

# Efficient calculation of covariances for astrometric data in the Gaia Catalogue



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**What kind of covariances do we expect between astrometric parameters?**



**How should we use the covariances?**



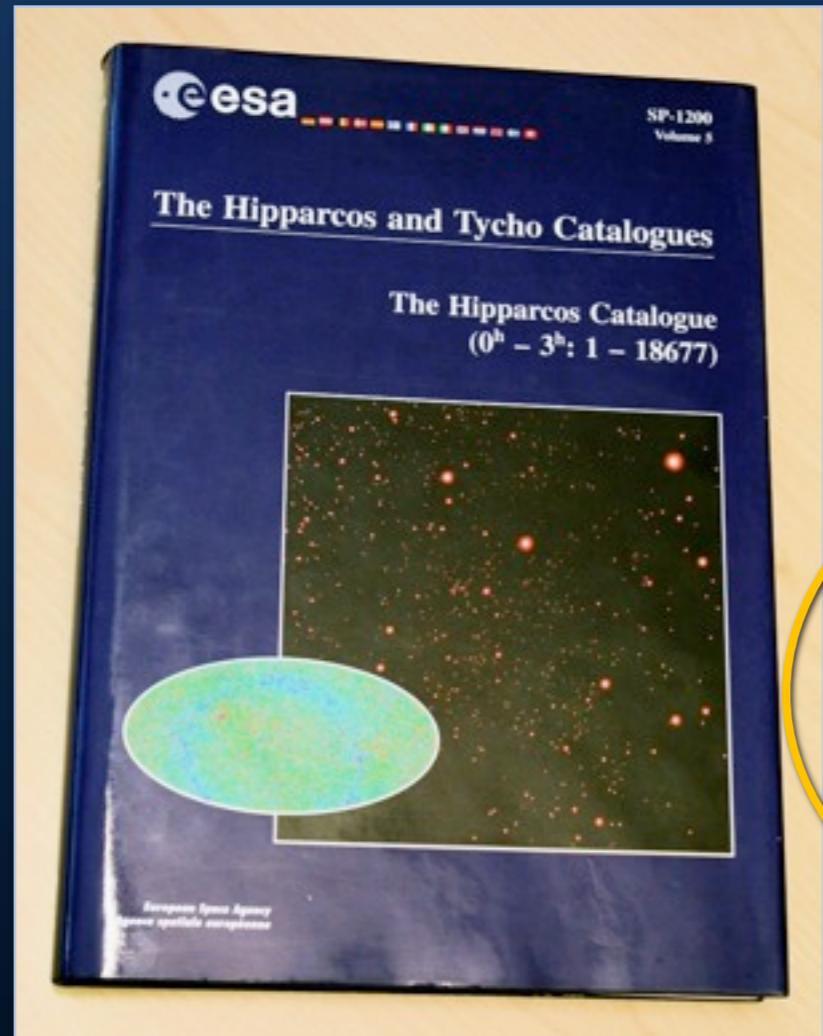
**How can this be done efficiently?**

# Covariance & correlation

$x_i$  and  $x_j$  are estimated values of astrometric parameter  $i$  and  $j$ .

$$\begin{aligned} \text{Cov}[x_i, x_j] &= E[e_i e_j] & e_i &= (x_i - x_{i,\text{true}}) \\ &= \rho_{ij} \sigma_i \sigma_j & (= \sigma_i^2 \text{ for } i = j) \\ \rho_{ij} &= \frac{E[e_i e_j]}{\sqrt{E[e_i^2] E[e_j^2]}} & \text{correlation coefficient} \end{aligned}$$

# Hipparcos catalogue



5 astrometric parameters

- estimated values and their standard errors

$$\alpha \ \delta \ \varpi \ \mu_\alpha^* \ \mu_\delta$$

$$\sigma_\alpha \ \sigma_\delta \ \sigma_\varpi \ \sigma_{\mu_\alpha^*} \ \sigma_{\mu_\delta}$$

- Correlation between astrometric parameters of each star

all 10 combinations

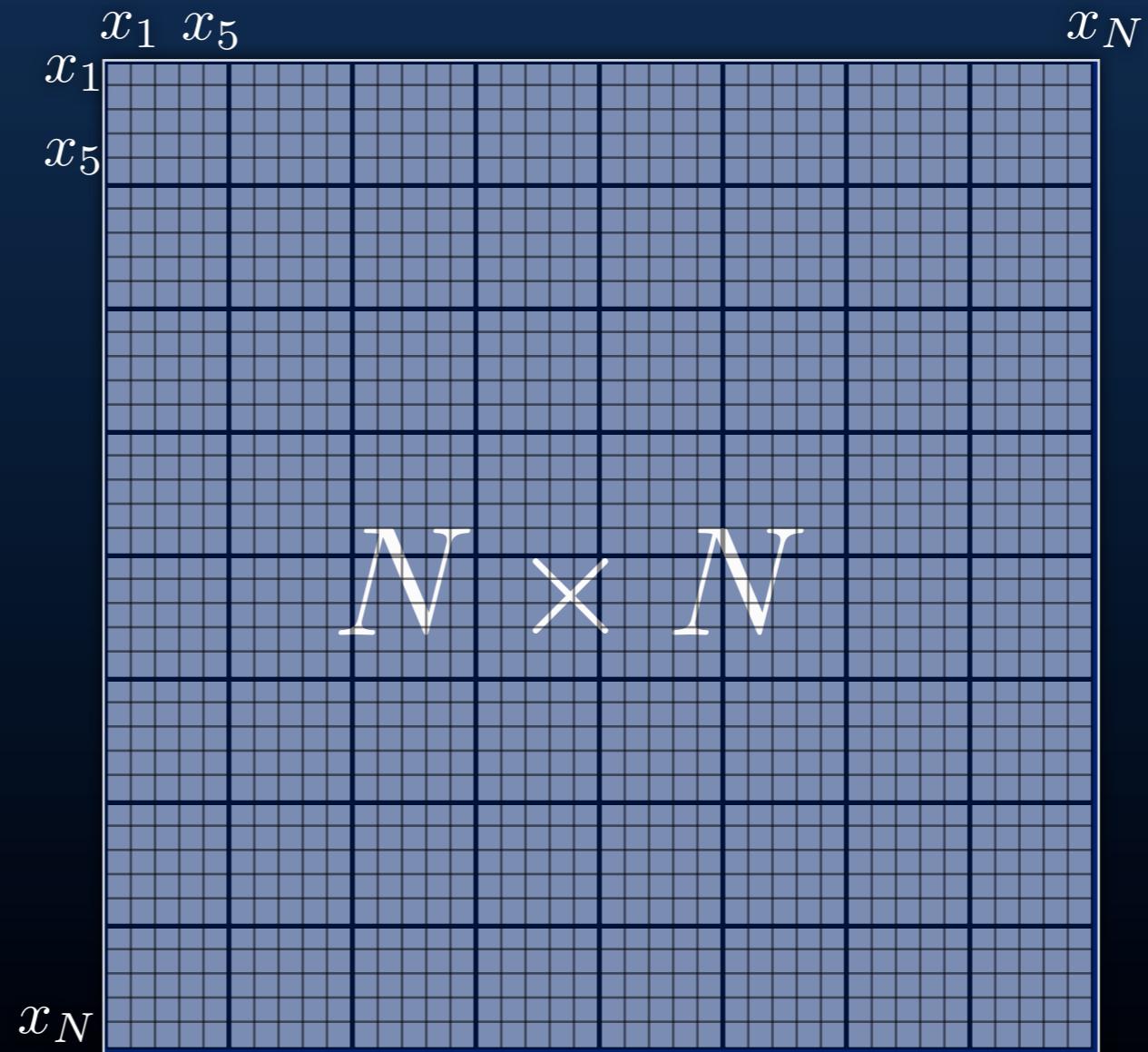
Covariances

Number HIP 1 2	Descriptor: epoch J1991.25							Position: epoch J1991.25				Par. π mas 11	Proper Motion			Standard Errors					Astrometric Correlations (%)										Soln	
	RA h m s			Dec ±° ' "		V mag		α deg 8	(ICRS)		δ deg 9		μ <sub>α*</sub> mas/yr 12	μ <sub>δ</sub> mas/yr 13	α* mas 14	δ mas 15	π mas 16	μ <sub>α*</sub> mas/yr 17	μ <sub>δ</sub> mas/yr 18	δ α*	π α*	π δ	μ <sub>α*</sub> α*	μ <sub>α*</sub> δ	μ <sub>α*</sub> π	μ <sub>δ</sub> δ	μ <sub>δ</sub> π	μ <sub>δ</sub> α*	F1 %	F2		
1	00 00 00.22	+01 05 20.4	9.10	H	0.000 911 85	+01.089 013 32		3.54	-5.20	-1.88	1.32	0.74	1.39	1.36	0.81	+32	-7	-11	-24	+9	-1	+10	-1	+1	+34	0	0.74					
2	00 00 00.91	-19 29 55.8	9.27	G	0.003 797 37	-19.498 837 45	+	21.90	181.21	-0.93	1.28	0.70	3.10	1.74	0.92	+12	-14	-24	-29	+1	+21	-2	-19	-28	+14	2	1.45					
3	00 00 01.20	+38 51 33.4	6.61	G	0.005 007 95	+38.859 286 08		2.81	5.24	-2.91	0.53	0.40	0.63	0.57	0.47	+6	+9	+4	+43	-1	-6	+3	+24	+7	+21	0	-0.45					
4	00 00 02.01	-51 53 36.8	8.06	H	0.008 381 70	-51.893 546 12		7.75	62.85	0.16	0.53	0.59	0.97	0.65	0.65	-22	-9	-3	+24	+20	+8	+18	+8	-31	-18	0	-1.46					
5	00 00 02.39	-40 35 28.4	8.55	H	0.009 965 34	-40.591 224 40		2.87	2.53	9.07	0.64	0.61	1.11	0.67	0.74	+10	+24	+6	+26	-10	+20	-16	-30	-19	+6	0	-1.24					

# Source parameter covariance matrix: $C_{SS}$

$N$  astrometric parameters:  $\vec{x} = (x_1, \dots, x_N)$

$$\vec{C}_{SS} = \text{Cov}(\vec{x}) =$$



# Within-source correlations

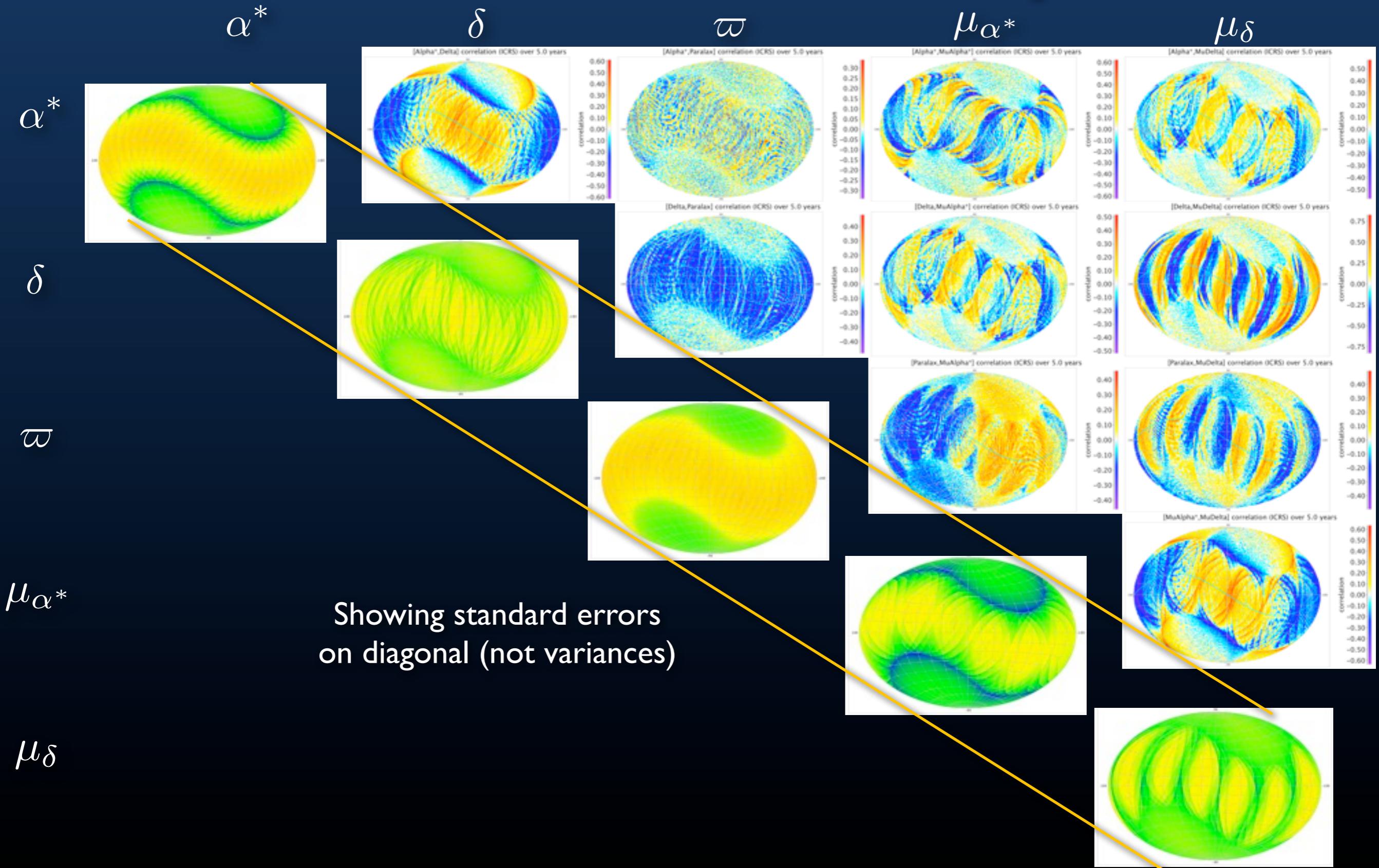
$N$  astrometric parameters:  $\vec{x} = (x_1, \dots x_N)$

Symmetric covariance  
matrix per source =

	$x_1$				$x_5$
$x_1$					
$x_5$					

# Within-source correlations

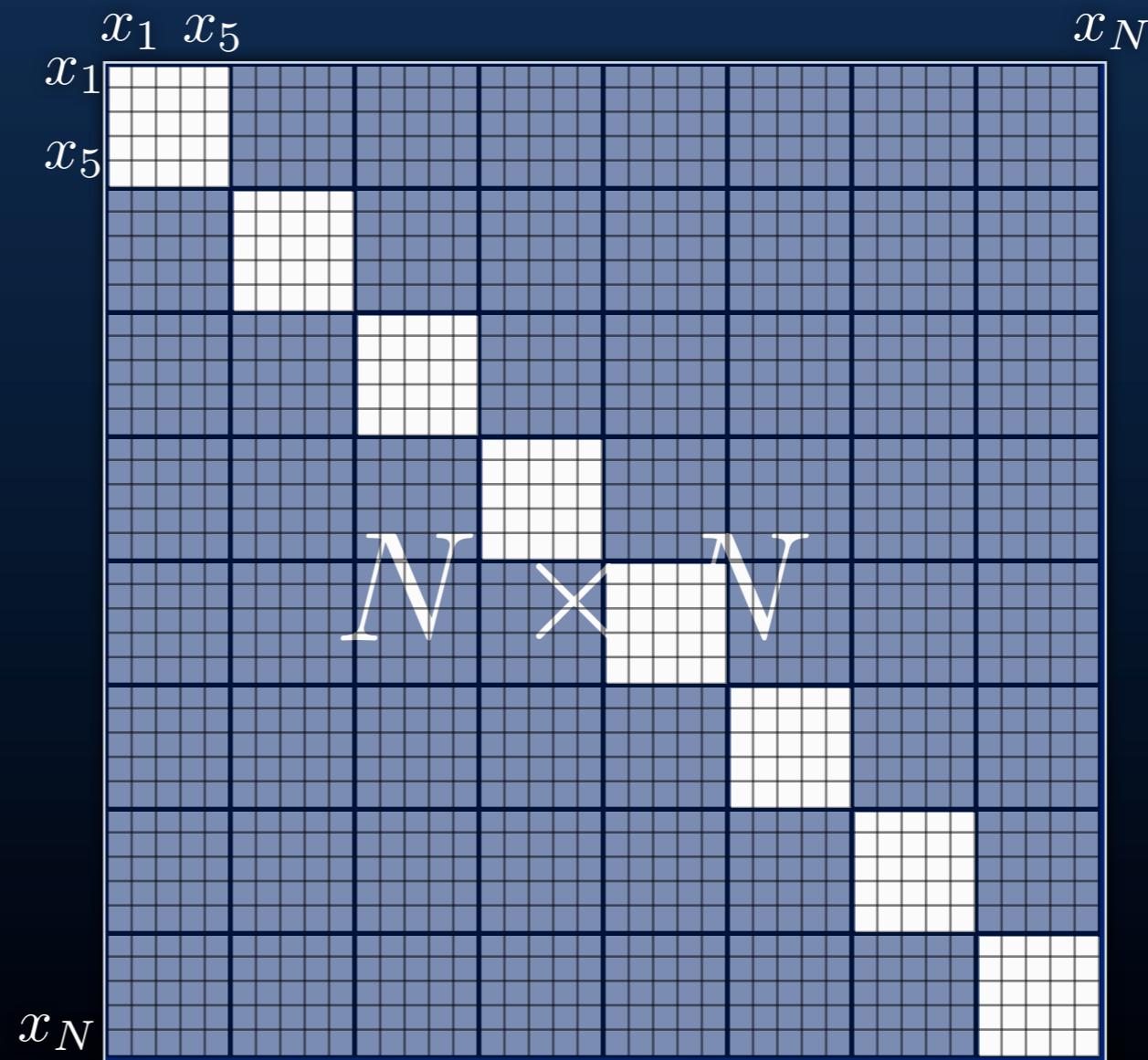
Correlation coefficients off-diagonal:



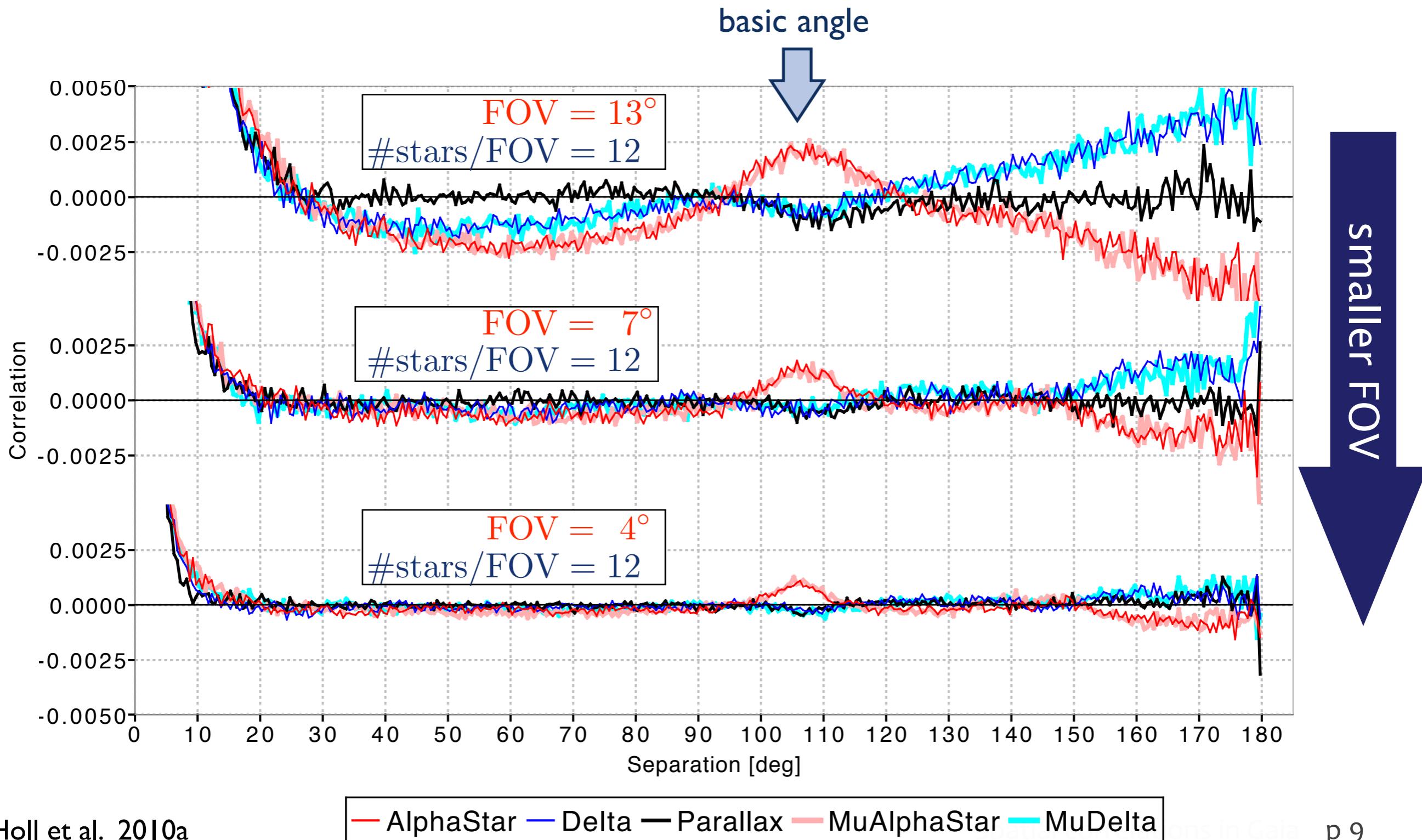
# Between-sources correlations

$N$  astrometric parameters:  $\vec{x} = (x_1, \dots x_N)$

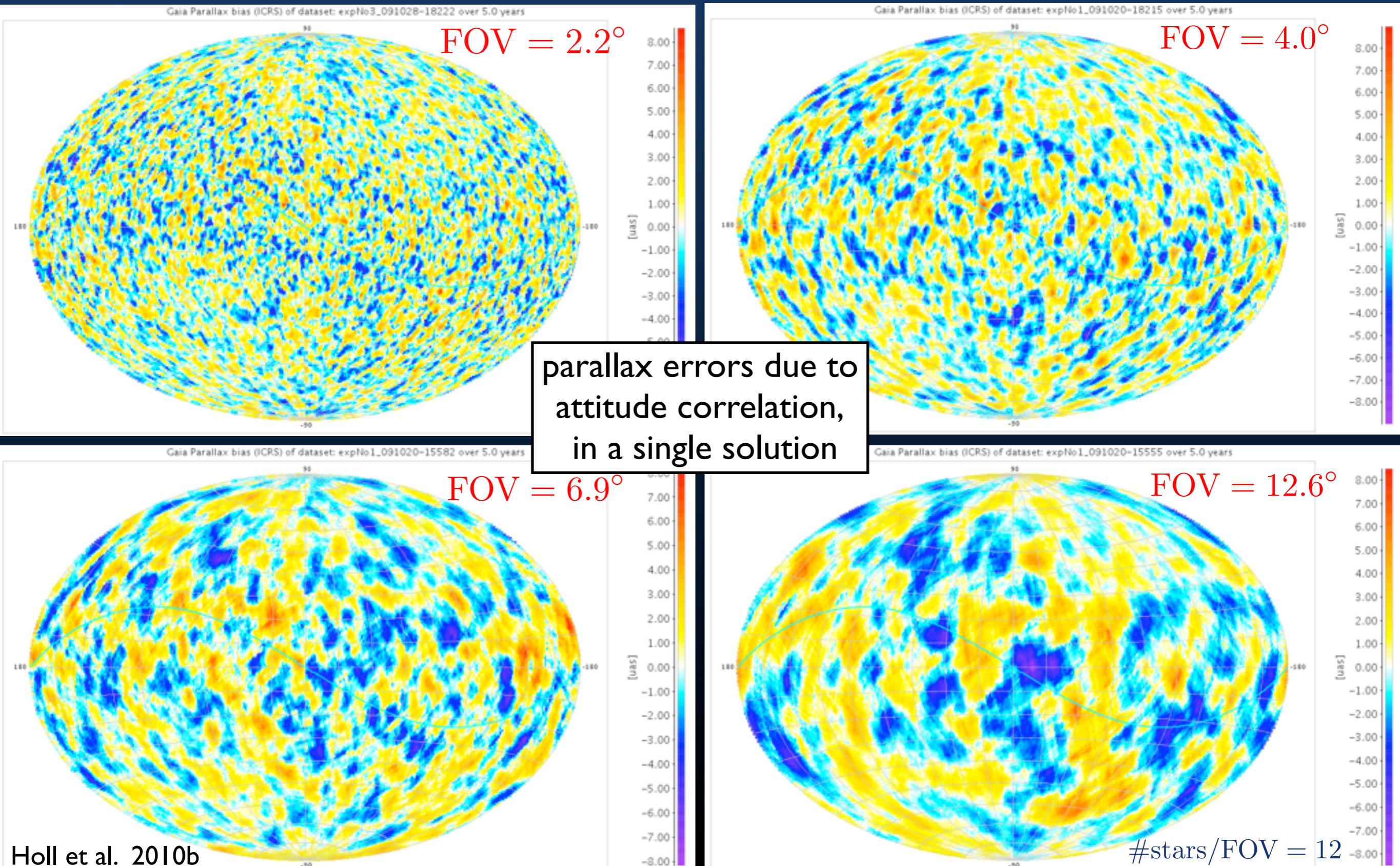
$$\vec{C}_{SS} = \text{Cov}(\vec{x}) =$$



# Between-sources correlations (scaled)



# Spatial correlations due to attitude





What kind of covariances do we expect between astrometric parameters?



**How should we use the covariances?**



How can this be done efficiently?

# What you want to know

Mean parallax to a star cluster with  $n$  stars.

$$y = (\varpi_1 + \dots + \varpi_n)/n$$
$$\sigma_y^2 = ?$$

Assuming  $y = f(\vec{x})$  is linear for small errors, the variance of  $y$  is:

$$\text{Cov}(y) = \sigma_y^2 = \left( \frac{\partial y}{\partial \vec{x}} \right) \vec{C}_{SS} \left( \frac{\partial y}{\partial \vec{x}} \right)' = \sum_i \sum_j \frac{\partial y}{\partial x_i} \frac{\partial y}{\partial x_j} (C_{SS})_{ij}$$

$$= \begin{matrix} x_1 & & x_N \\ & 1 \times N & \\ x_N & & x_N \end{matrix} \quad \boxed{\begin{matrix} x_1 & & x_N \\ & \vec{C}_{SS} = \text{Cov}(\vec{x}) & \\ & N \times N & \end{matrix}}$$

$\begin{bmatrix} 1 \\ \vdots \\ N \end{bmatrix}$

# What you want to know

Mean parallax to a star cluster with  $n$  stars.

$$y = (\varpi_1 + \dots + \varpi_n)/n \quad y = f(\vec{x})$$
$$\sigma_y^2 = n^{-2} \sum_{i \in n} \sum_{j \in n} (C_{SS})_{ij}$$

Assuming  $y = f(\vec{x})$  is linear for small errors, the variance of  $y$  is:

$$\text{Cov}(y) = \sigma_y^2 = \left( \frac{\partial y}{\partial \vec{x}} \right) \vec{C}_{SS} \left( \frac{\partial y}{\partial \vec{x}} \right)' = \sum_i \sum_j \frac{\partial y}{\partial x_i} \frac{\partial y}{\partial x_j} (C_{SS})_{ij}$$

$$\vec{C}_{SS} = \text{Cov}(\vec{x})$$

$N \times N$

$$= \begin{matrix} x_1 & & x_N \\ & 1 \times N & \\ x_N & & x_N \end{matrix}$$

$\begin{bmatrix} 1 \\ \vdots \\ N \end{bmatrix}$

# What you want to know

In general for  $M$  different scalar quantities:

$$\vec{y} = (y_1, \dots, y_M) \quad y_m = f_m(\vec{x})$$

$$\text{Cov}(\vec{y}) = \vec{J} \vec{C}_{SS} \vec{J}' \quad J_{mj} = \partial y_m / \partial x_j$$

$$\begin{matrix} \text{Cov}(\vec{y}) \\ M \times M \end{matrix} = \begin{matrix} y_1 & x_1 & x_N \\ & \vec{J} & \\ y_M & & M \times N \end{matrix}$$

$$\vec{C}_{SS} = \text{Cov}(\vec{x})$$

$N \times N$

$$\vec{J}'$$

$N \times M$

Only consider  $n$  active parameters (non-zero columns in  $\vec{J}'$ ),  
then typically:

$$\begin{matrix} \text{Cov}(\vec{y}) \\ M \times M \end{matrix} \quad M(M-1)/2 \ll n(n-1)/2$$

(number of non-redundant elements)

$$\vec{C}_{SS} = \text{Cov}(\vec{x})$$

$N \times N$

$\square \leftarrow n \times n$



What kind of covariances do we expect between astrometric parameters?



How should we use the covariances?



**How can this be done efficiently?**

# How large is $C_{SS}$ ?

- ▶  $N = 5 \times 10^9$  for  $10^9$  sources (so  $\sim 10^{19}$  elements).
- ▶ Takes  $\sim 10^8$  TeraByte of space!
- ▶ Computing directly from normal matrix is unfeasible (Bombrun et al. 2011).

$$\vec{C}_{SS} = \text{Cov}(\vec{x})$$

$N \times N$

$$\begin{matrix} & x_1 & & x_N \\ x_1 & & & \\ & & & \\ & & & \\ x_N & & & \end{matrix}$$



Its **clearly** desirable that the covariance, between any pair of astrometric parameters, can be **computed** from a **reduced** amount of data.



How to compute  $(\vec{C}_{SS})_{ij}$  **accurately** and **efficiently**?



How do we actually estimate  
astrometric parameters?

**Formula alert!**  
**(stay right for easy passing...)**

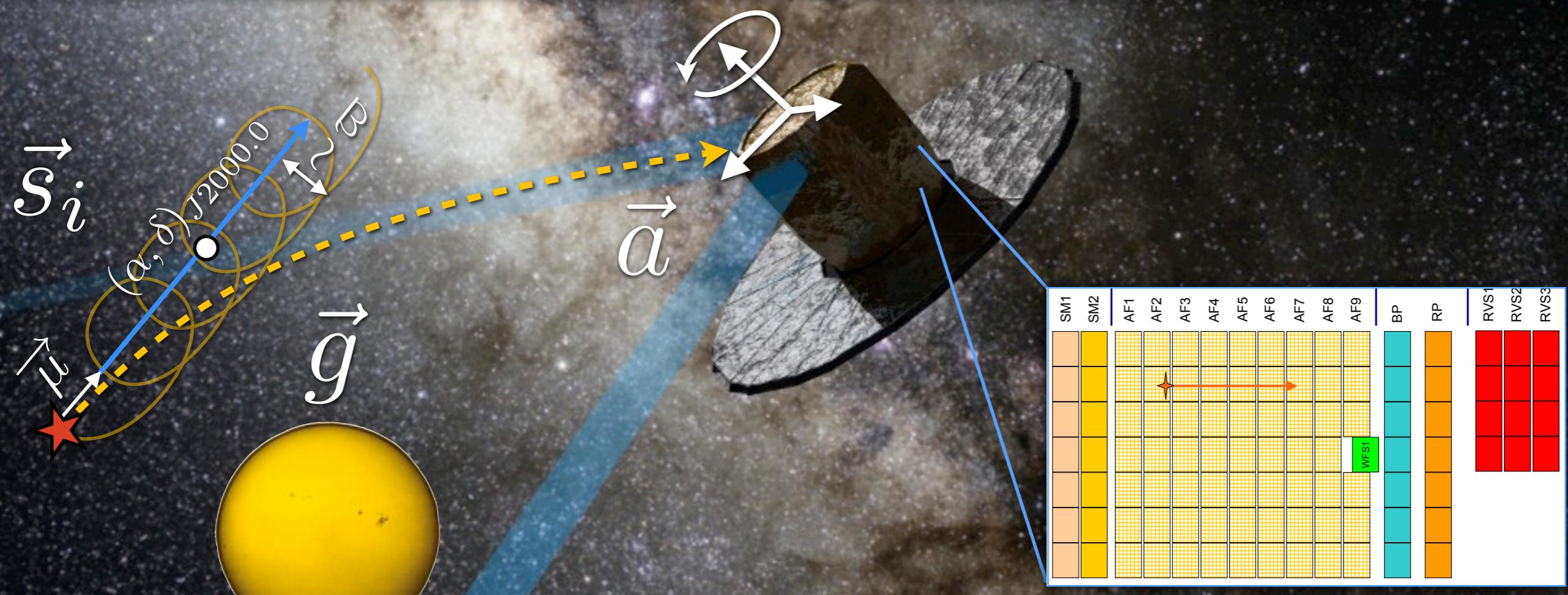


How to compute  $(\vec{C}_{SS})_{ij}$  **accurately** and **efficiently**?

# Astrometric parameters

'Nuisance' parameters (from astrometric point of view):

- ▶  $\vec{a}$  attitude
- ▶  $\vec{c}$  calibration
- ▶  $\vec{g}$  global



# How to estimate astrometric parameters?

Least squares solution with  $\sim 10^9$  parameters!

$$\min_{\vec{s}, \vec{a}} \sum_l \left[ \frac{t_l - f_l(\vec{s}_l, \vec{a}, \vec{c}, \vec{g})}{\sigma_l} \right]^2$$

$t_l$  = measured observation time  $l$  (with uncertainty  $\sigma_l$  )

$f_l$  = modelled observation time  $l$

In this talk **we ignore**  $\vec{c}$  and  $\vec{g}$  since they are affected by large number of observations over large part of the sphere.

