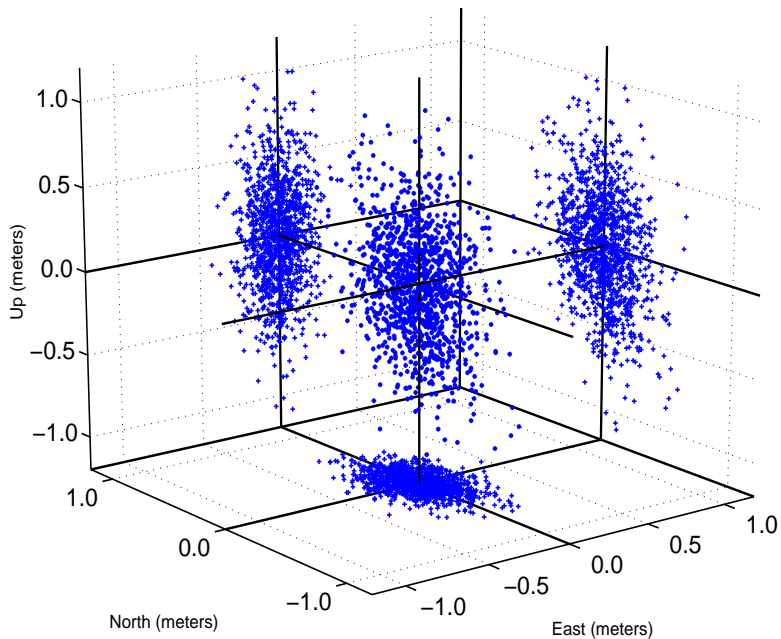
(baseline coordinates, attitude angles, orbit  
parameters, atmospheric delays)