A direct solution is unfeasible (Bombrun et al. 2011),  
therefore solved iteratively using **AGIS** (Lindegren et al. 2011):  
**Astrometric Global Iterative Solution**

# Normal matrix

Least-squares estimation of  $\vec{s}$  and  $\vec{a}$

$$\forall l : \vec{S}_l \Delta \vec{s}_i + \vec{A}_l \Delta \vec{a} = h_l$$

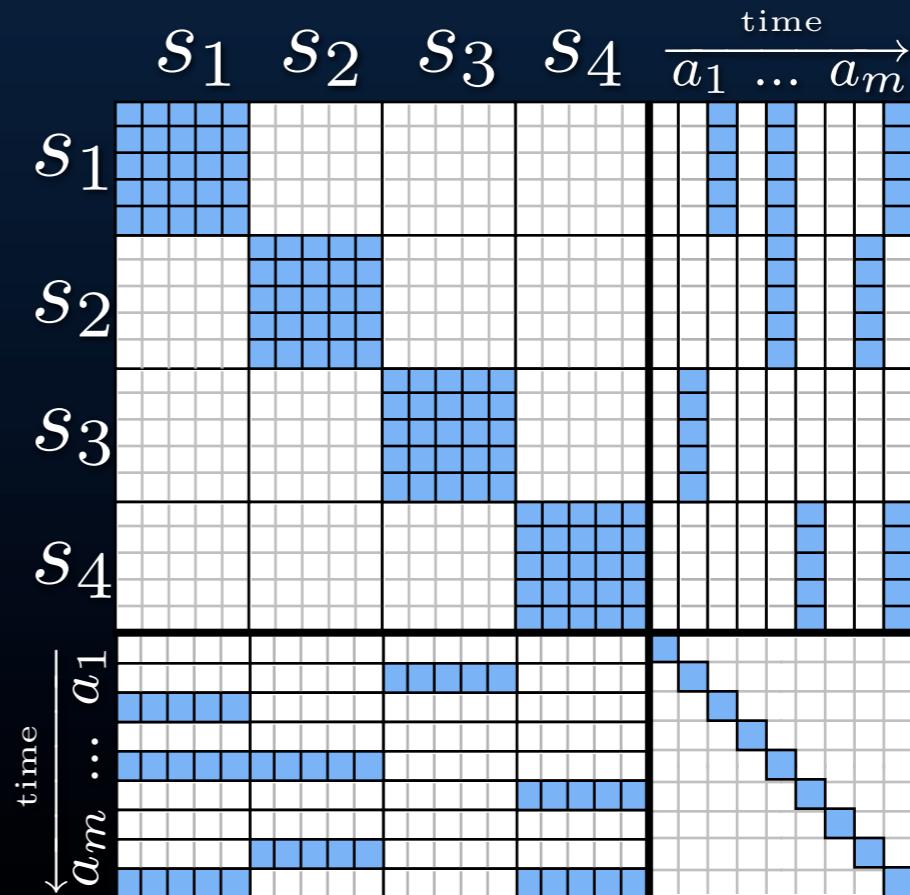
$$\vec{S}_l = (\partial f_l / \partial \vec{s}_i) / \sigma_l$$

$$\vec{A}_l = (\partial f_l / \partial \vec{a}) / \sigma_l$$

$$h_l = (t_l - f_l(\vec{s}_i, \vec{a})) / \sigma_l$$

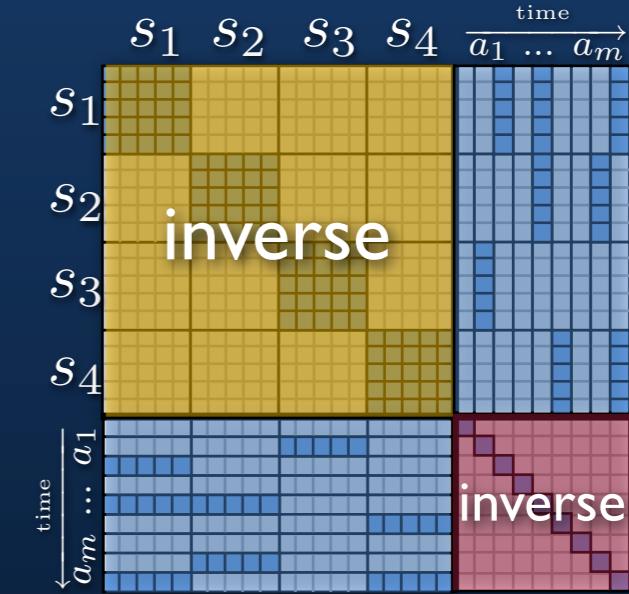
$$\vec{N} = \begin{bmatrix} \vec{N}_{SS} & \vec{N}_{SA} \\ \vec{N}_{AS} & \vec{N}_{AA} \end{bmatrix} =$$

**observation equations:**  
for each observations,  
how to modify parameters  
to get zero residual between  
measured and modelled time.



# Source parameter covariance $C_{SS}$

normal  $\vec{N} = \begin{bmatrix} \vec{N}_{SS} & \vec{N}_{SA} \\ \vec{N}_{AS} & \vec{N}_{AA} \end{bmatrix} =$



covariance  $\vec{C} = \begin{bmatrix} \vec{C}_{SS} & \vec{C}_{SA} \\ \vec{C}_{AS} & \vec{C}_{AA} \end{bmatrix} = \vec{N}^{-1}$  **unfeasible!**

Expansion of source parameter covariance\*:

$$\begin{aligned} \vec{C}_{SS} = & \vec{N}_{SS}^{-1} + \\ & \vec{N}_{SS}^{-1} \vec{N}_{SA} \left( \vec{N}_{AA}^{-1} \right) \vec{N}_{AS} \vec{N}_{SS}^{-1} + \\ & \vec{N}_{SS}^{-1} \vec{N}_{SA} \left( \vec{N}_{AA}^{-1} \vec{N}_{AS} \left( \vec{N}_{SS}^{-1} \right) \vec{N}_{SA} \vec{N}_{AA}^{-1} \right) \vec{N}_{AS} \vec{N}_{SS}^{-1} + \\ & \dots \end{aligned}$$

$$= \vec{C}_{SS}^{(1)} + \vec{C}_{SS}^{(2)} + \vec{C}_{SS}^{(3)} + \dots + \vec{C}_{SS}^{(p)} + \dots$$

\*Holl et al. 2011, in prep.

# Source parameter covariance $C_{SS}$

$$(\vec{C}_{SS}^{(1)})_{ij}$$

Only estimating source parameters from observations  
 = assuming attitude is known.  
 No source coupling, so zero for  $i \neq j$ .

$$(\vec{C}_{SS}^{(2)})_{ij}$$

Only estimating attitude parameters from observations  
 = and propagating those attitude cov back to source cov.  
 Sources get coupled to attitude parameters at which  
 they were observed.

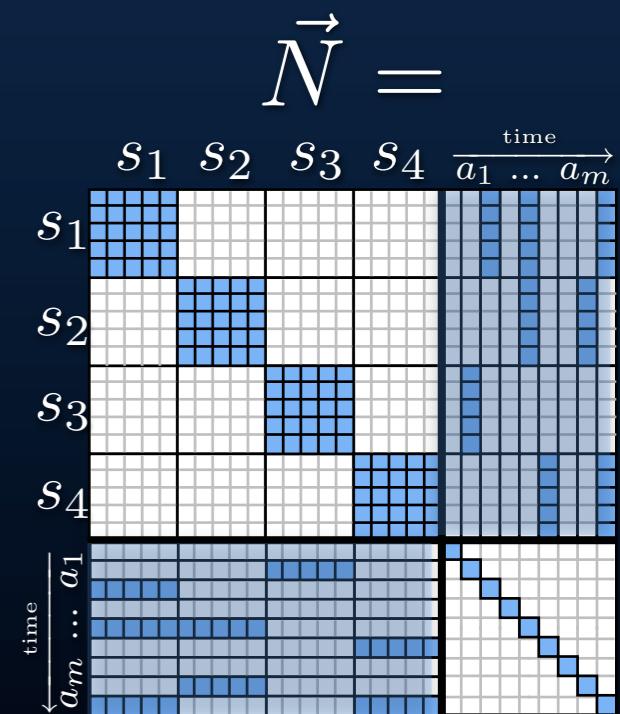
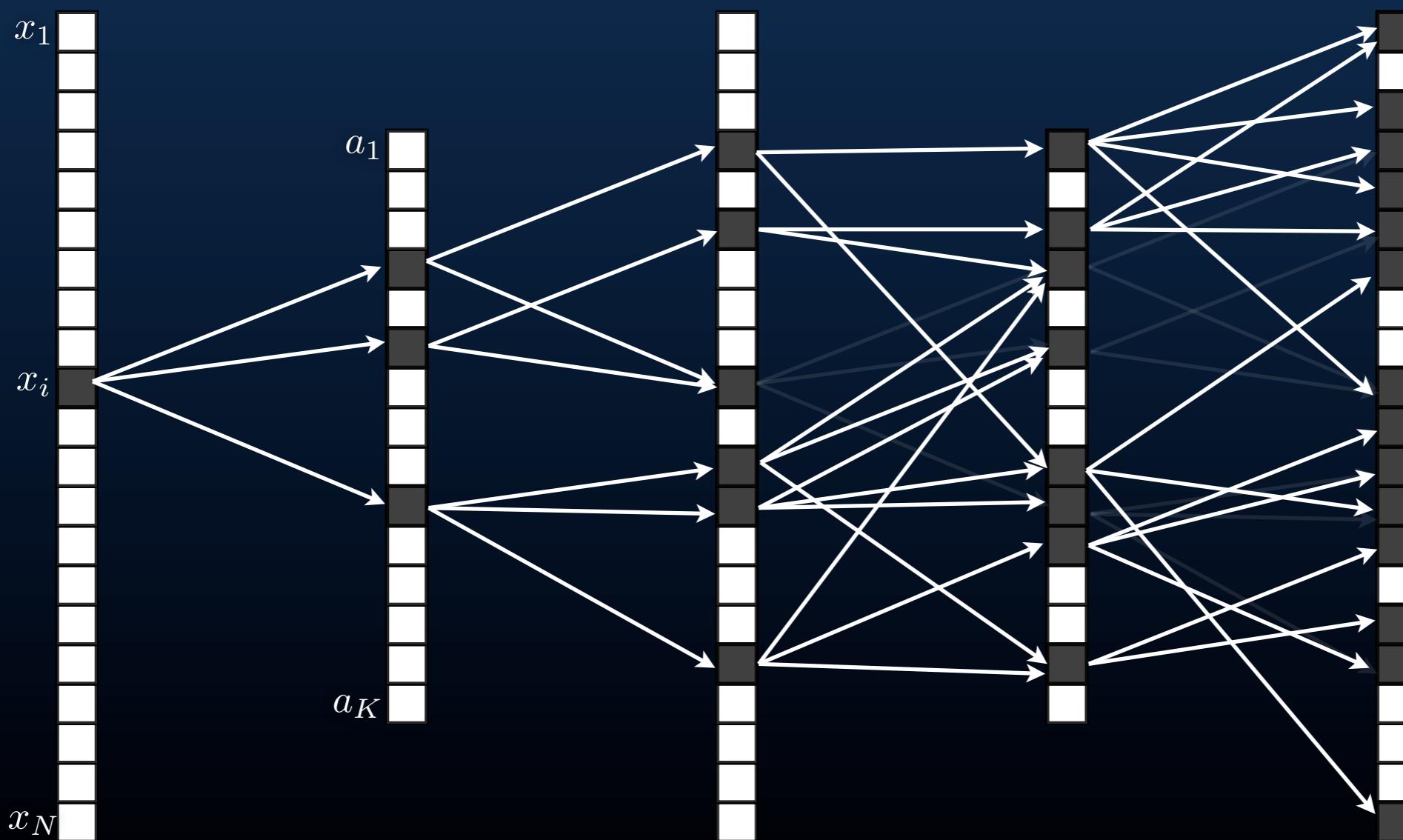
Expansion of source parameter covariance\*:

$$\vec{C}_{SS} = \vec{N}_{SS}^{-1} + \vec{N}_{SS}^{-1} \vec{N}_{SA} \left( \vec{N}_{AA}^{-1} \right) \vec{N}_{AS} \vec{N}_{SS}^{-1} + \vec{N}_{SS}^{-1} \vec{N}_{SA} \left( \vec{N}_{AA}^{-1} \vec{N}_{AS} \left( \vec{N}_{SS}^{-1} \right) \vec{N}_{SA} \vec{N}_{AA}^{-1} \right) \vec{N}_{AS} \vec{N}_{SS}^{-1} + \dots = \vec{C}_{SS}^{(1)} + \vec{C}_{SS}^{(2)} + \vec{C}_{SS}^{(3)} + \dots + \vec{C}_{SS}^{(p)} + \dots$$

\*Holl et al. 2011, in prep.

# Parameters connectivity

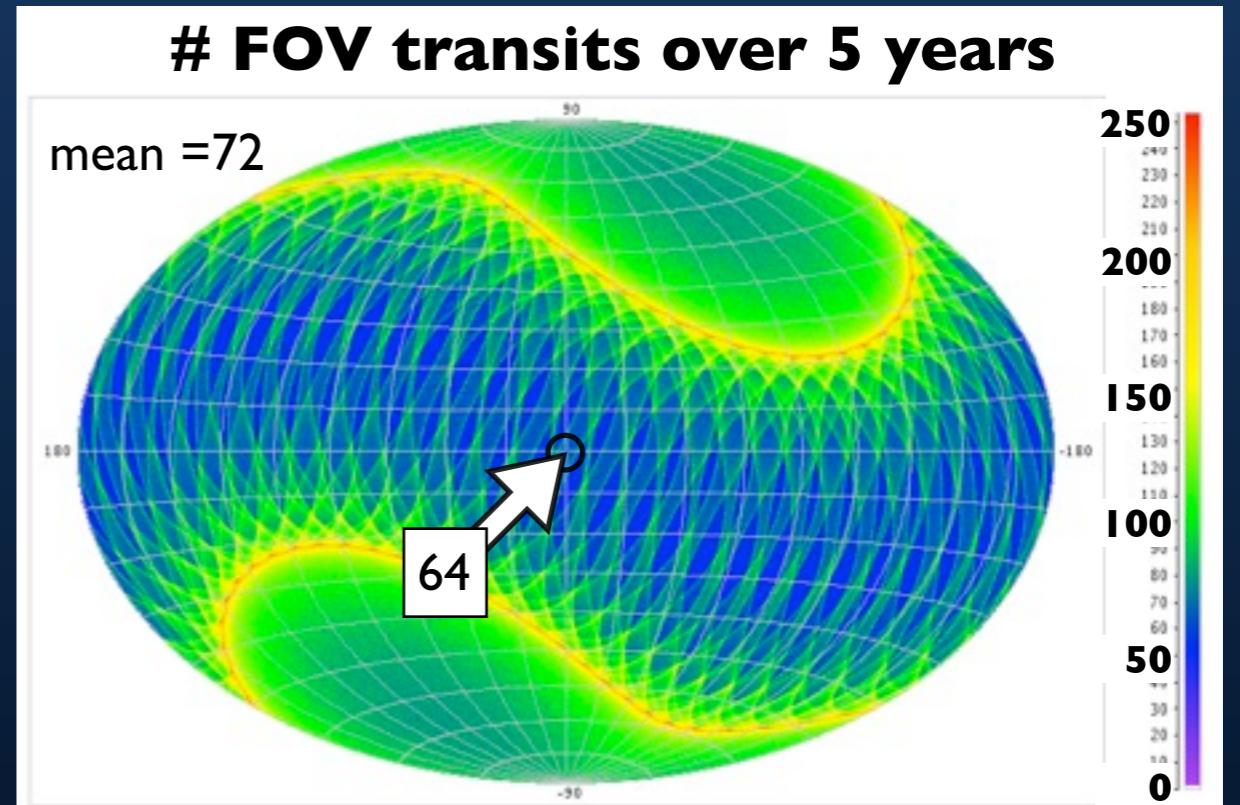
$$\begin{array}{c}
 (\vec{C}_{SS}^{(1)})_{ii} \\
 \xleftarrow{\hspace{1cm}} \quad \quad (\vec{C}_{SS}^{(2)})_{ii} \\
 \xleftarrow{\hspace{1cm}} \quad \quad (\vec{C}_{SS}^{(3)})_{ii} \\
 \xleftarrow{\hspace{1cm}} \quad \quad (\vec{C}_{SS}^{(4)})_{ii} \\
 \xleftarrow{\hspace{1cm}} \quad \quad (\vec{C}_{SS}^{(5)})_{ii}
 \end{array}$$



# Parameters connectivity study

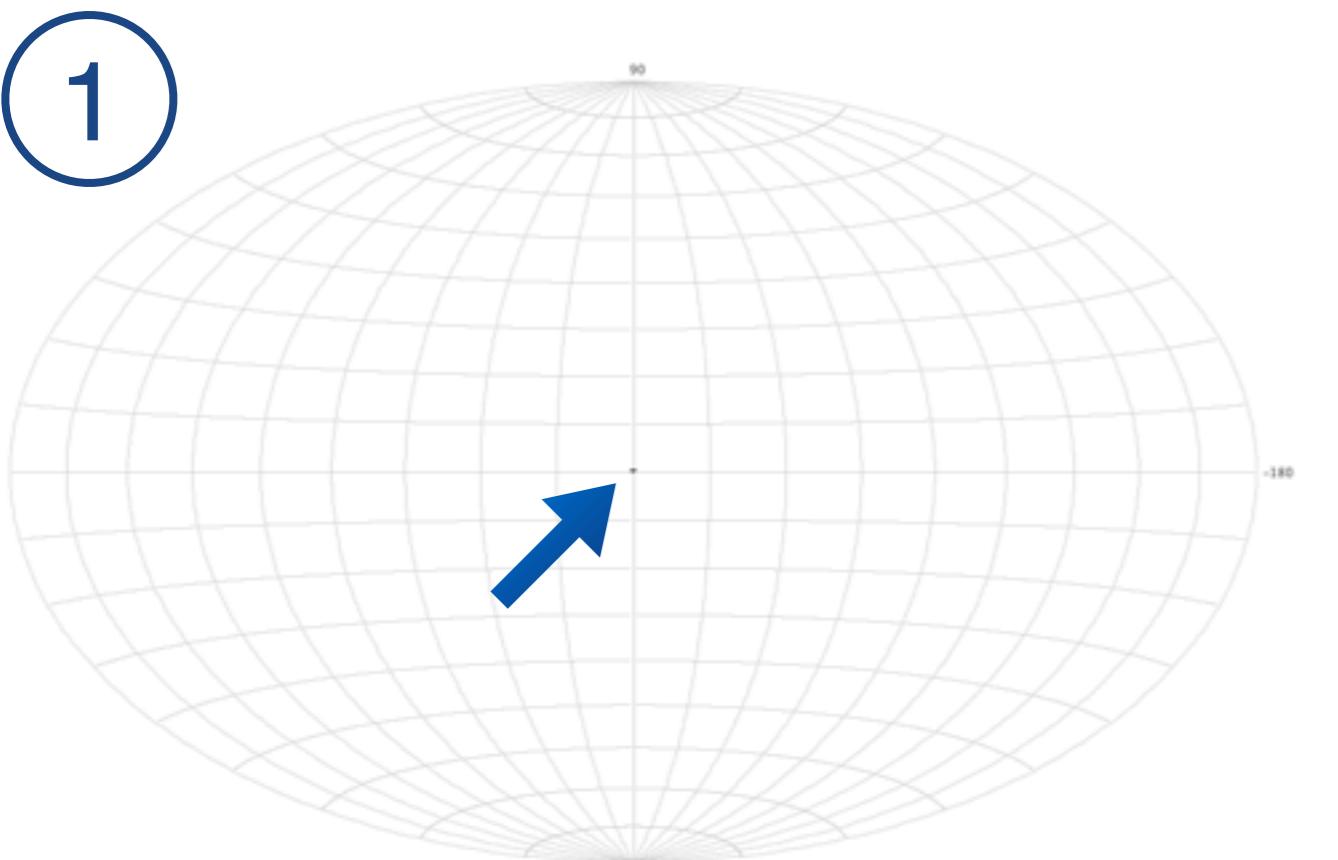
All connections for each term  $p$ :

- ▶  $(0^\circ, 0^\circ)$  ecliptic, 64 FOV transits.
- ▶ 196,608 sources,  $\sim 0.46^\circ$  separation  
(Gaia FOV  $\sim 0.7 \times 0.7^\circ$ ).
- ▶ 2,629,800 attitude intervals,  
60 sec per interval  
(Gaia FOV transit time  $\sim 45$  sec).
- ▶ Simulated 5 years observations.

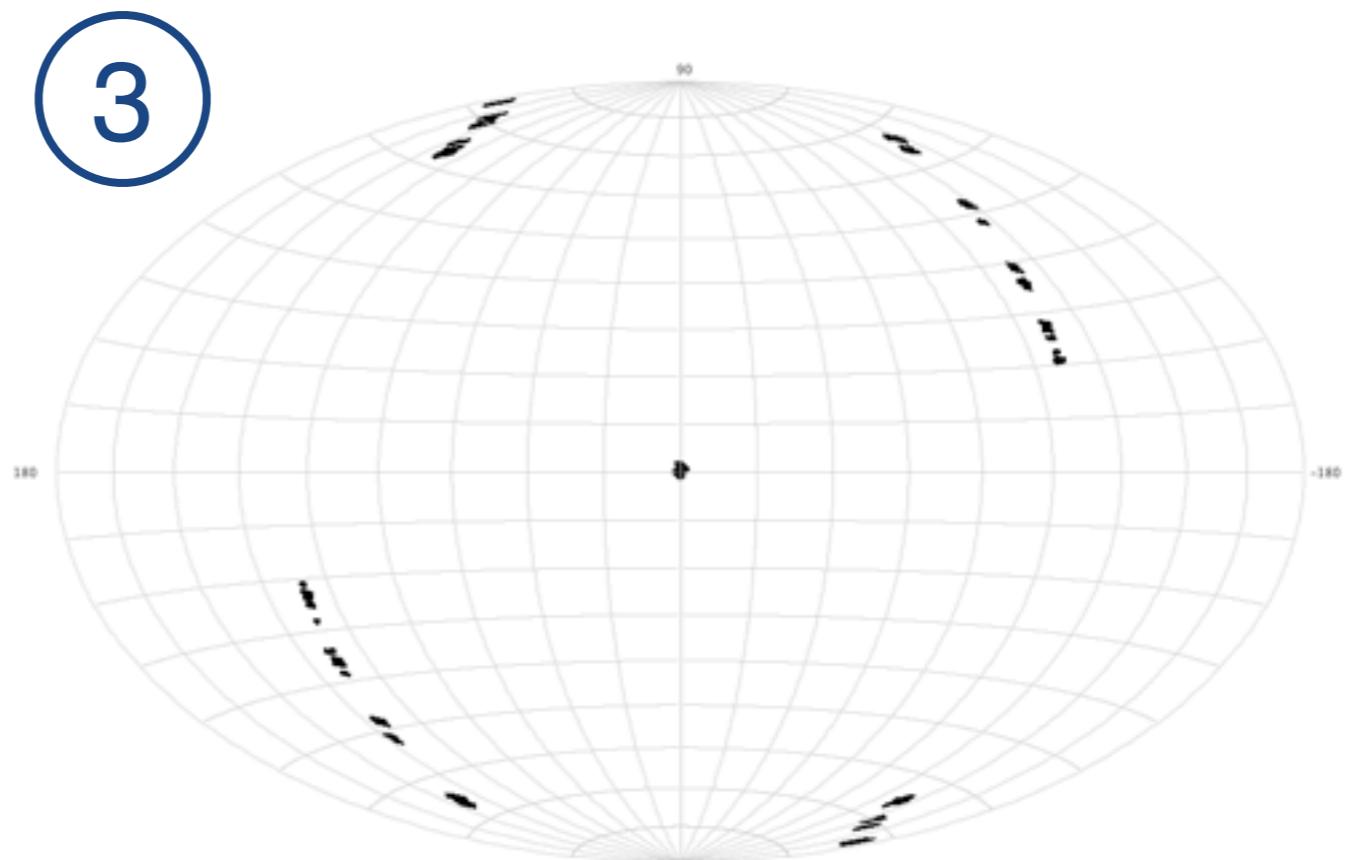


# Source connections to $(0^\circ, 0^\circ)$

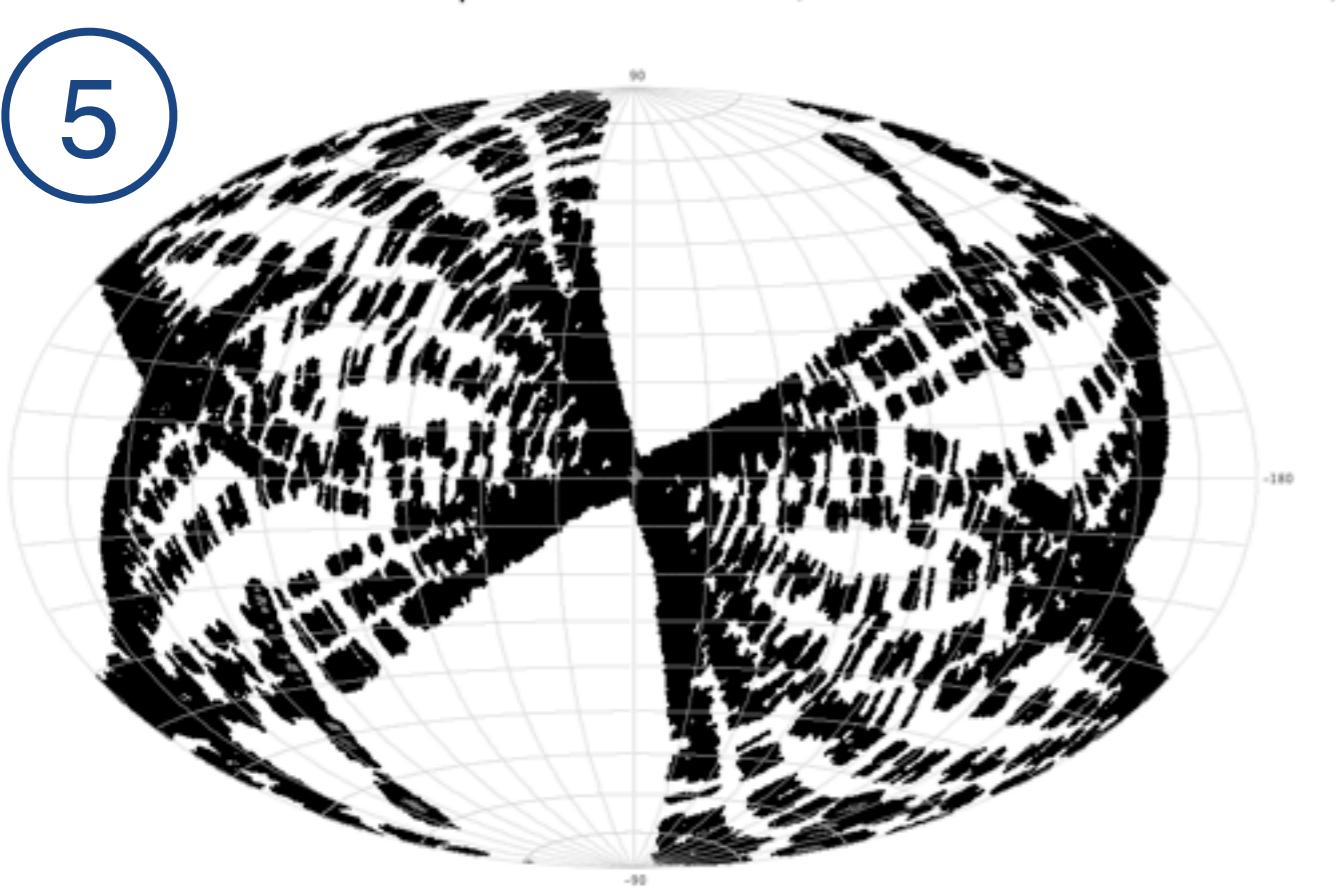
Connections Cov term 1 alpha: 0.0 delta: 0.0 ( 0.00% of total 196608 sources)



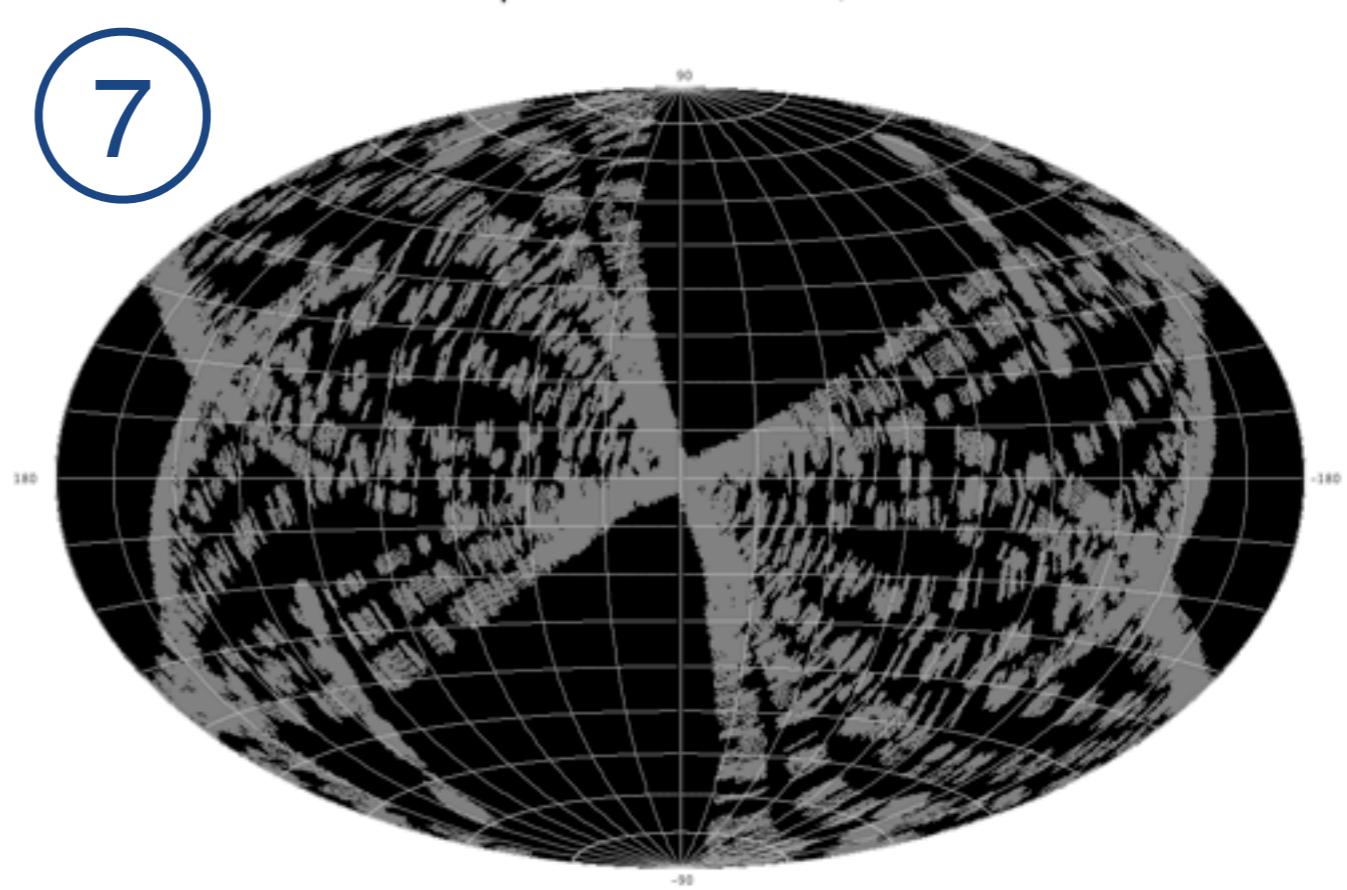
Connections Cov term 3 alpha: 0.0 delta: 0.0 ( 0.24% of total 196608 sources)