# Integer estimation

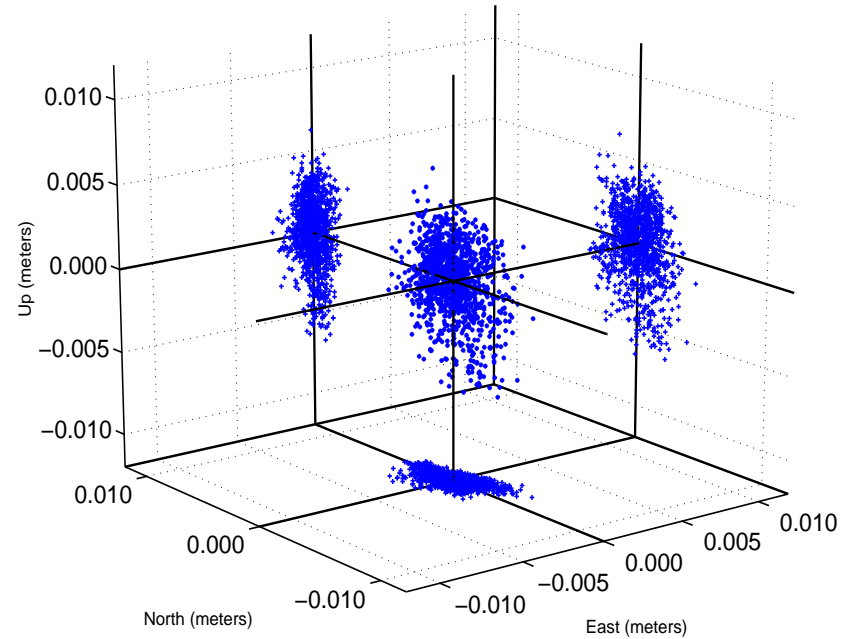


# Float and fixed solution

## Ambiguities not fixed



## Ambiguities fixed

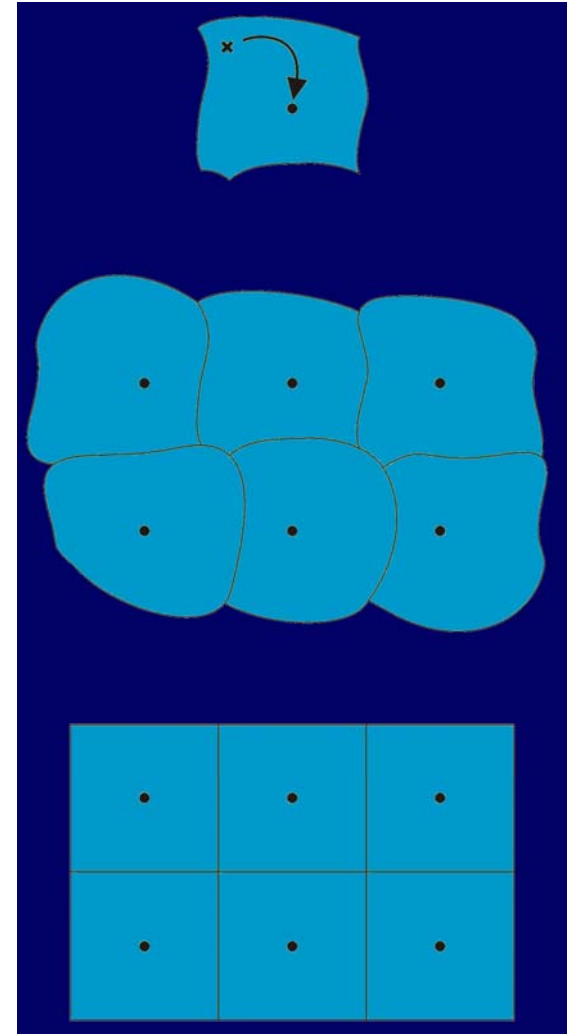


# Integer estimation

integer map  $\hat{\mathbf{a}} \in \mathbb{R}^n \rightarrow S(\hat{\mathbf{a}}) = \check{\mathbf{a}} \in \mathbb{Z}^n$

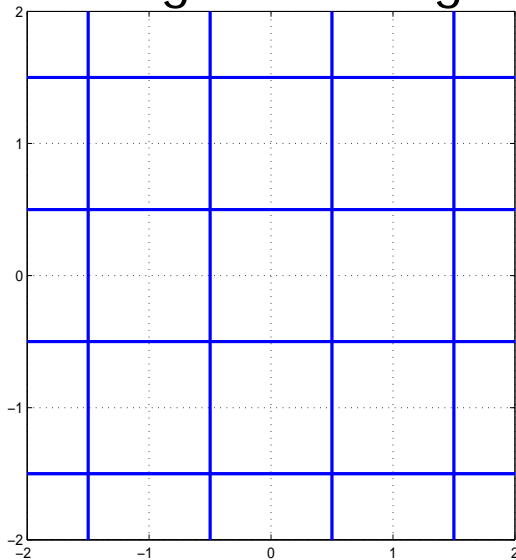
no holes & no overlap  
→ there will always be **ONE** solution