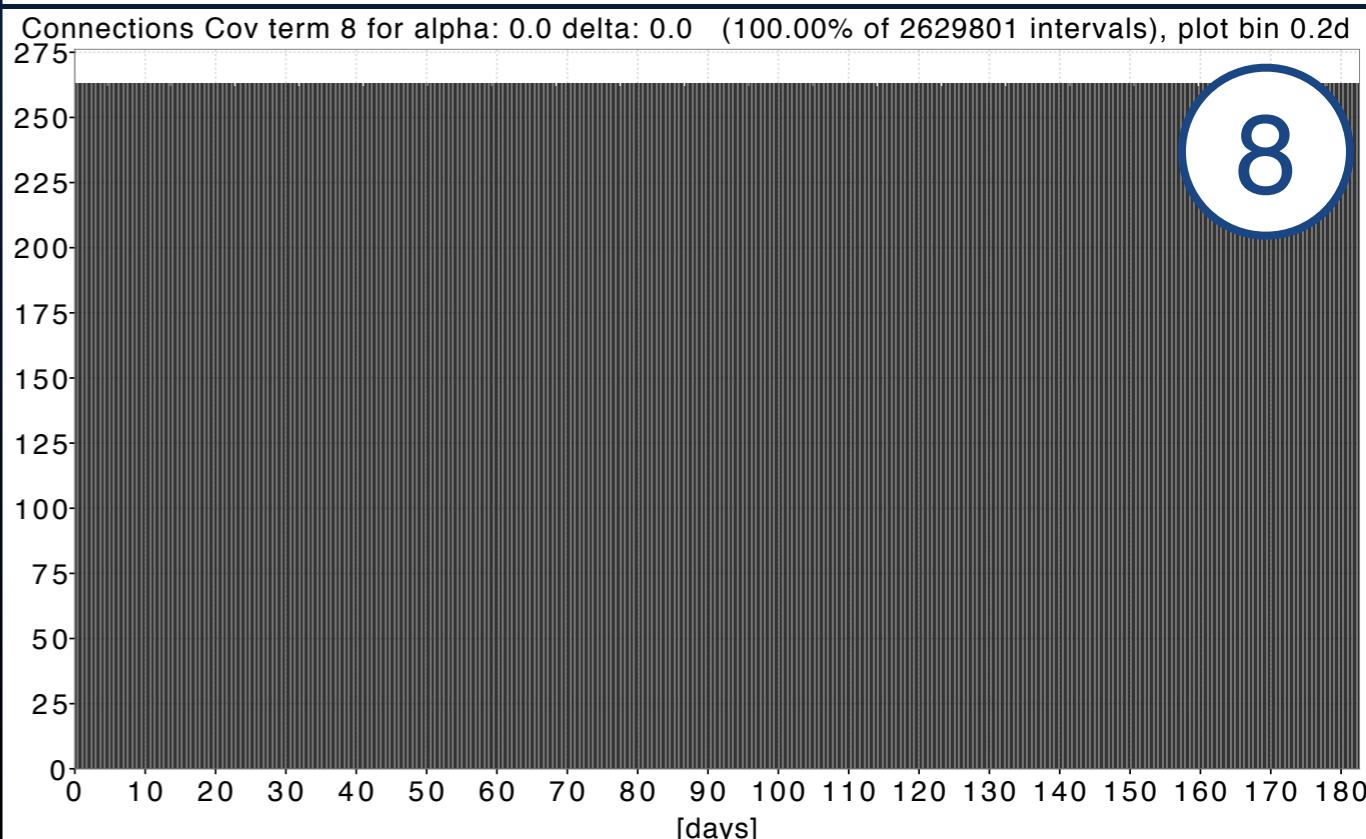
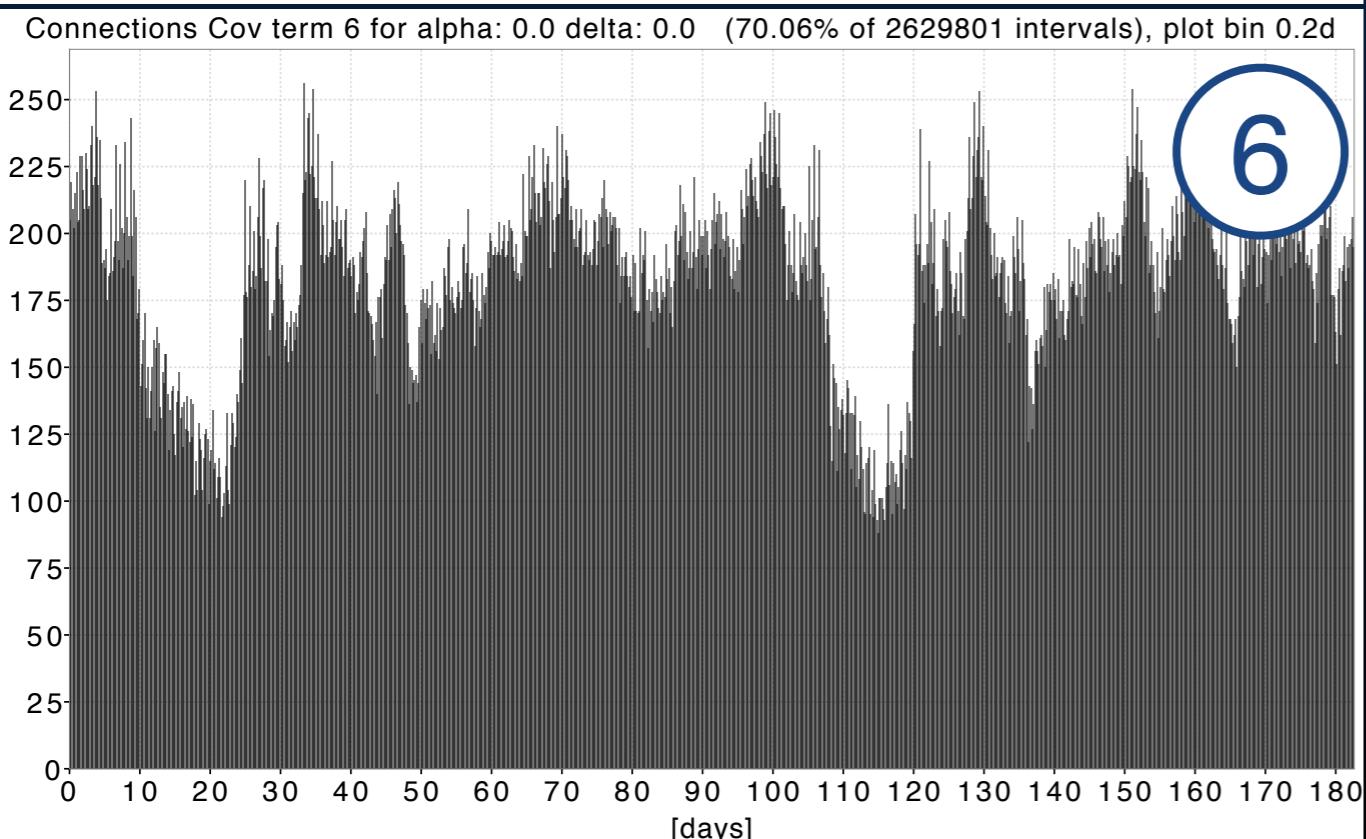
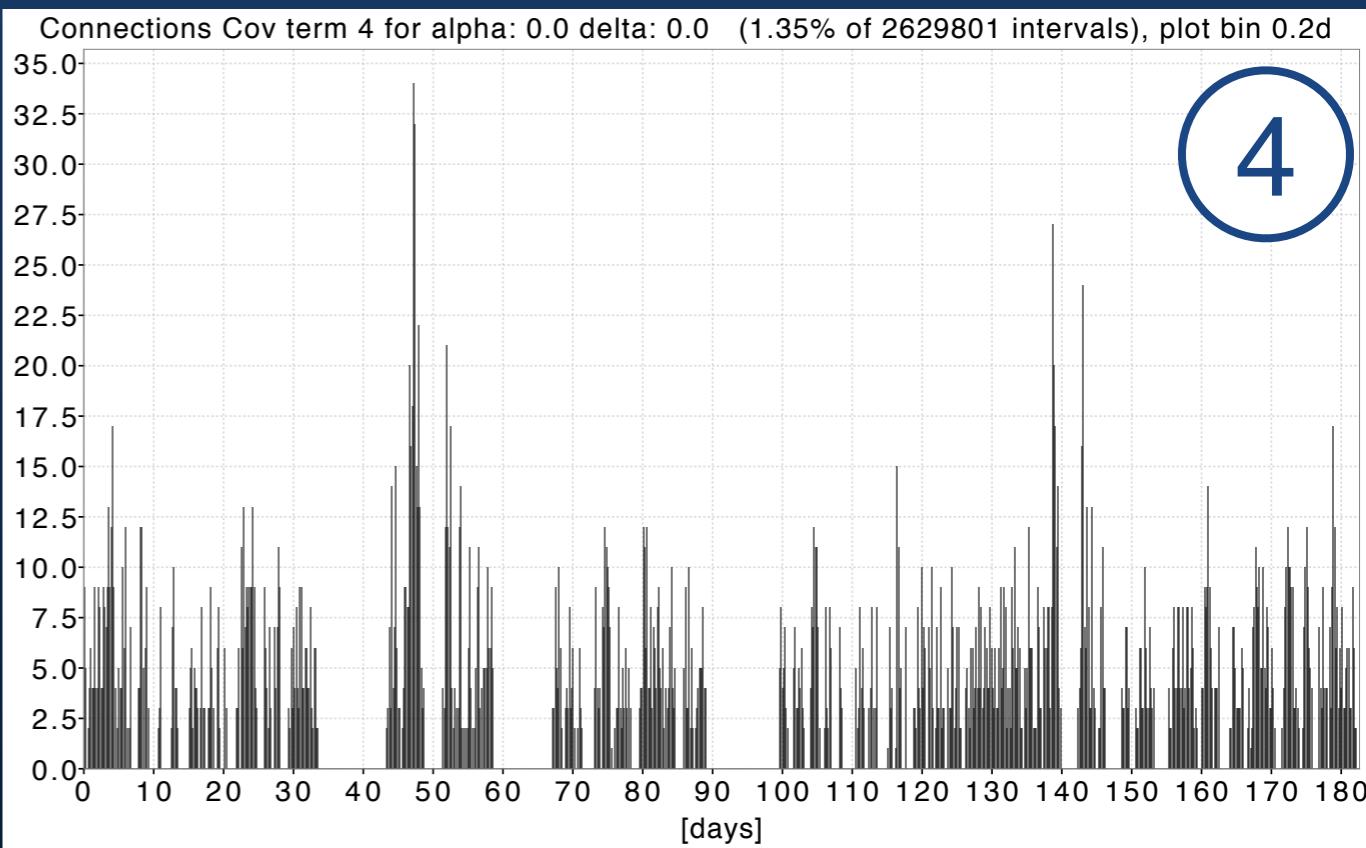
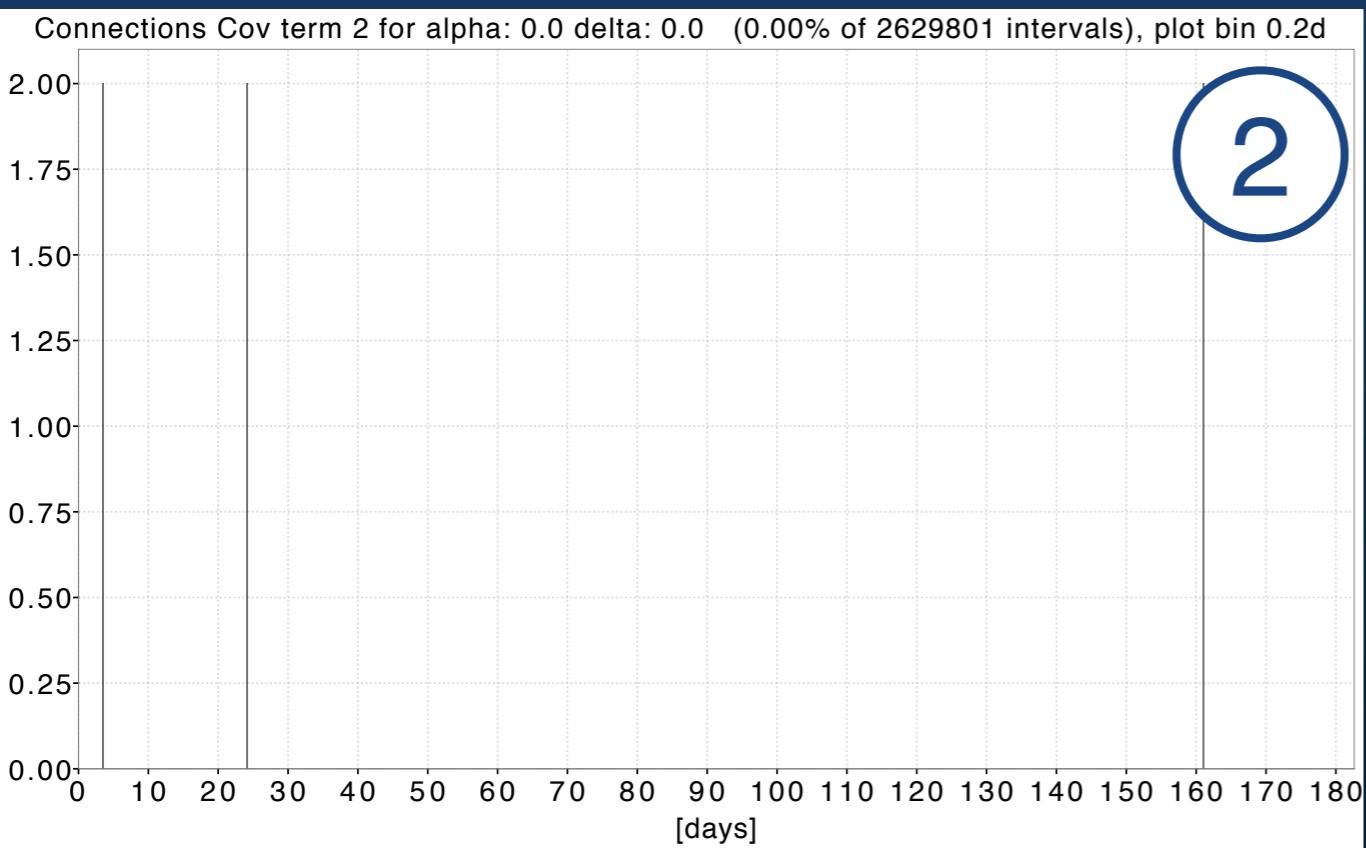
Connections Cov term 5 alpha: 0.0 delta: 0.0 ( 32.56% of total 196608 sources)



Connections Cov term 7 alpha: 0.0 delta: 0.0 (100.00% of total 196608 sources)



# Attitude connections to $(0^\circ, 0^\circ)$



plot bin size 0.18 days (containing 263 attitude intervals)

# Covariance expansion model

Input data:

- ▶ Per source, per field-of-view (FOV) transit:
  - ◆ partial derivatives of observations w.r.t. the (5) source parameters.
  - ◆ observation time
  - ◆ combined weight of the observations

On average 72 FOV transits  $\sim 2$  TB for  $10^9$  sources.

**$\sim 10^8$  times less than storing the full covariance matrix!**

# Conclusions

- ▶ No chance to provide the whole  $\vec{C}_{SS}$ .
- ▶ We found recursive algorithm to compute any element of  $\vec{C}_{SS}$  (based on reduced amount of structural data), to any accuracy at increasing computational cost.
- ▶ But in reality we want  $\text{Cov}(\vec{y})$ , which might involve a very large number of elements of  $\vec{C}_{SS}$ , making it impractical/unfeasible, so:
- ▶ If the user provides  $\vec{J}$ , apply recursive algorithm to compute  $\text{Cov}(\vec{y}) = \vec{J}\vec{C}_{SS}\vec{J}'$  directly, without explicitly computing any part of  $\vec{C}_{SS}$  (is possible).

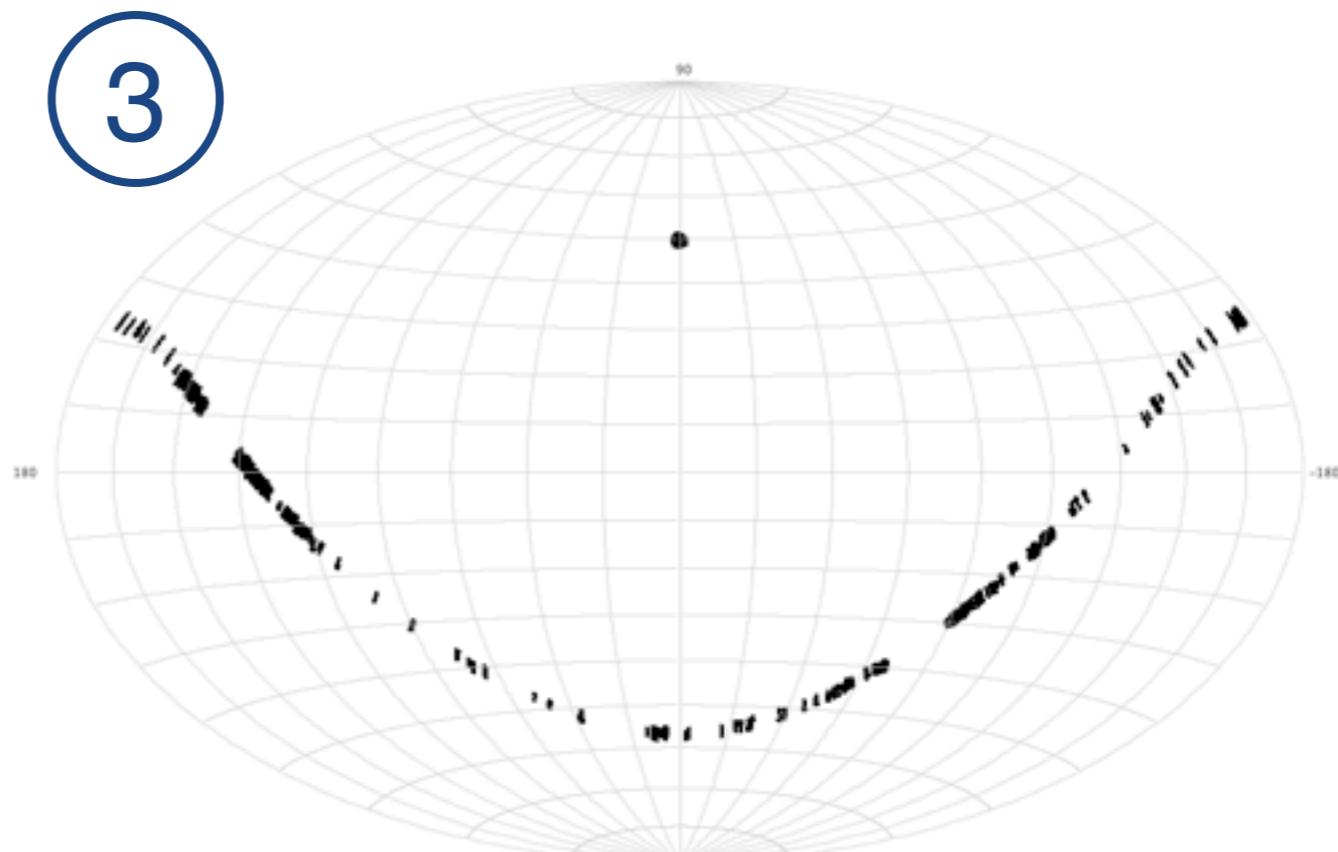
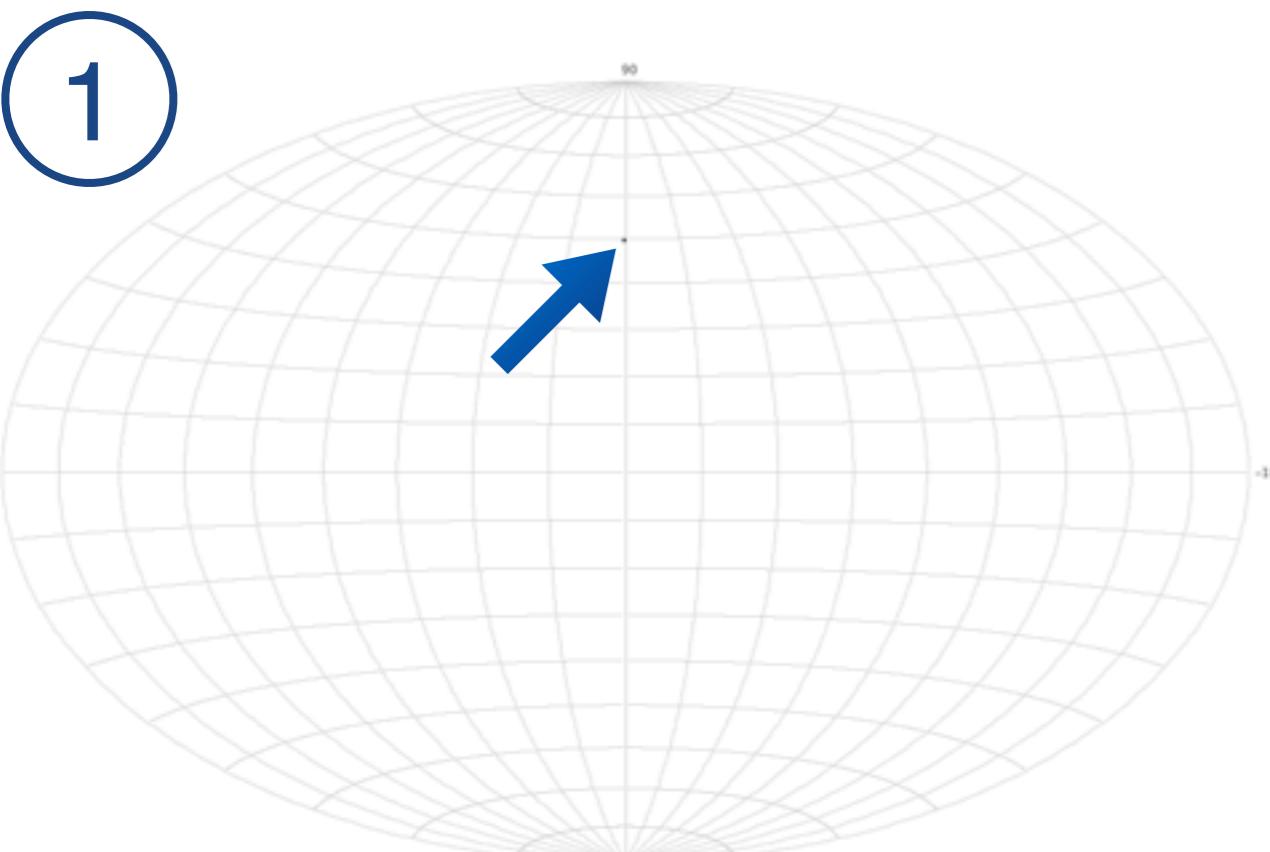
# Contact details

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- ▶ Lund Observatory  
Telephone: +46 46 22 21567

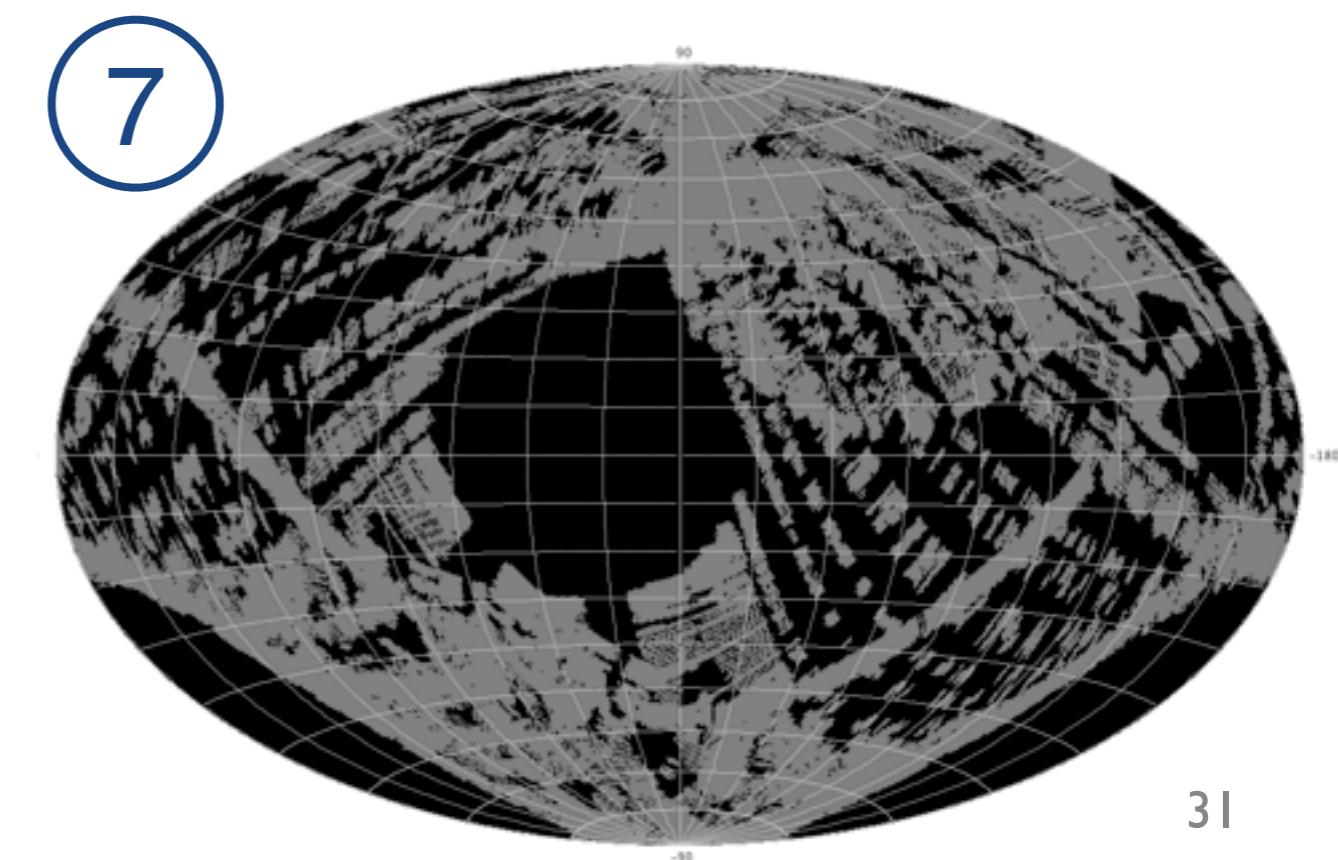
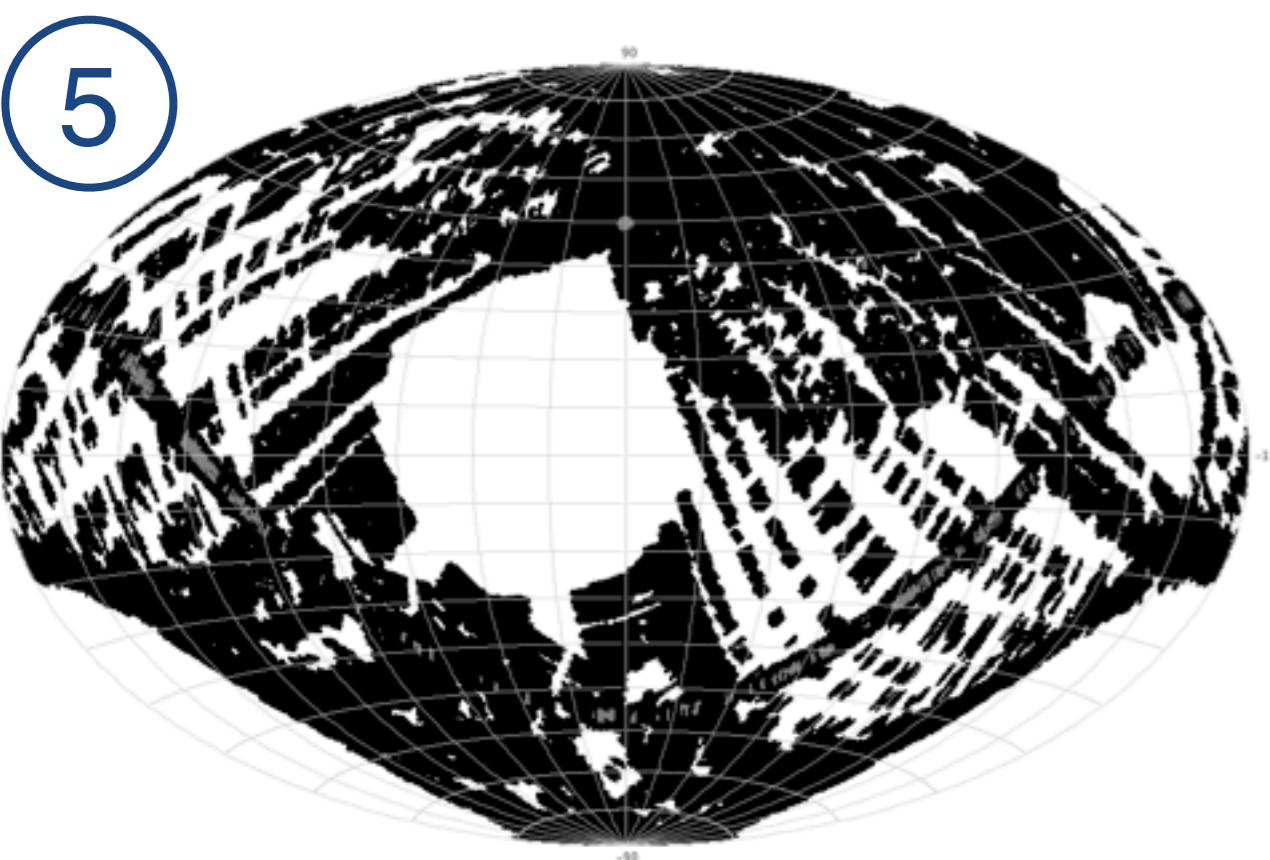
# Appendices

# Source connections to $(0^\circ, 50^\circ)$

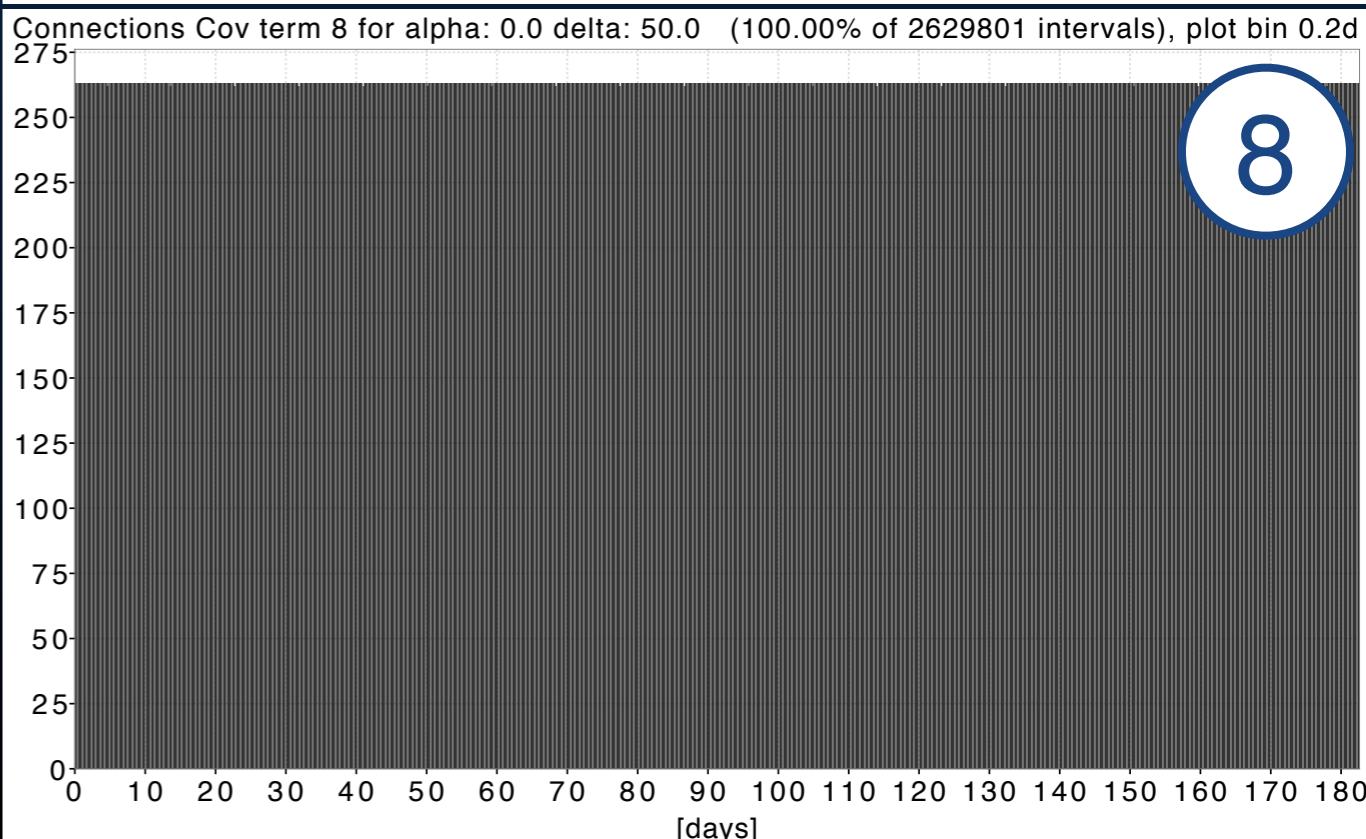
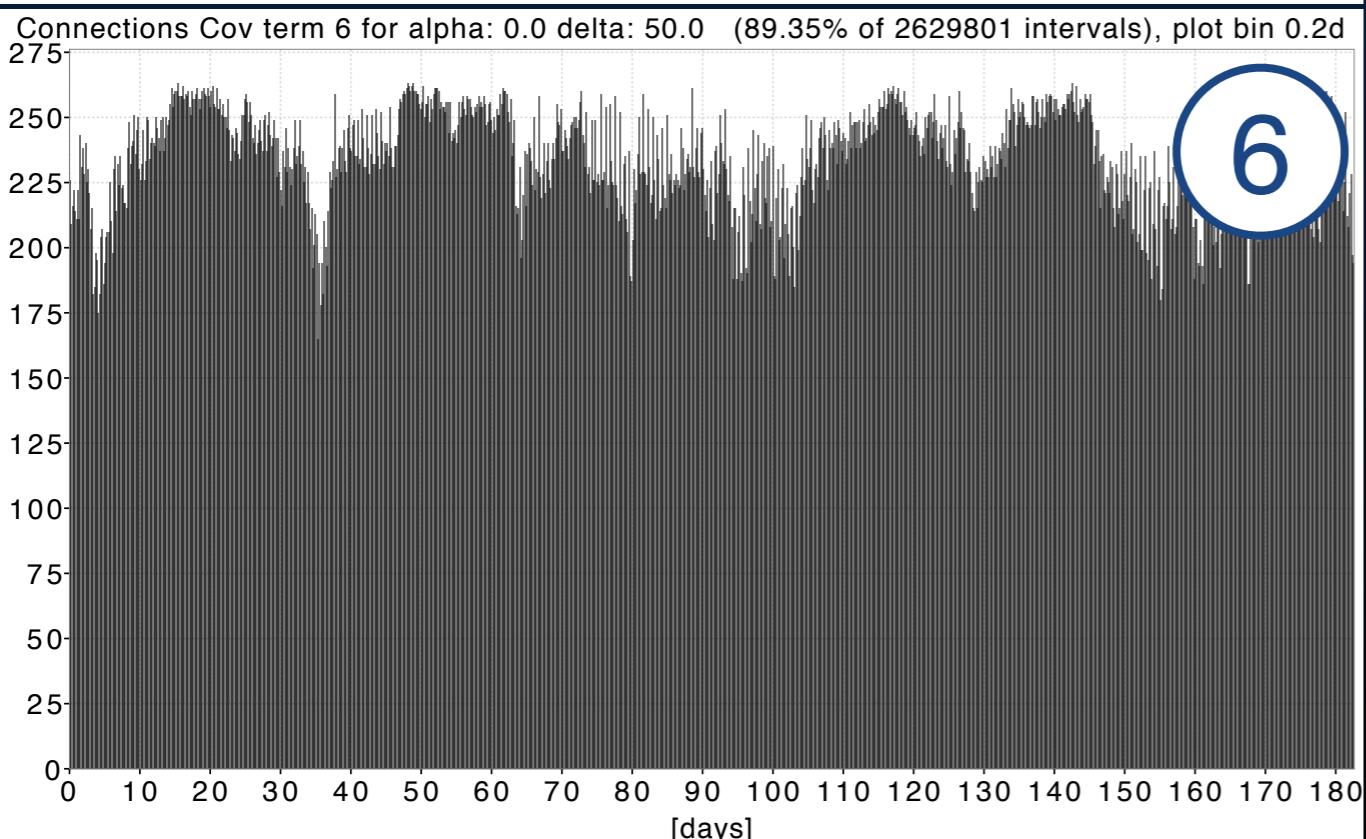
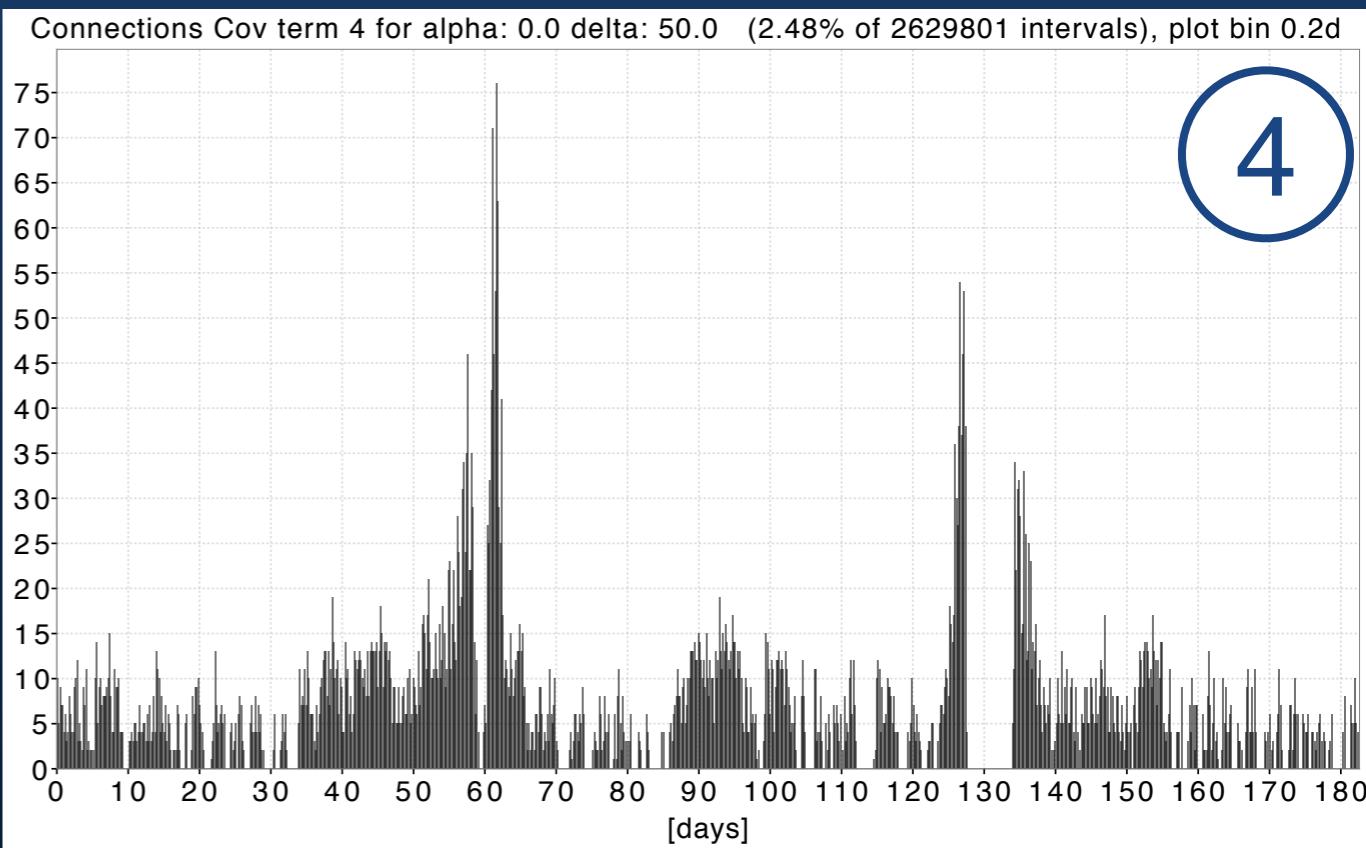
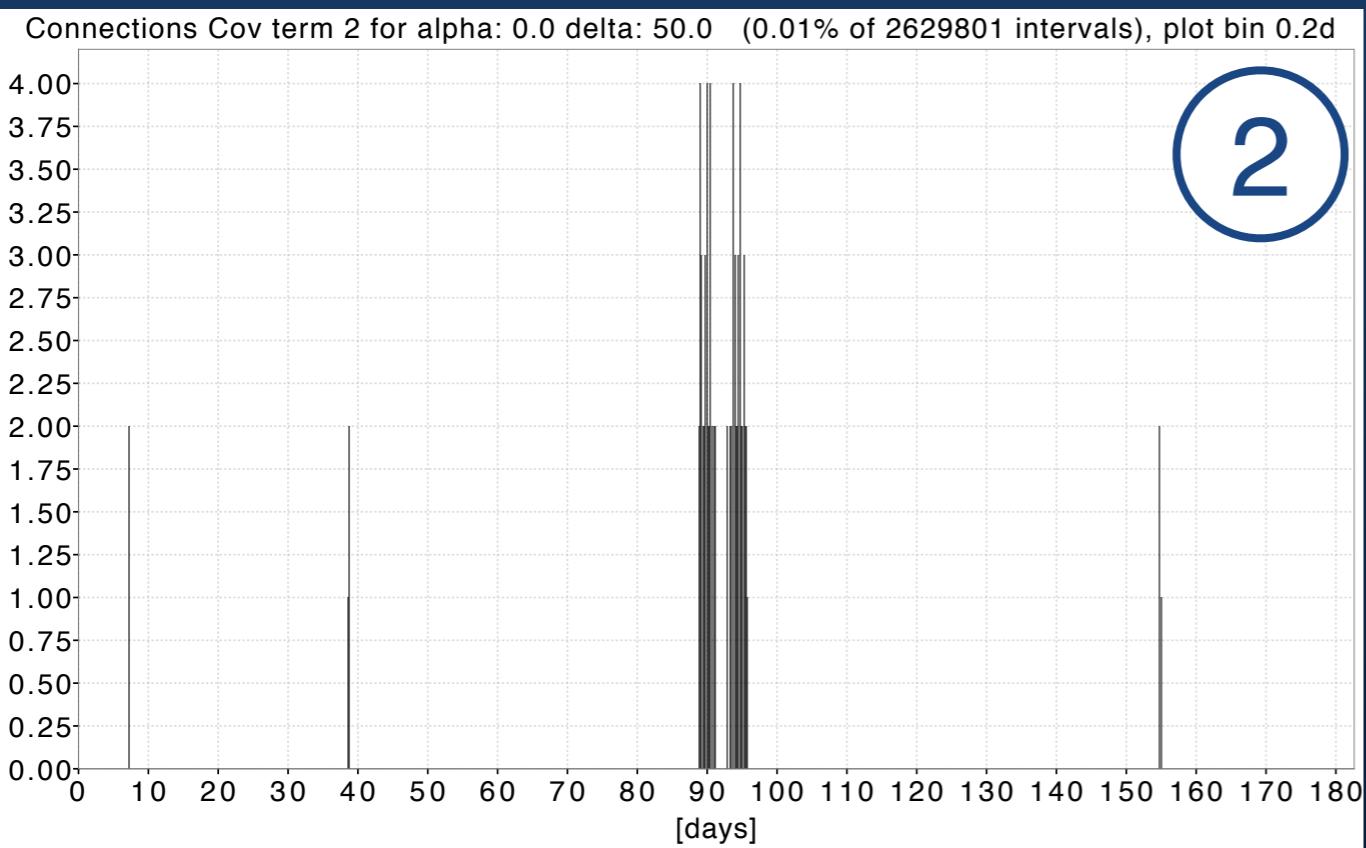
Connections Cov term 1 alpha: 0.0 delta: 50.0 ( 0.00% of total 196608 sources) Connections Cov term 3 alpha: 0.0 delta: 50.0 ( 0.66% of total 196608 sources)