translation invariant

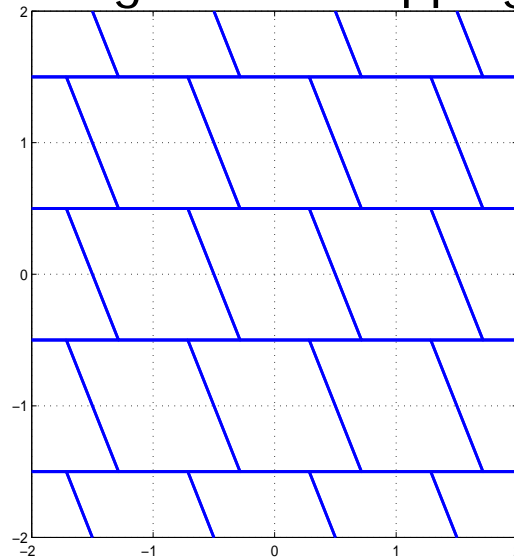


# Different choices of integer estimators

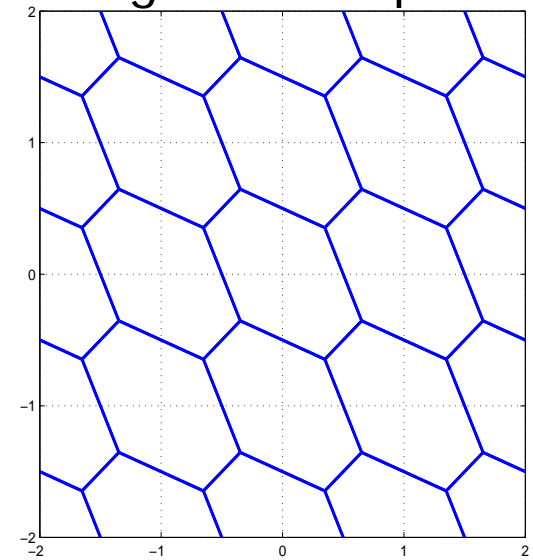
integer rounding



integer bootstrapping



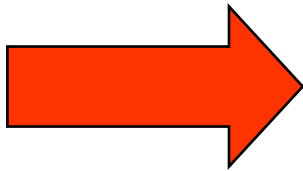
integer least-squares



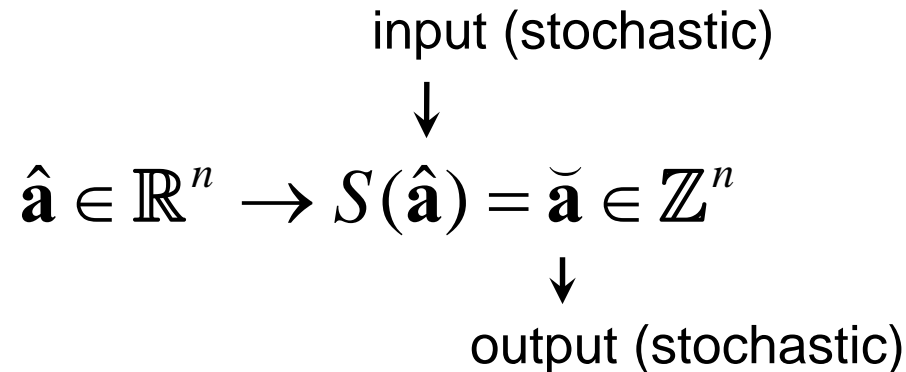
after their pull-in region

# Ambiguity resolution

Integer ambiguities are derived from stochastic observations



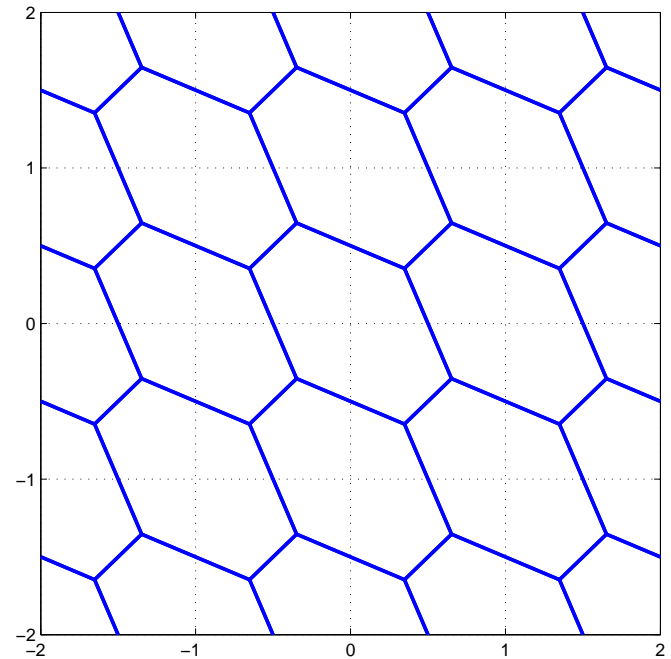
Integer ambiguities are **not deterministic** but *stochastic*



# Integer estimation

Optimal integer estimator: integer least-squares

$$\check{\mathbf{a}} = \arg \min_{\mathbf{z} \in \mathbb{Z}^n} \|\hat{\mathbf{a}} - \mathbf{z}\|_{\mathbf{Q}_{\hat{\mathbf{a}}}}^2$$