Connections Cov term 5 alpha: 0.0 delta: 50.0 ( 49.50% of total 196608 sources) Connections Cov term 7 alpha: 0.0 delta: 50.0 (100.00% of total 196608 sources)



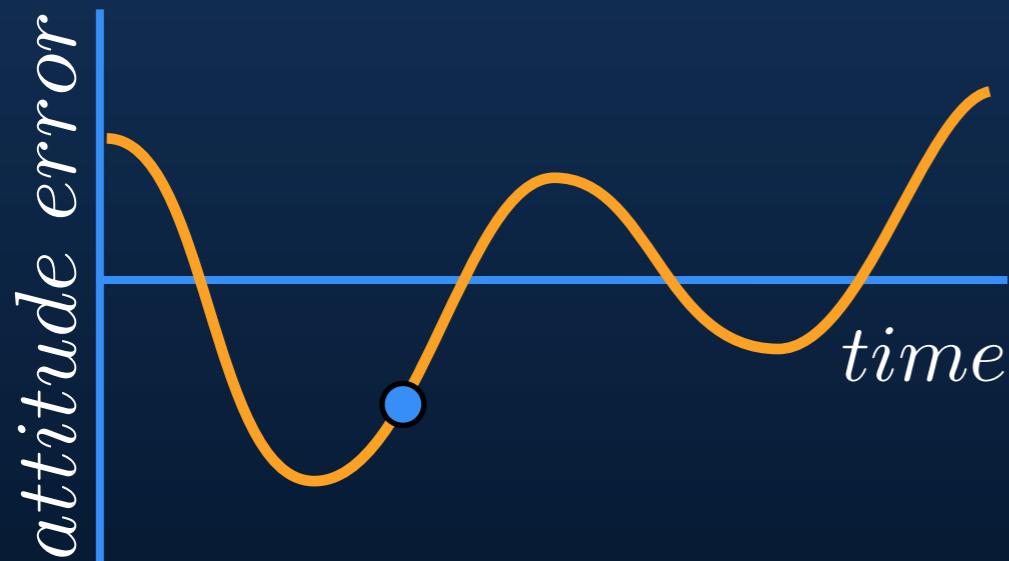
# Attitude connections to (0°, 50°)



plot bin size 0.18 days (containing 263 attitude intervals)

# Theory: attitude causes spatial correlations

Attitude is solved as part of solution,  
so it cannot be **exactly** determined!



For **attitude error at time  $t$** , all observations  
in the focal plane will be affected:

- ▶ Correlation between stars **in the FOV**.
- ▶ Correlation between stars **in the fields separated by the basic angle**.

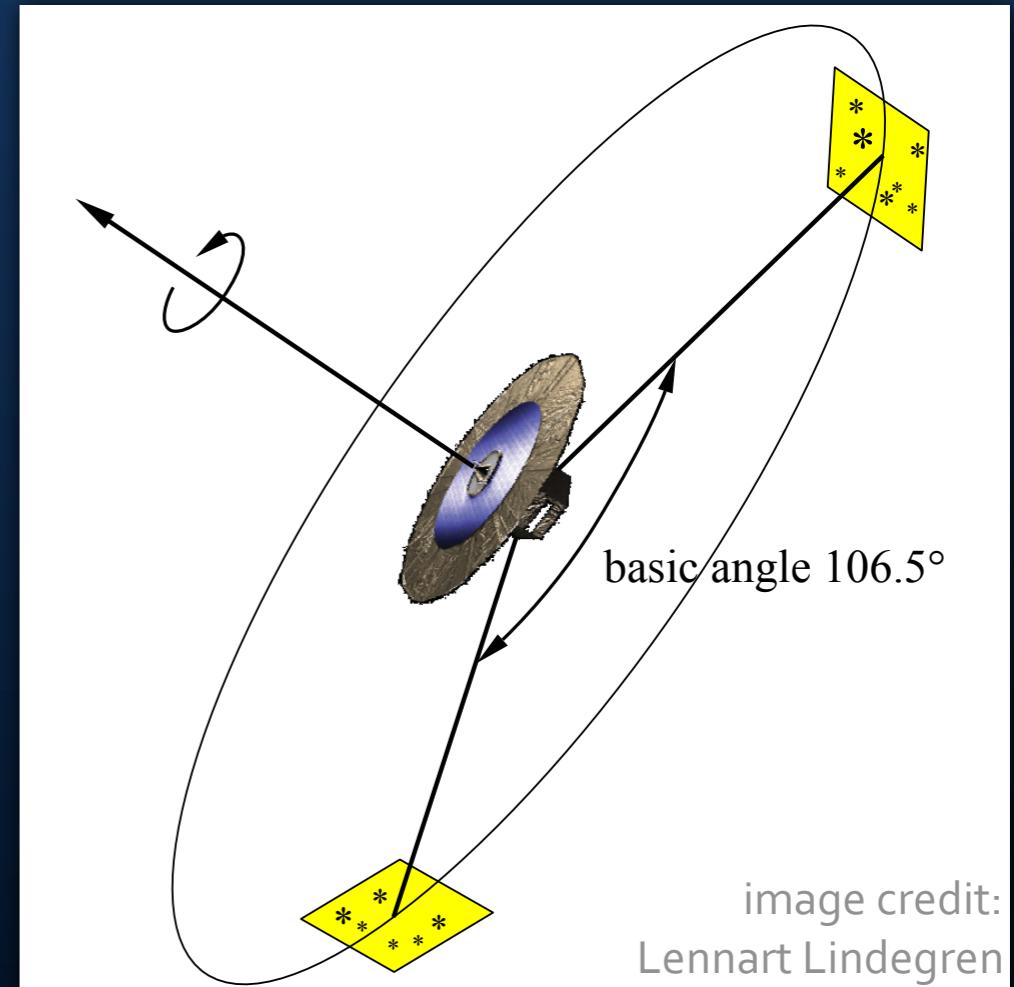
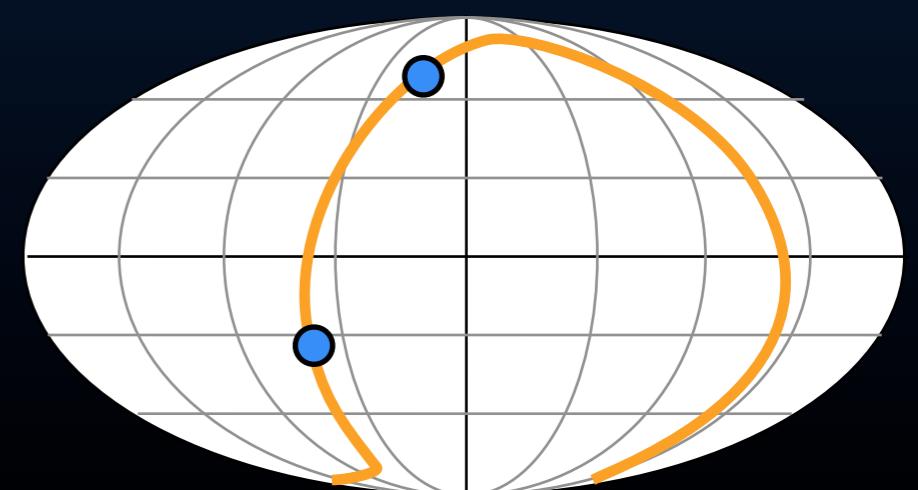


image credit:  
Lennart Lindegren



# Gaia scanning law

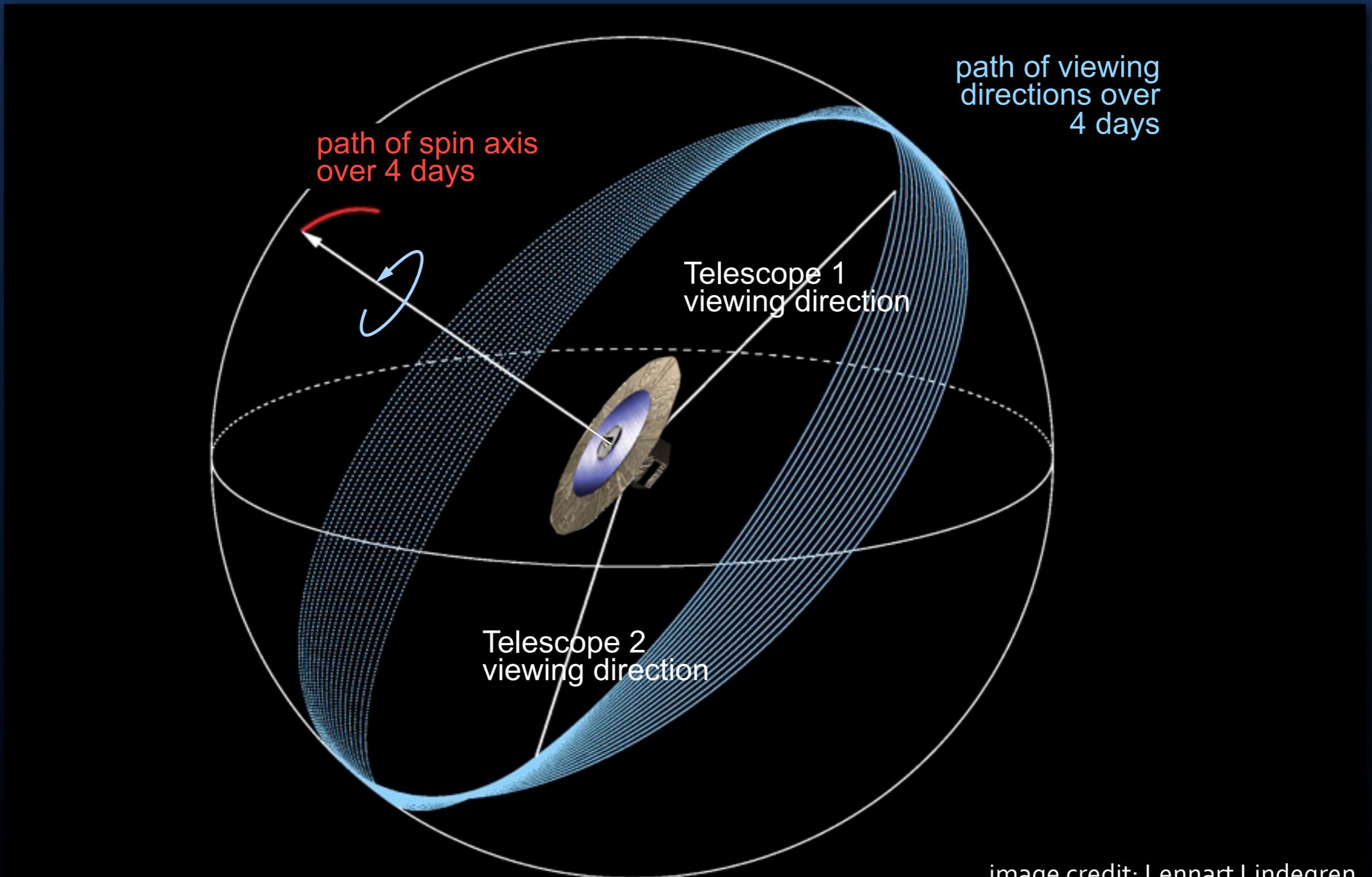
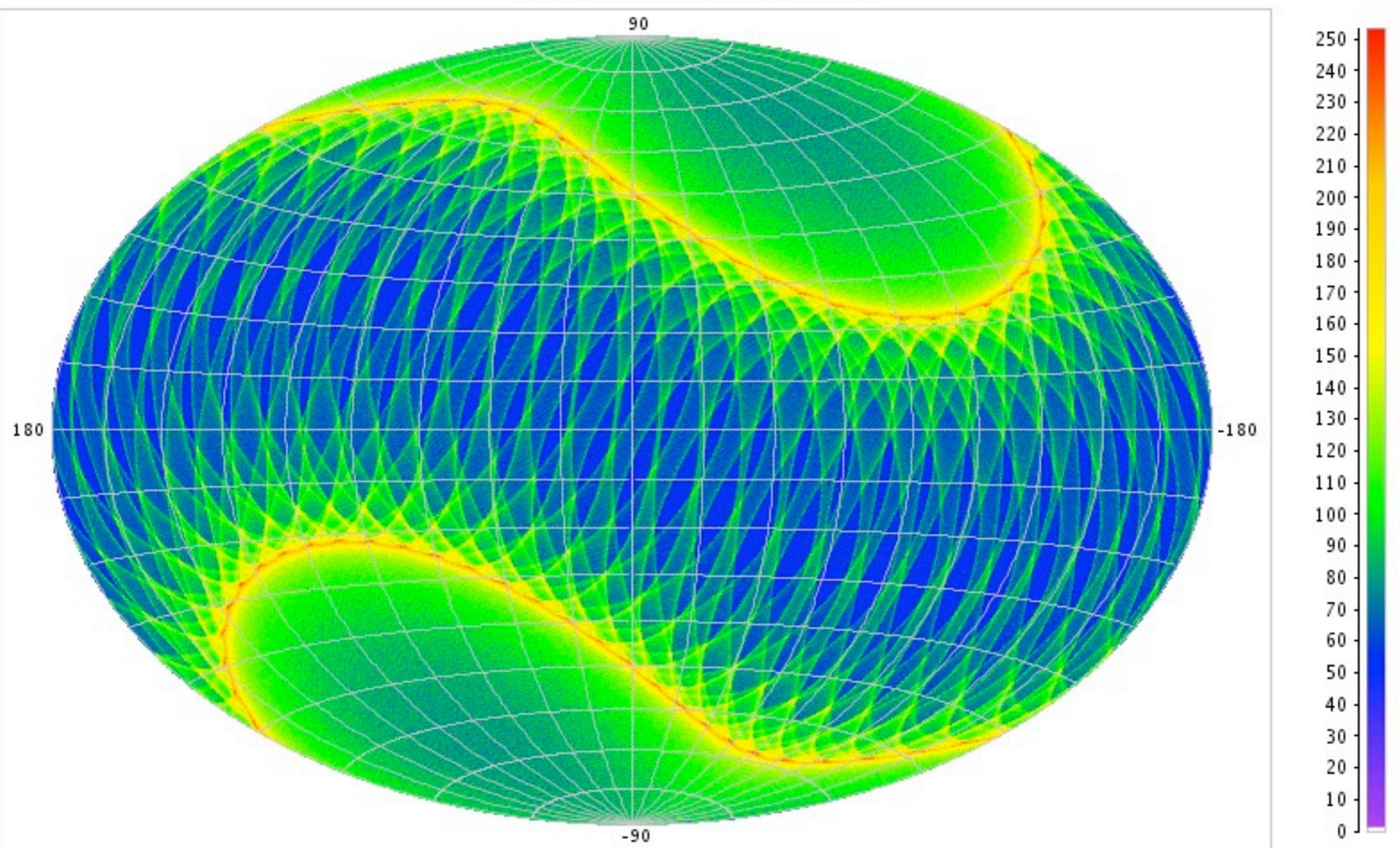