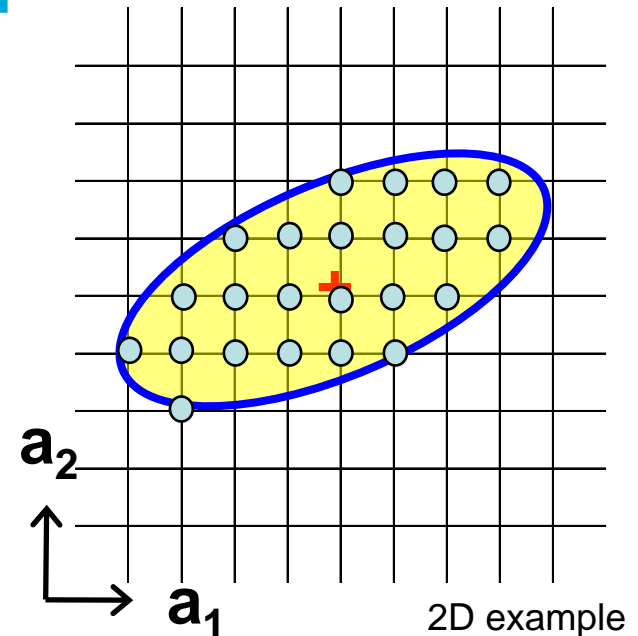


# Ambiguity search space: a (hyper-) ellipsoid

- **centered** at  $\hat{\mathbf{a}}$
- **shape** governed by  $\mathbf{Q}_{\hat{\mathbf{a}}\hat{\mathbf{a}}}$
- find all integers  $\mathbf{z}$  for which

$$(\hat{\mathbf{a}} - \mathbf{z})^T \mathbf{Q}_{\hat{\mathbf{a}}\hat{\mathbf{a}}}^{-1} (\hat{\mathbf{a}} - \mathbf{z}) \leq \chi^2$$

- $\chi^2$  should be set such that search space contains at least **one** integer vector
- select the  $\mathbf{z}$  which provides minimum

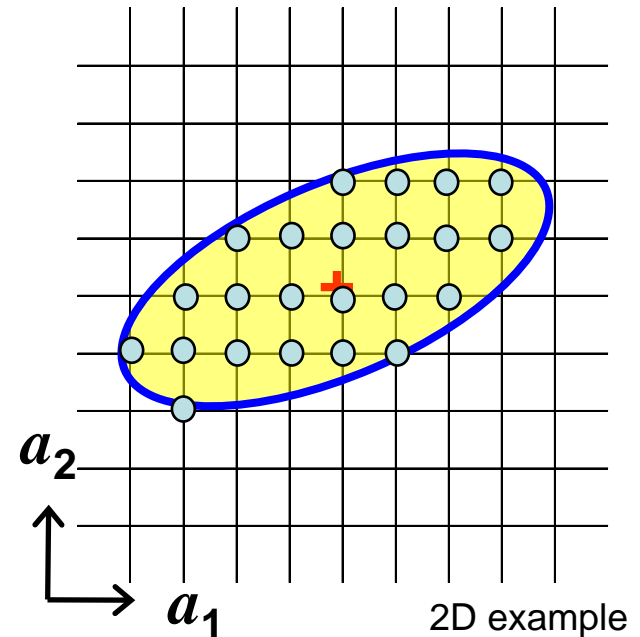


- +  $\hat{\mathbf{a}}_1, \hat{\mathbf{a}}_2$  (float solution)
- candidate integer solution



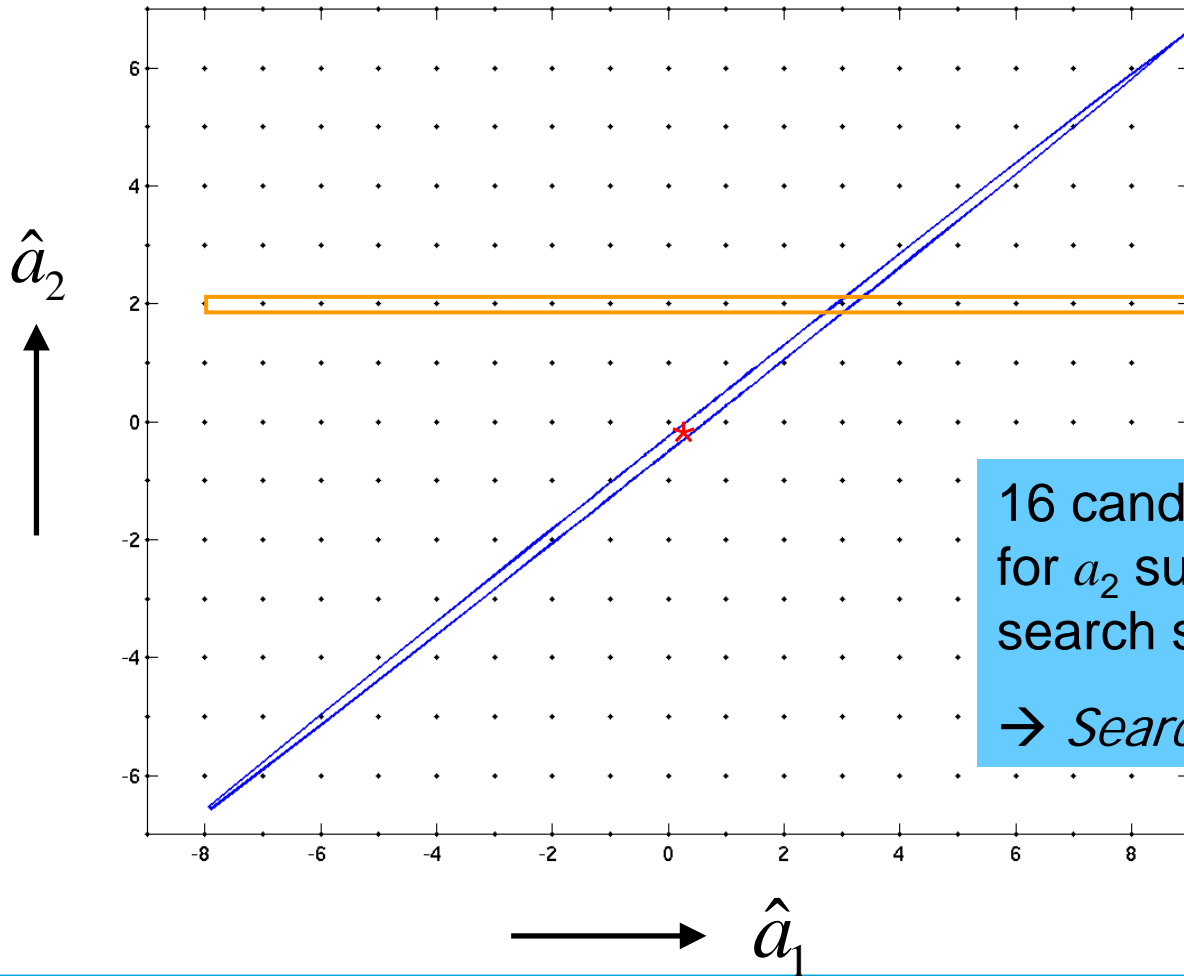
# Integer ambiguity resolution

- Float solution: least-squares
- Integer search: find integer solution with shortest weighted distance to float solution (weighted by variance-covariance matrix of float ambiguities)
- Search difficult due to correlations
- LAMBDA: transformation of search space to make it efficient



# Example: Ambiguity search space

## Two dimensions, geometry-free, short baseline

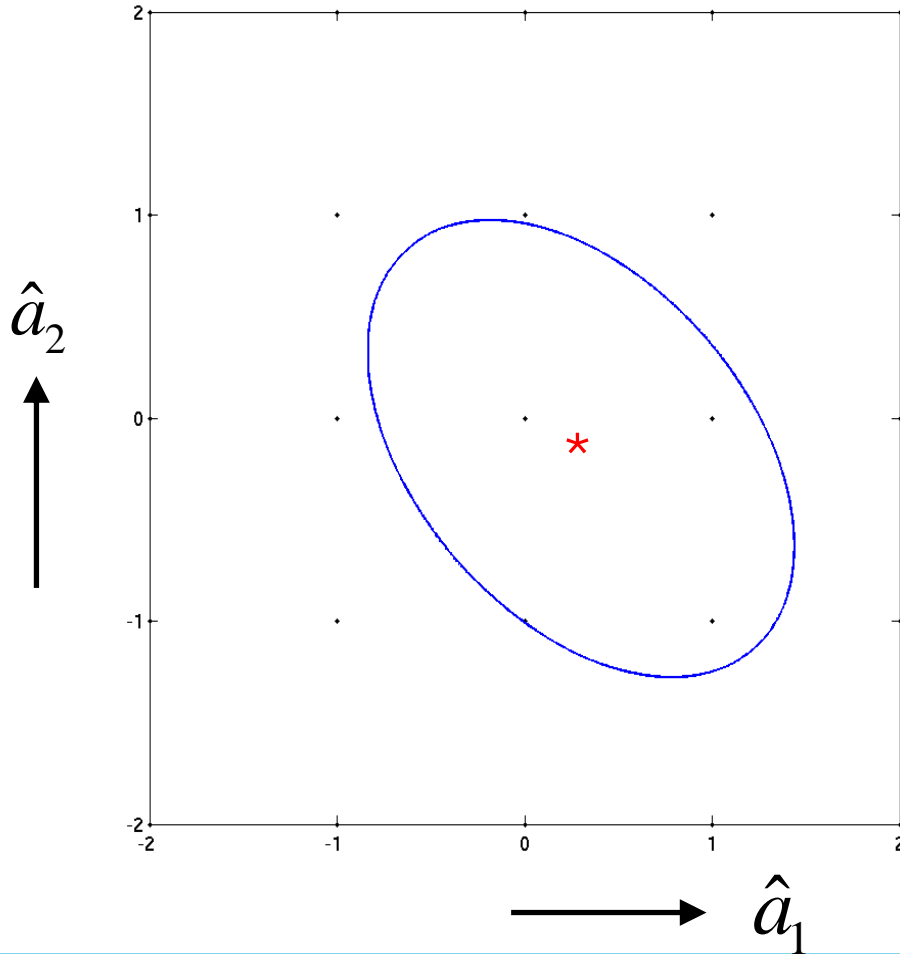


16 candidates for  $a_1$ , but only one for  $a_2$  such that  $(a_1, a_2)$  inside search space

→ *Search is inefficient*

# Example: Ambiguity search space

## Two dimensions, geometry-free, short baseline



After decorrelation

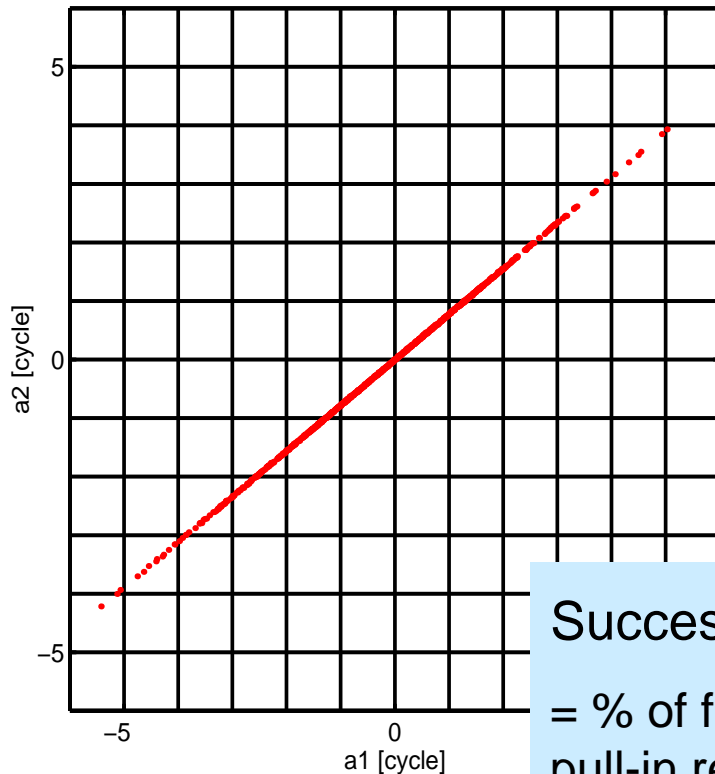
Number of candidates **INSIDE** search space is same

→ *Search is efficient*

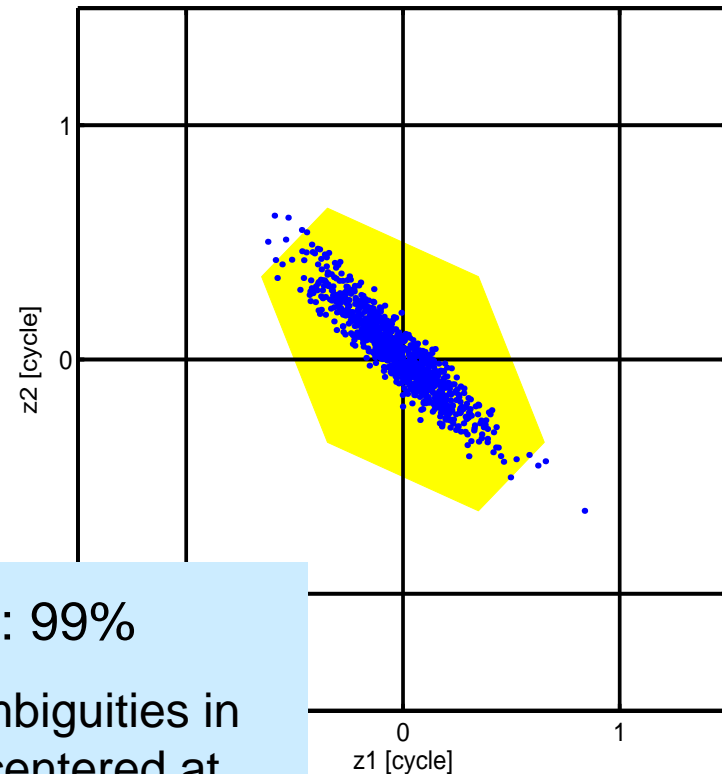
# Ambiguity estimation and success rate

## Example based on real data (1000 epochs)

Distribution of original ambiguities



Distribution of transformed ambiguities



Success rate: 99%

= % of float ambiguities in pull-in region centered at correct integer value

# Integer ambiguity resolution

Successful ambiguity resolution depends on precision of float solution, which depends on:

- baseline length (tropo + iono delays)
- satellite geometry
- precision of code and phase observations
- # frequencies

→ Change in satellite geometry helps (long duration)

# LAMBDA method

Integer estimation:

- optimal : maximum success rate
- efficient : (near) real-time

LAMBDA

Applicable to wide variety of models

- With or without relative satellite-receiver geometry
- Stationary or moving receivers
- With or without atmospheric delays
- Single- or multi-baseline
- One, two, three or more frequencies (any GNSS)

LAMBDA

# Baseline models

## Parameters

	Geometry-free	Roving-receiver	Stationary-receiver
Ranges	N		
Station coordinates		N	C
Ambiguities	C	C	C
Ionospheric delays	N <sup>*)</sup>	N <sup>*)</sup>	N <sup>*)</sup>

**N** - New parameter introduced for each observation epoch

**C** - Constant parameter for entire observation period

**\*)** - Long baselines only

# Ambiguity Resolution Methods

- Search in the 3-dimensional position space (e.g. ambiguity function method); *Now deprecated*
- Linear combination of code and phase (using widelane/narrowlane combinations)
  - performance worse with AS
  - has been improved by LAMBDA: 2-dimensional ambiguity resolution/search problem
    - **Geometry-free model**
- Search in the  $n$ -dimensional ambiguity space
  - **Geometry-based model**



# Summary and outlook

- We covered it all! (except for the applications)

Next:

Applications: your presentations

Exam preparation: check blackboard!