image credit: Lennart Lindegren

# Nominal Scanning Law

Gaia field transits (ICRS) for 5 years



# Covariance & correlation

$x_i$  and  $x_j$  are estimated values of astrometric parameter  $i$  and  $j$ .

These values are correlated if:

$$Cov[x_i, x_j] = E[e_i e_j] \neq 0 \quad \text{for } i \neq j$$

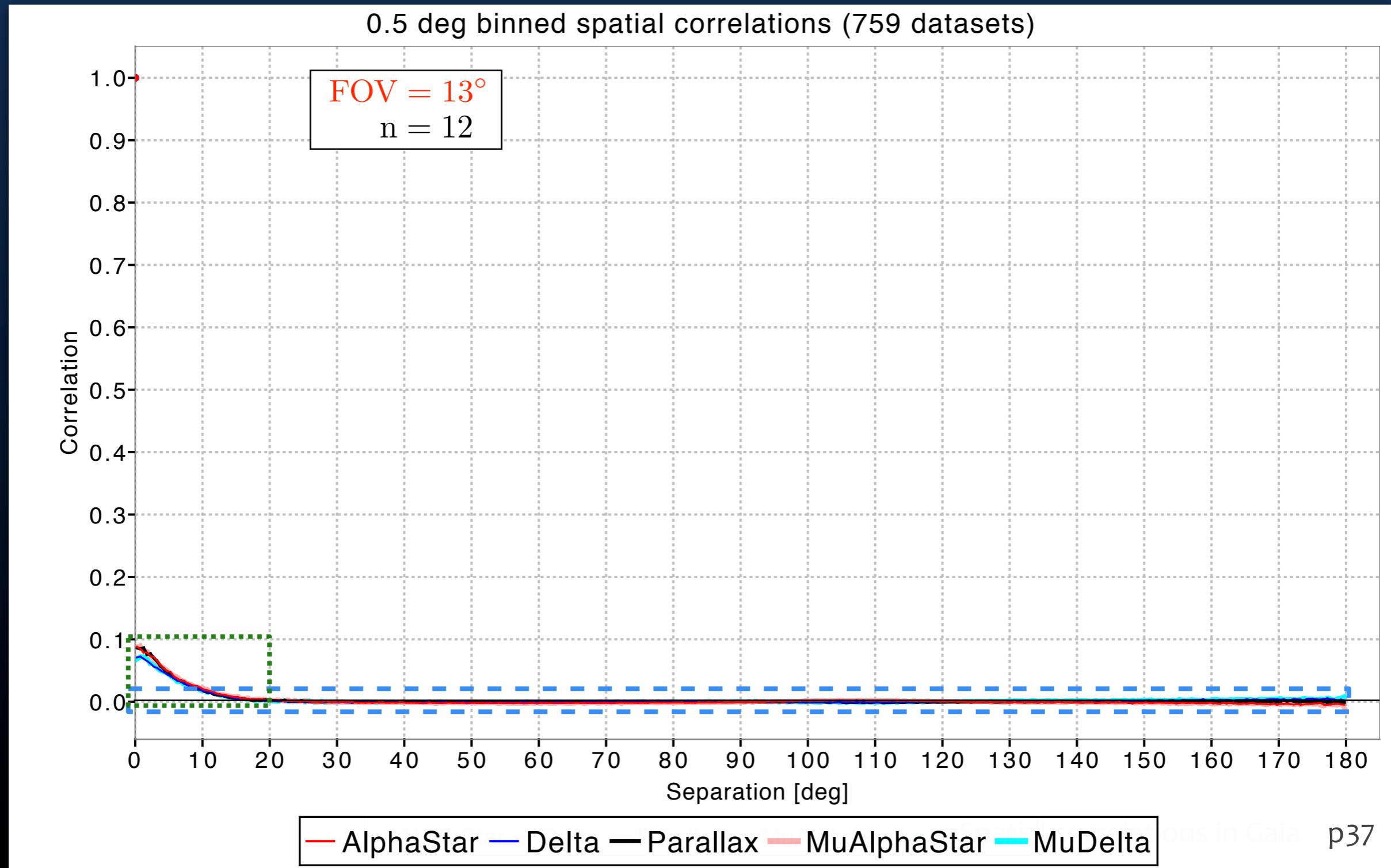
$$e_i = (x_i - x_{i, \text{true}}) \quad \text{the error}$$

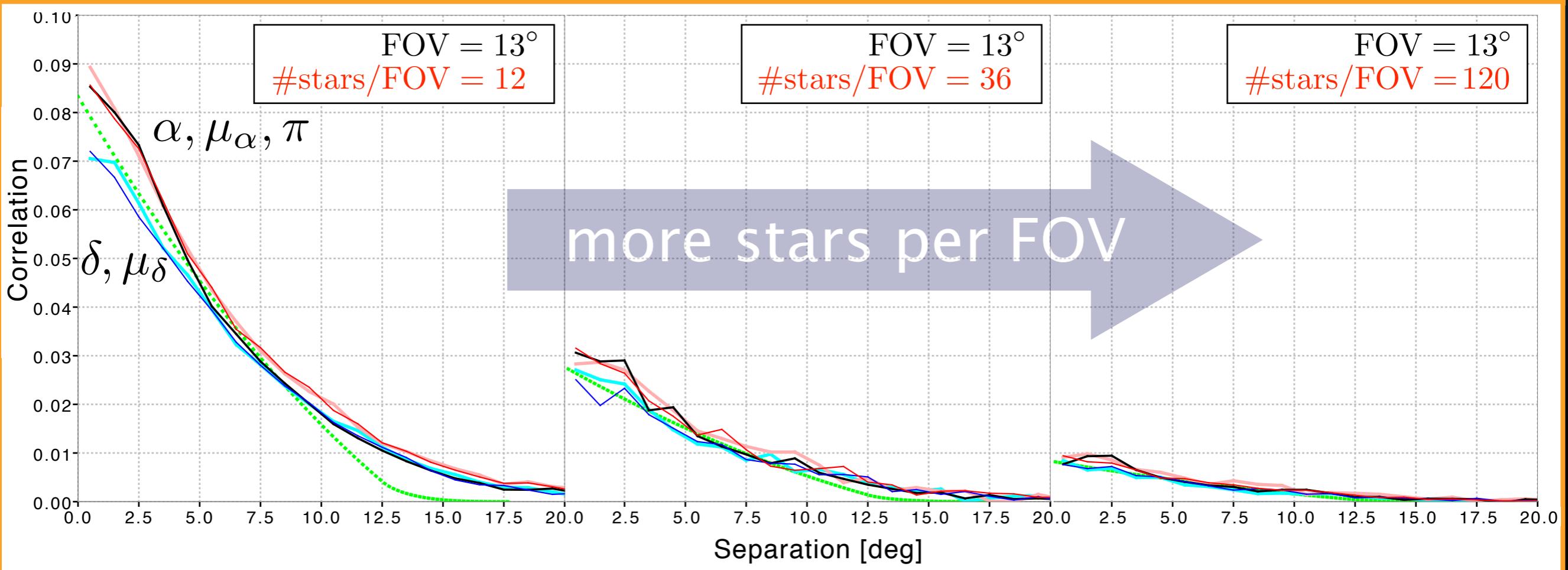
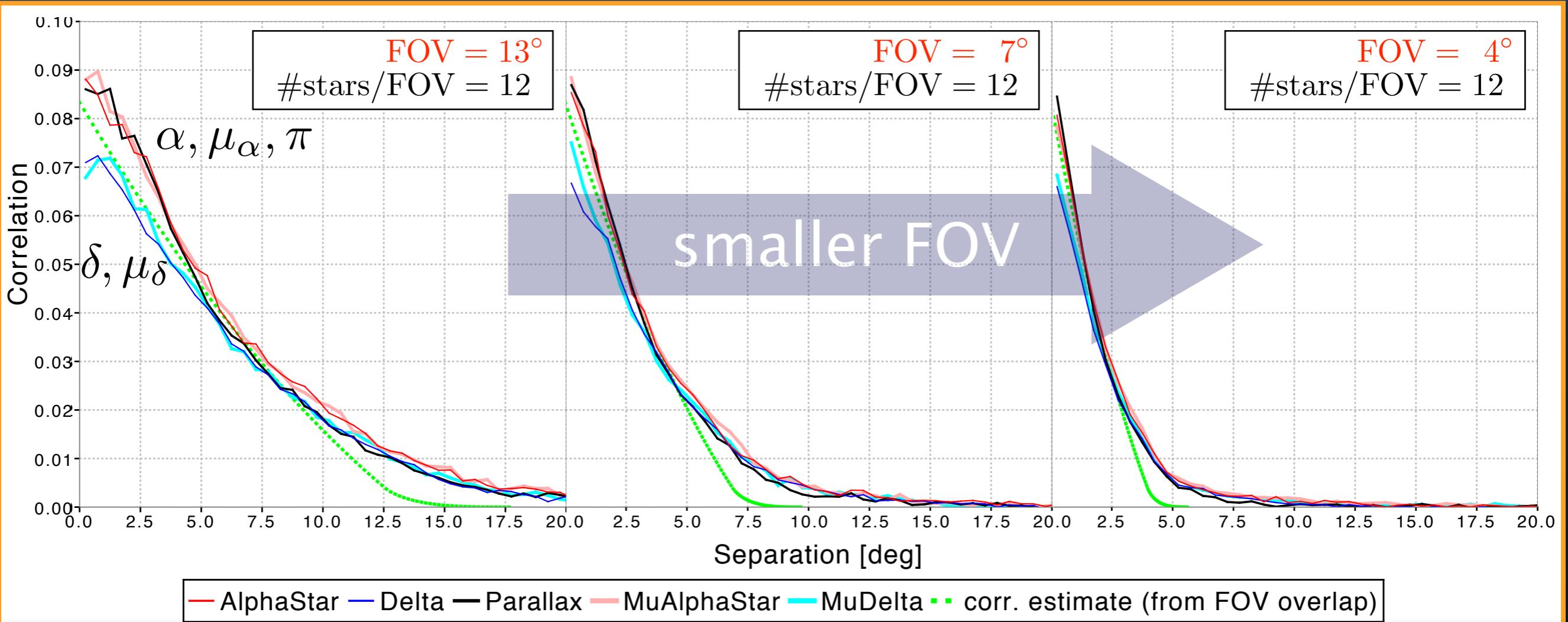
$$E[e_i] \equiv 0 \quad \text{assume no systematic error}$$

$$Cov[\hat{x}_i, \hat{x}_j] = \rho_{ij} \sigma_i \sigma_j \quad (= \sigma_i^2 \text{ for } i = j)$$

$$\rho_{ij} = \frac{E[e_i e_j]}{\sqrt{E[e_i^2] E[e_j^2]}} \quad \text{correlation coefficient}$$

# AGISLab results

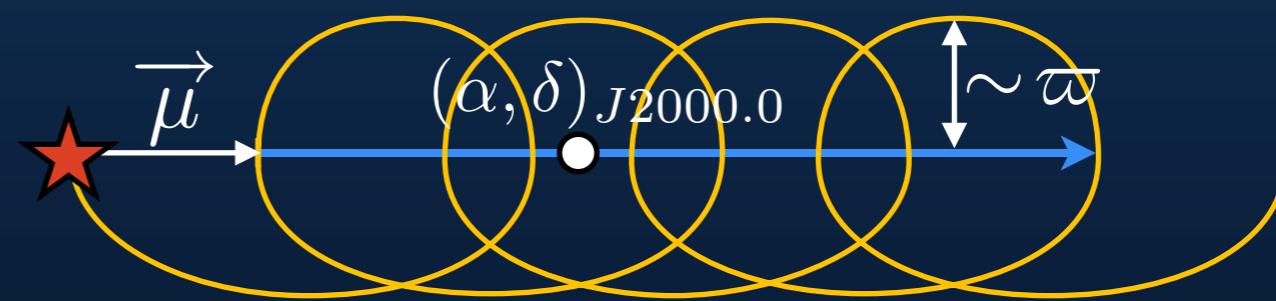
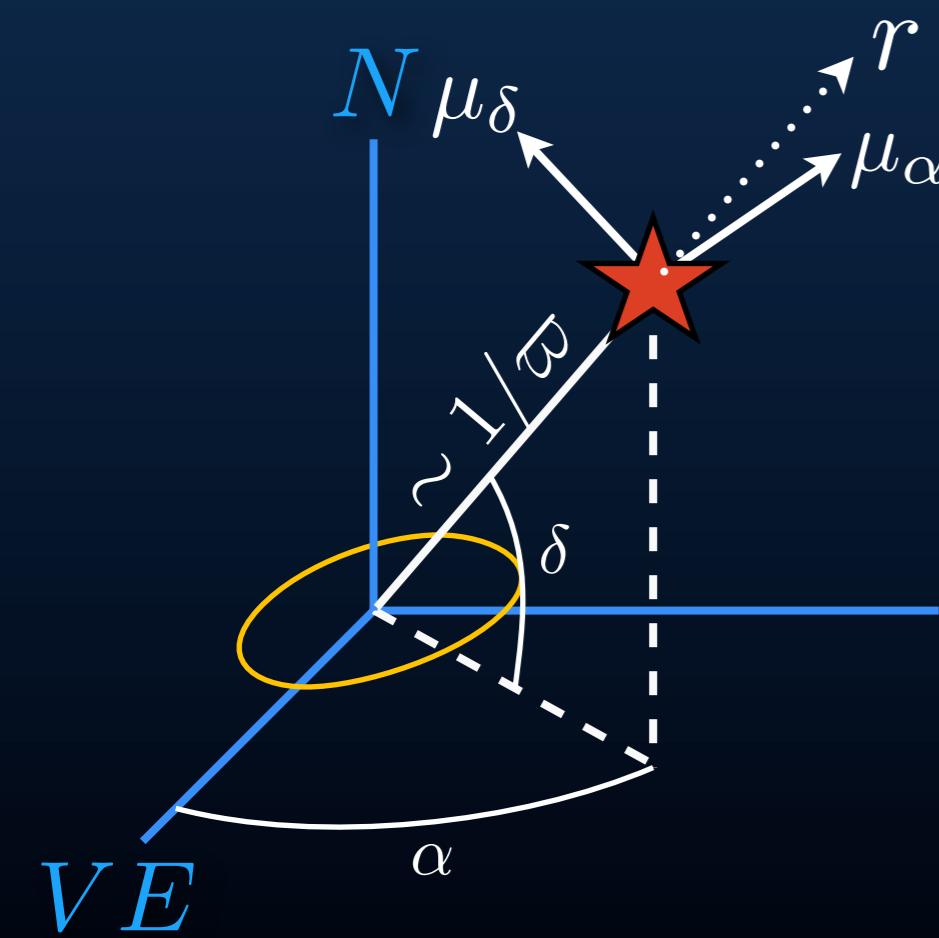




# Astrometric parameters

Which parameters?

Observed position on the sky,  
measured  $\sim 75$  times over 5 year:



Only for close-by fast moving stars,  
radial velocity is apparent:

