



Mappe della radiazione di fondo cosmico

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What a bolometer measures

Let's consider one single bolometer in the focal plane of a scanning telescope.

In every time sample the bolometer measures a convolution of the sky with the beam in the band:

$$V(t) = \mathcal{R} A \int_0^\infty d\nu \int_{4\pi} d\Omega R(\theta - \theta_0, \phi - \phi_0) B(\theta, \phi, \nu) \eta F(\nu) + n$$

Diagram illustrating the components of the bolometer output equation:

- Output [Volt] (points to $V(t)$)
- Responsivity [Volt/W] (points to $\mathcal{R}$)
- Area of the telescope [m²] (points to A)
- Solid angle [sr] (points to $d\Omega$)
- Angular response (points to $R(\theta - \theta_0, \phi - \phi_0)$)
- Direction in the sky (pointing) (points to $\theta - \theta_0, \phi - \phi_0$)
- Brightness of the sky [Jy/sr] (points to $B(\theta, \phi, \nu)$)
- Telescope efficiency (points to η)
- Band transmission (points to $F(\nu)$)
- Noise (points to n)

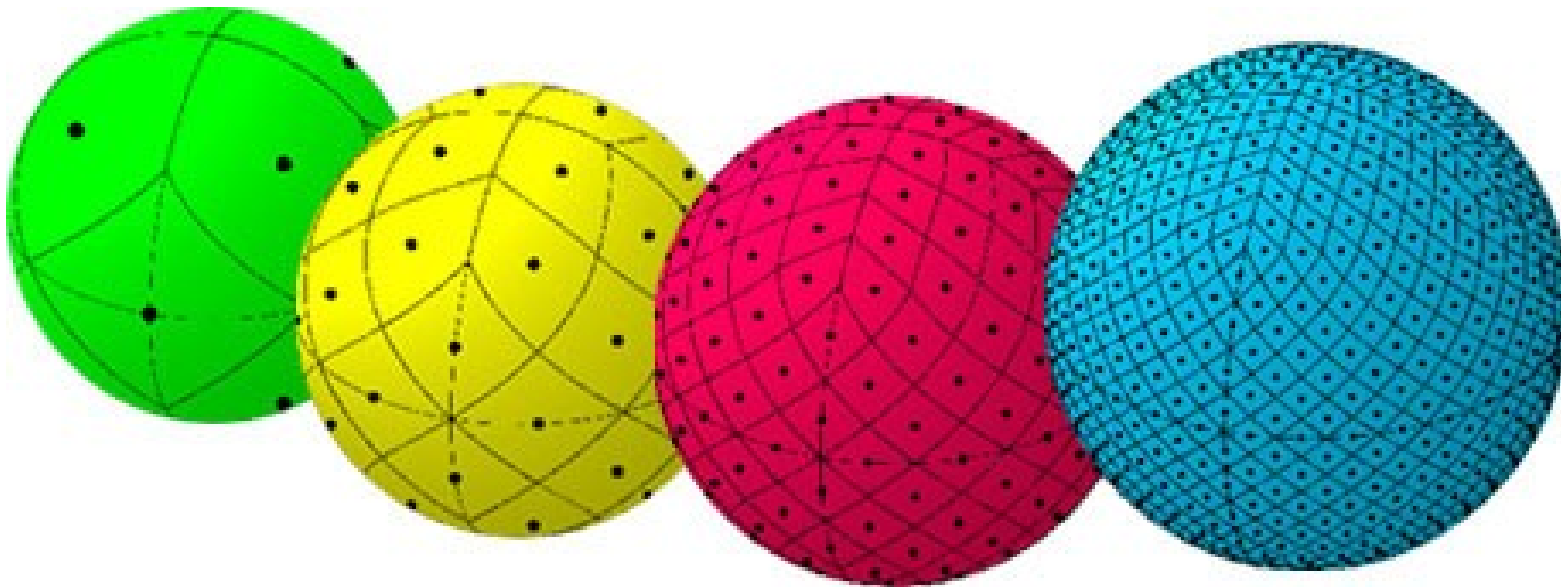
Pixellization

The sky is divided in pixels. To every pixel a pixel number is associated.

Every direction in sky fall in a pixel.

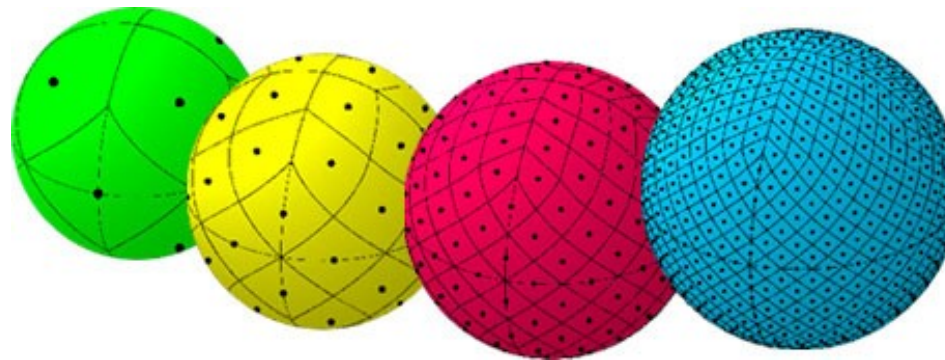
Pixel size is chosen to be $\sim \text{FWHM}/3$

See e.g. <http://healpix.jpl.nasa.gov/index.shtml>



■ HEALPix

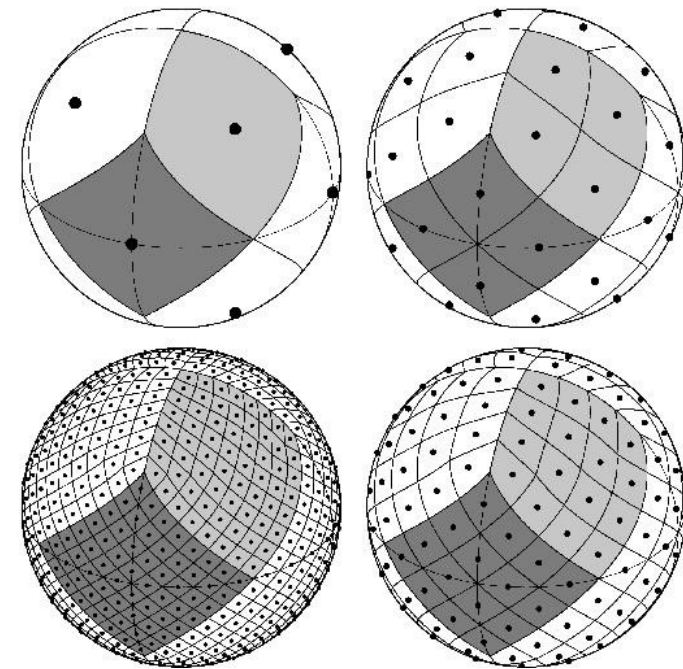
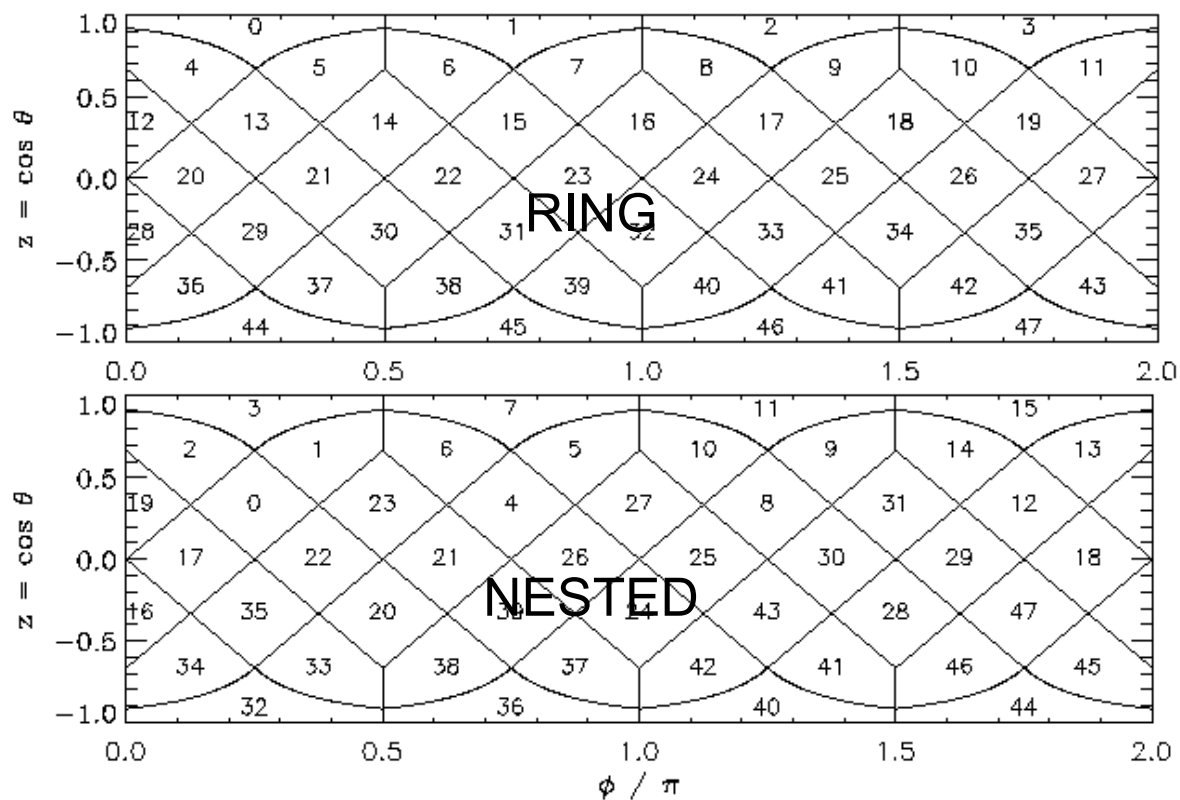
- ➔ HEALPix is an acronym for Hierarchical Equal Area isoLatitude Pixelization of a sphere.
- ➔ Subdivision of a spherical surface
- ➔ Each pixel covers the same surface area as every other pixel.
- ➔ The figure below shows the partitioning of a sphere at progressively higher resolutions, from left to right.
- ➔ The green sphere represents the lowest resolution possible: base partitioning of the sphere surface into 12 equal sized pixels. The yellow sphere has a HEALPix grid of 48 pixels, the red sphere has 192 pixels, and the blue sphere has a grid of 768 pixels (~7.3 degree resolution)



- ➔ The pixel centers, represented by the black dots, occur on a discrete number of rings of constant latitude.
- ➔ The number of constant-latitude rings is dependent on the resolution of the HEALPix grid. For the green, yellow, red, and blue spheres shown, there are 3, 7, 15, and 31 constant-latitude rings, respectively.

- The principal requirements in the development of HEALPix were to create a mathematical structure which supports a suitable discretization of functions on a sphere at sufficiently high resolution, and to facilitate fast and accurate statistical and astrophysical analysis of massive full-sky data sets.
- HEALPix satisfies these requirements because it possesses the following three essential properties:
 - The sphere is hierarchically tessellated into curvilinear quadrilaterals.
 - Areas of all pixels at a given resolution are identical.
 - Pixels are distributed on lines of constant latitude. This property is essential for all harmonic analysis applications involving spherical harmonics.

- Ogni mappa e' costituita da una serie di parametri che ne definiscono le caratteristiche fondamentali
 - NSIDE = [1,2,4,8,16,32,64,128,256,512,1024,2048,...2^n]
 - ➔ definisce il numero di pixel $npix = 12 * nside^2$
 - ➔ Definisce la risoluzione angolare
 - ORDERING = [RING, NESTED]
 - ➔ Esistono due diversi ordinamenti [ad anelli, a nido]



- Disponibile a questo link <http://healpix.jpl.nasa.gov/>
- Fornisce numerosi tools per
 - Leggere e scrivere mappe (fits format)
 - Convertire tra pixel e coordinate
 - Operare sulle mappe (IDL tools)
 - Rappresentare mappe (IDL tools)
 - Trattare polarizzazione con i parametri di Stokes (IDL tools, Fortran tools)
 - Sviluppo in armoniche sferiche e calcolo dello spettro di potenza angolare (Fortran, anafast)
 - Generazione di mappe a partire da armoniche sferiche o da spettri di potenza (Fortran, synfast)
 - ...



What a bolometer measures

Let's consider one single bolometer in the focal plane of a scanning telescope.

In every time sample the bolometer measures a convolution of the sky with the beam in the band.

In a more compact form:

$$V_i = \kappa B_j + n_i$$

Output of sample i

Calibration

Brightness of sky pixel j

Noise at sample i

Calibration includes: spectral behavior of the source, responsivity, efficiency.

The **sky** is considered **pixellized** and convolved with the beam.



What a bolometer measures

Let's consider one single bolometer in the focal plane of a scanning telescope.

In every time sample the bolometer measures a convolution of the sky with the beam in the band.

In matrix form:

$$\vec{d} = \mathcal{K} \mathbf{A} \vec{m} + \vec{n}$$

Data vector Calibration Pointing matrix Map of the sky Noise vector

nd elements (very large)
(np x nd) elements
np elements (large)
nd elements (very large)

The problem is now a large linear system.

Pointing matrix has only one 1 and a lot of 0 per line. It converts from pixel space to time space.



Pointing matrix

$$A_{tp} = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} \quad \begin{matrix} nd = 5 \\ np = 3 \end{matrix}$$

A has (nd x np) elements

At time t=0 the telescope is observing pixel 2

At time t=1 the telescope is observing pixel 1

.....

At time t=4 the telescope is observing pixel 1

A matrix creates a Time Ordered Data from a sky map:

$$\vec{d} = A \vec{m} \quad \longrightarrow \quad \begin{pmatrix} d_{t_0} \\ d_{t_1} \\ \dots \\ d_{t_4} \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} m_{p_1} \\ m_{p_2} \\ m_{p_3} \end{pmatrix} = \begin{pmatrix} m_{p_2} \\ m_{p_1} \\ \dots \\ m_{p_1} \end{pmatrix} \quad nd = 5$$

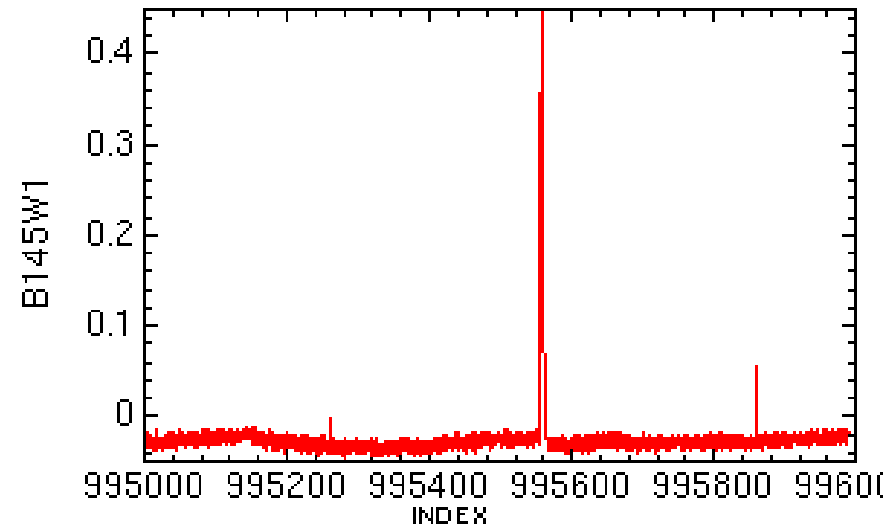
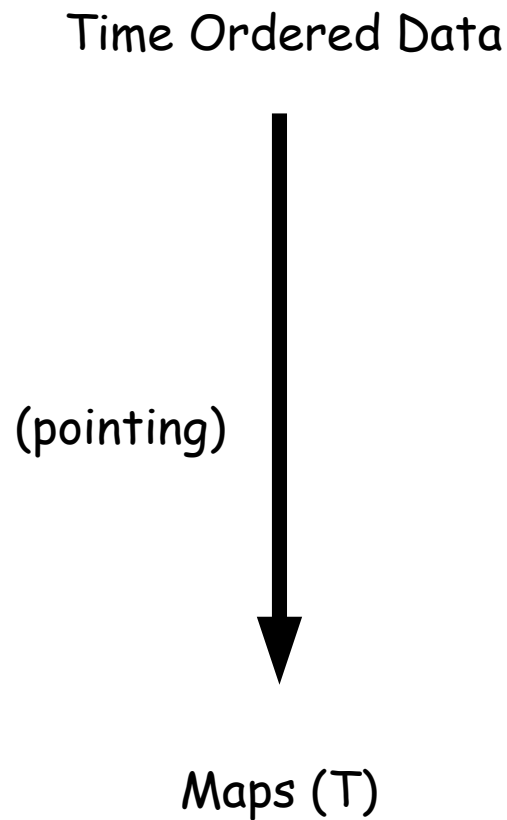
A^T sums all the data collected observing the same pixel:

$$A^T \vec{d} = \begin{pmatrix} 0 & 1 & 0 & 0 & 1 \\ 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} d_{t_0} \\ d_{t_1} \\ d_{t_2} \\ d_{t_3} \\ d_{t_4} \end{pmatrix} = \begin{pmatrix} 2 m_{p_1} \\ 2 m_{p_2} \\ m_{p_3} \end{pmatrix} \quad np = 3$$

$A^T A$ is a diagonal matrix encoding the number of observation falling in the same pixel:

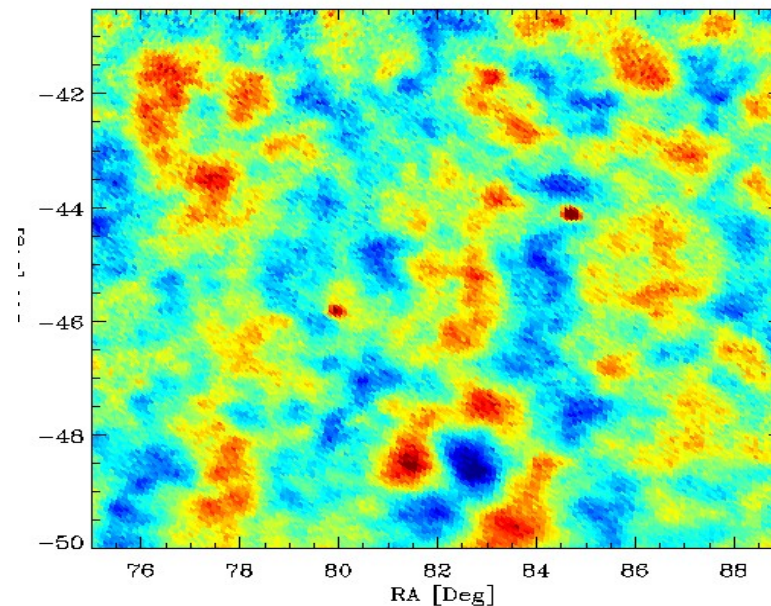
$$A^T A = \begin{pmatrix} 0 & 1 & 0 & 0 & 1 \\ 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad np = 3$$

Map-making problem



IGLS T map of all 145GHz detectors

-300 -200 -100 0 100 200 300 uK





Map-making solution, naive solution

Ideal time ordered data: $\vec{d} = A \vec{m} + \vec{n}$

A simple solution: $\tilde{m} = (A^T A)^{-1} A^T \vec{d}$

- in every pixel, put the average of all the data falling in that pixel
- this solution is correct if the noise is white (no correlations between observation at different time)



Come fare in pratica, algoritmo naïve

- Supponiamo di avere: $ra(t)$, $dec(t)$, $segnale(t)$, tre vettori della stessa lunghezza, in un file ASCII
- Codice IDL

```
init_healpix                ; inizializza healpix
nside = 1024L               ; definisce la risoluzione
ATd = fltarr(nside^2*12)    ; numeratore
ATA = lonarr(nside^2*12)    ; denominatore (contatore)
map = fltarr(nside^2*12)    ; mappa da costruire

readcol, 'data.dat', ra, dec, signal, count=ndata ; legge i dati

th = (90-dec)/180!*pi       ; converte in coordinate polari
phi = ra/180!*pi
pix2ang_ring, nside, th, phi, pixel ; genera vettore pixel

for i=0L, ndata-1 do begin   ; popola numeratore e denominatore
    ATd[pixel[i]] += signal[i]
    ATA[pixel[i]] += 1
endfor

map[*] = !healpix.bad_value ; valore "bad" per definizione
good = where(ATA gt 0)      ; identifico i pixel osservati
map[good] = ATd[good]/ATA[good] ; costruisco la mappa nei pixel osservati

write_fits_map, file='mappa.fits', map, nside=nside, coordsys='C', /ring, units='K_cmb'

mollview, 'mappa.fits'      ; disegna la mappa nel file salvato
```



- Metodo di massima verosimiglianza
- Abbiamo dei dati ($\mathbf{d}$) e un modello che cerchiamo di ricavare: $\mathbf{m}$
- Possiamo definire una verosimiglianza (likelihood):

$$\mathcal{L} = \frac{1}{(2\pi)^{n/2} |\mathbf{N}|^{1/2}} \exp \left[-\frac{1}{2} (\mathbf{d} - \mathbf{A}\mathbf{m})^T \mathbf{N}^{-1} (\mathbf{d} - \mathbf{A}\mathbf{m}) \right]$$

- Dove $\mathbf{N}$ e' la matrice delle covarianze del rumore

$$\mathbf{N} = \langle (\mathbf{d} - \mathbf{A}\mathbf{m})(\mathbf{d} - \mathbf{A}\mathbf{m})^T \rangle_{ens}$$

- Questa likelihood viene usata spesso, quando possiamo ricondurci a variabili distribuite in modo gaussiano (cmb, rumore, no spettro di potenza)



- Abbiamo che un rivelatore misura dati ordinati in tempo

$$\mathbf{d} = \mathbf{A}\mathbf{m} + \mathbf{n}$$

$$\mathbf{n} = \mathbf{d} - \mathbf{A}\mathbf{m}$$

$$\mathbf{N} = \langle \mathbf{n} \cdot \mathbf{n}^T \rangle_{ens}$$

$$\mathcal{L} = \frac{1}{(2\pi)^{n/2} |\mathbf{N}|^{1/2}} \exp \left[-\frac{1}{2} \mathbf{n}^T \mathbf{N}^{-1} \mathbf{n} \right]$$

$$\mathcal{L} = \frac{1}{(2\pi)^{n/2} |\mathbf{N}|^{1/2}} \exp \left[-\frac{1}{2} (\mathbf{d} - \mathbf{A}\mathbf{m})^T \mathbf{N}^{-1} (\mathbf{d} - \mathbf{A}\mathbf{m}) \right]$$

$$\frac{\partial \mathcal{L}}{\partial \mathbf{m}} = 0 \quad \frac{\partial}{\partial \mathbf{m}} \left((\mathbf{d} - \mathbf{A}\mathbf{m})^T \mathbf{N}^{-1} (\mathbf{d} - \mathbf{A}\mathbf{m}) \right) = 0$$



Map-making solution, ideal case

Derivative with $\mathbf{x}$ and $\mathbf{y}$ vectors

$$\frac{\partial \mathbf{y}}{\partial \mathbf{x}} = \begin{bmatrix} \frac{\partial y_1}{\partial x_1} & \frac{\partial y_1}{\partial x_2} & \dots & \frac{\partial y_1}{\partial x_n} \\ \frac{\partial y_2}{\partial x_1} & \frac{\partial y_2}{\partial x_2} & \dots & \frac{\partial y_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial y_m}{\partial x_1} & \frac{\partial y_m}{\partial x_2} & \dots & \frac{\partial y_m}{\partial x_n} \end{bmatrix} \Rightarrow \frac{\partial}{\partial \mathbf{x}} \mathbf{A} \mathbf{x} = \mathbf{A}$$

$$\frac{\partial}{\partial \mathbf{m}} [(\mathbf{d} - \mathbf{A} \mathbf{m})^T \mathbf{N}^{-1} (\mathbf{d} - \mathbf{A} \mathbf{m})] = 0$$

$$-\mathbf{A}^T \mathbf{N}^{-1} (\mathbf{d} - \mathbf{A} \mathbf{m}) - (\mathbf{d} - \mathbf{A} \mathbf{m})^T \mathbf{N}^{-1} \mathbf{A} = 0$$

$$-\mathbf{A}^T \mathbf{N}^{-1} (\mathbf{d} - \mathbf{A} \mathbf{m}) = 0$$

$$\mathbf{A}^T \mathbf{N}^{-1} \mathbf{d} = \mathbf{A}^T \mathbf{N}^{-1} \mathbf{A} \mathbf{m}$$

$$\tilde{\mathbf{m}} = (\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A})^{-1} \mathbf{A}^T \mathbf{N}^{-1} \mathbf{d}$$

Comparison with the naive solution:

$$\tilde{\mathbf{m}} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \vec{\mathbf{d}} \quad \neq \quad \tilde{\mathbf{m}} = (\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A})^{-1} \mathbf{A}^T \mathbf{N}^{-1} \vec{\mathbf{d}}$$

Average of all the observations
falling in the same pixel

Optimal noise weighted average of all
the observations falling in the same pixel



Map-making solution, ideal case

$$\vec{d} = \mathbf{A}\vec{m} + \vec{n}$$

Ideal time ordered data

$$\tilde{m} = (\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A})^{-1} \mathbf{A}^T \mathbf{N}^{-1} \vec{d}$$

Map solved by inversion
regularized by noise

$$\mathbf{N}_{i,j} = \langle n_i n_j^T \rangle \simeq \xi(|t_j - t_i|) \quad \text{time-time noise covariance matrix}$$

Check:

$$\begin{aligned} \tilde{m} &= (\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A})^{-1} \mathbf{A}^T \mathbf{N}^{-1} \vec{d} \\ &= (\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A})^{-1} \mathbf{A}^T \mathbf{N}^{-1} (\mathbf{A}\vec{m} + \vec{n}) \\ &= (\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A})^{-1} \mathbf{A}^T \mathbf{N}^{-1} \mathbf{A} \vec{m} + (\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A})^{-1} \mathbf{A}^T \mathbf{N}^{-1} \vec{n} \\ &= \vec{m} + n_p \end{aligned}$$

Solved map = sky + pixel noise



Map-making vs naive solutions

If the noise is white, i.e. no correlations between different samples, it can be described using only its RMS amplitude σ , and the noise covariance matrix N can be written as:

$$N = \sigma^2 I \quad \longrightarrow \quad N^{-1} = 1/\sigma^2 I$$

$$\begin{aligned}\tilde{m} &= (A^T N^{-1} A)^{-1} A^T N^{-1} \vec{d} \\ &= \sigma^2 (A^T I A)^{-1} A^T \sigma^{-2} I \vec{d} \\ &= (A^T A)^{-1} A^T \vec{d}\end{aligned}$$

Which is the naive solution!

In general the white noise assumption is not valid (e.g. $1/f$ noise), and we have to use the N matrix in the general form:

$$N_{i,j} = \langle n_i n_j^T \rangle \simeq \xi(|t_j - t_i|)$$

The problem is that both n_d (data) and n_p (pixels) are huge.

For Planck, $n_p = 50 \cdot 10^6$, $n_d = 25 \cdot 10^9$ per detector

The full noise matrix N_{ij} cannot be used.

The matrix $(A^T N^{-1} A)_{ij}$ is $n_p \times n_p$ and is hard to invert



Map-making, fast solutions

Natoli et al., astro-ph/0101252

De Gasperis et al, astro-ph/0502142

$$\vec{m} = \left(\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A} \right)^{-1} \mathbf{A}^T \mathbf{N}^{-1} \vec{d}$$

$$\left(\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A} \right) \vec{m} = \mathbf{A}^T \mathbf{N}^{-1} \vec{d}$$

$$\mathbf{A} \mathbf{x} = \mathbf{b} \quad \longrightarrow \quad \mathbf{A} \mathbf{x} - \mathbf{b} = 0$$

$$\varepsilon = \frac{\left\| \left(\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A} \right) \vec{m} - \mathbf{A}^T \mathbf{N}^{-1} \vec{d} \right\|}{\left\| \mathbf{A}^T \mathbf{N}^{-1} \vec{d} \right\|}$$

Note that $\frac{d}{dx} \left(\frac{1}{2} \mathbf{A} \mathbf{x}^2 - \mathbf{b} \mathbf{x} \right) = \mathbf{A} \mathbf{x} - \mathbf{b}$, hence,
solving the linear system is the same as finding
the minimum of the function $\frac{1}{2} \mathbf{A} \mathbf{x}^2 - \mathbf{b} \mathbf{x}$

The solution...

...can be written like
that...

...as a linear system

The algorithm iteratively
search the $\mathbf{m}$ than
minimizes this norm using

**Conjugate Gradient
Solver**

or

Jacobi iterative solver



Map-making, fast solutions

Natoli et al., astro-ph/0101252

De Gasperis et al, astro-ph/0502142

The tricky part is that under some hypothesis, we don't need to calculate $\mathbf{N}^{-1}$.

If the noise is stationary and the noise correlation is not too long, then $\mathbf{N}$ is band diagonal and circulant ($n_{ij} = n_{i+1,j+1 \bmod n_d}$).

$$\text{Then } \mathbf{N}^{-1} \vec{d} = \mathbf{F}_t^\dagger \mathbf{\Xi}^{-1} \mathbf{F}_t \vec{d}$$

where $\mathbf{F}_t$ is the Fourier transform operator and $\mathbf{\Xi}^{-1}$ is a diagonal matrix.

The diagonal part of $\mathbf{\Xi}^{-1}$ is the inverse of the noise power spectrum

Multiplying by $\mathbf{N}^{-1}$ is now a convolution by the $\mathbf{PS}^{-1}$.

Recipe:

the *noise filter* $\mathbf{\Xi}^{-1}$ is the Fourier transform of the inverse of the noise power spectrum

Convert the map estimated at the previous iteration in a TOD

$$\underbrace{(\mathbf{A}^\top \mathbf{F}_t^\dagger \mathbf{\Xi}^{-1} \mathbf{F}_t \mathbf{A})}_{\text{Same as R.H.S}} \tilde{\mathbf{m}} = \underbrace{\mathbf{A}^\top \mathbf{F}_t^\dagger}_{\text{go back to time domain}} \underbrace{\mathbf{\Xi}^{-1} \mathbf{F}_t}_{\text{do FFT of the TOD multiply by PS}^{-1}} \vec{d}$$

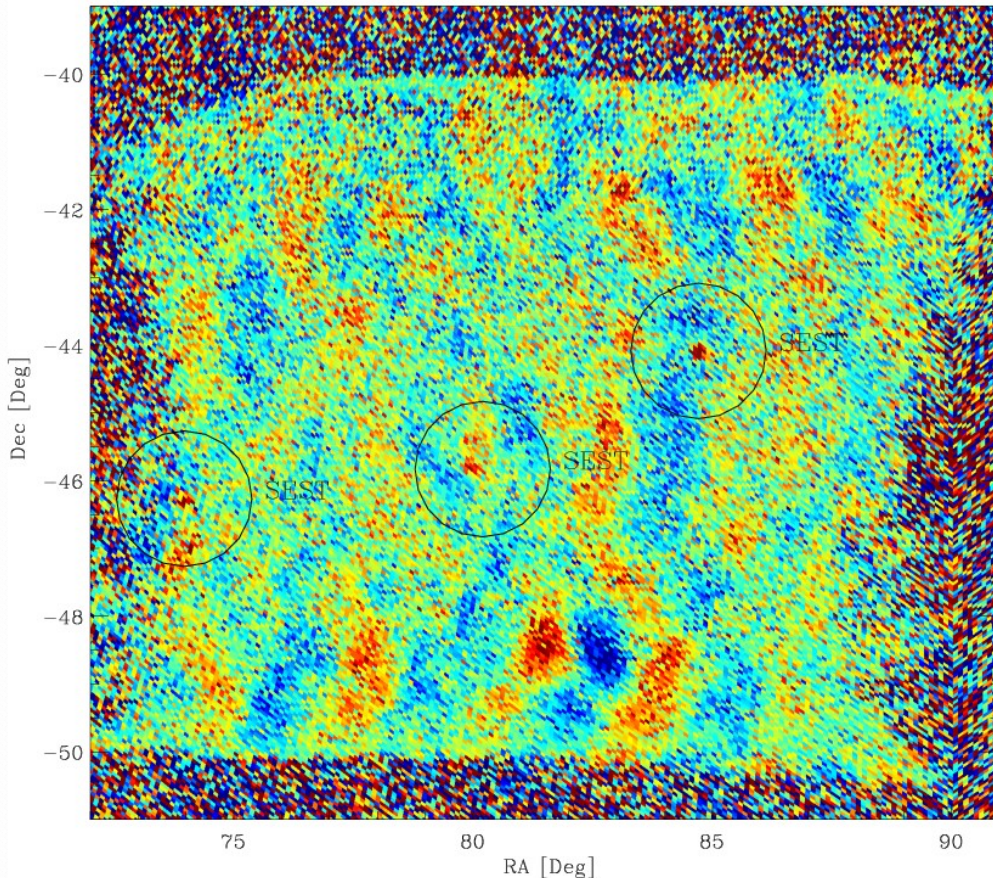
sum the observation in a map

Map-making: comparison

Naive map of W1 detector

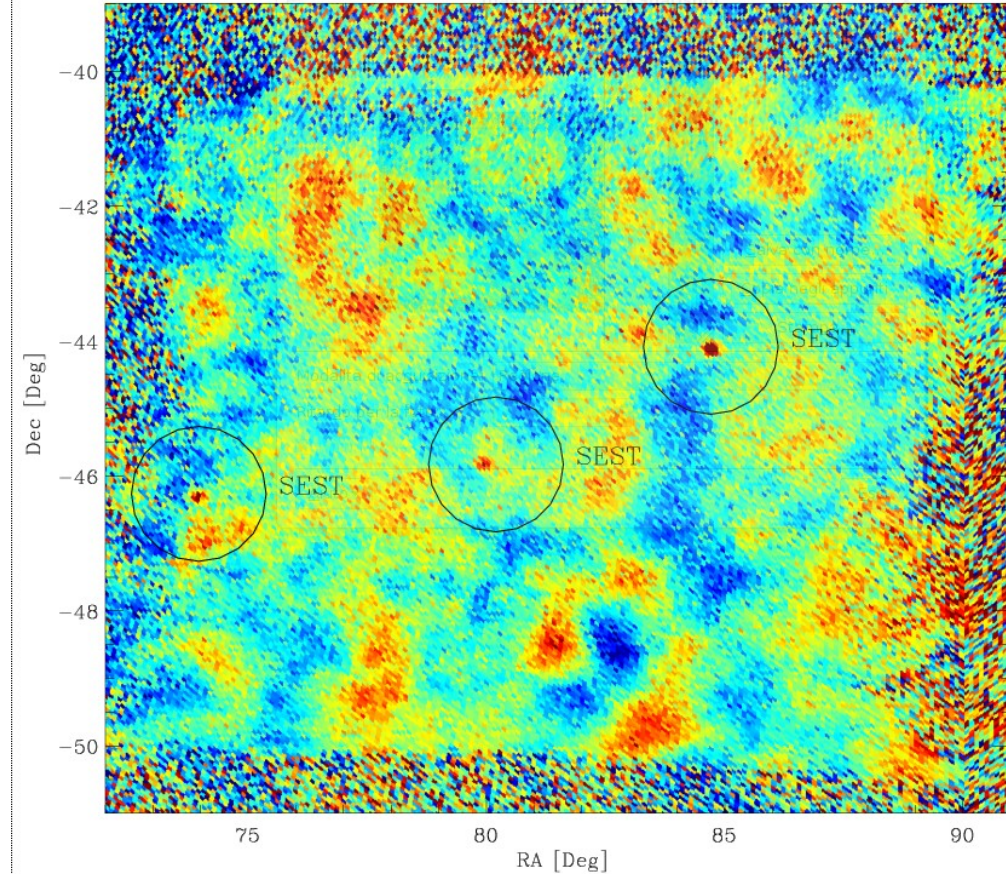
-400 -200 0 200 400 uK_CMB

igls_w2.png - Ksnapshot



IGLS map of W1 detector

-400 -200 0 200 400 uK_cmb



Being non-optimal, the naive map is more noisy than the IGLS one!

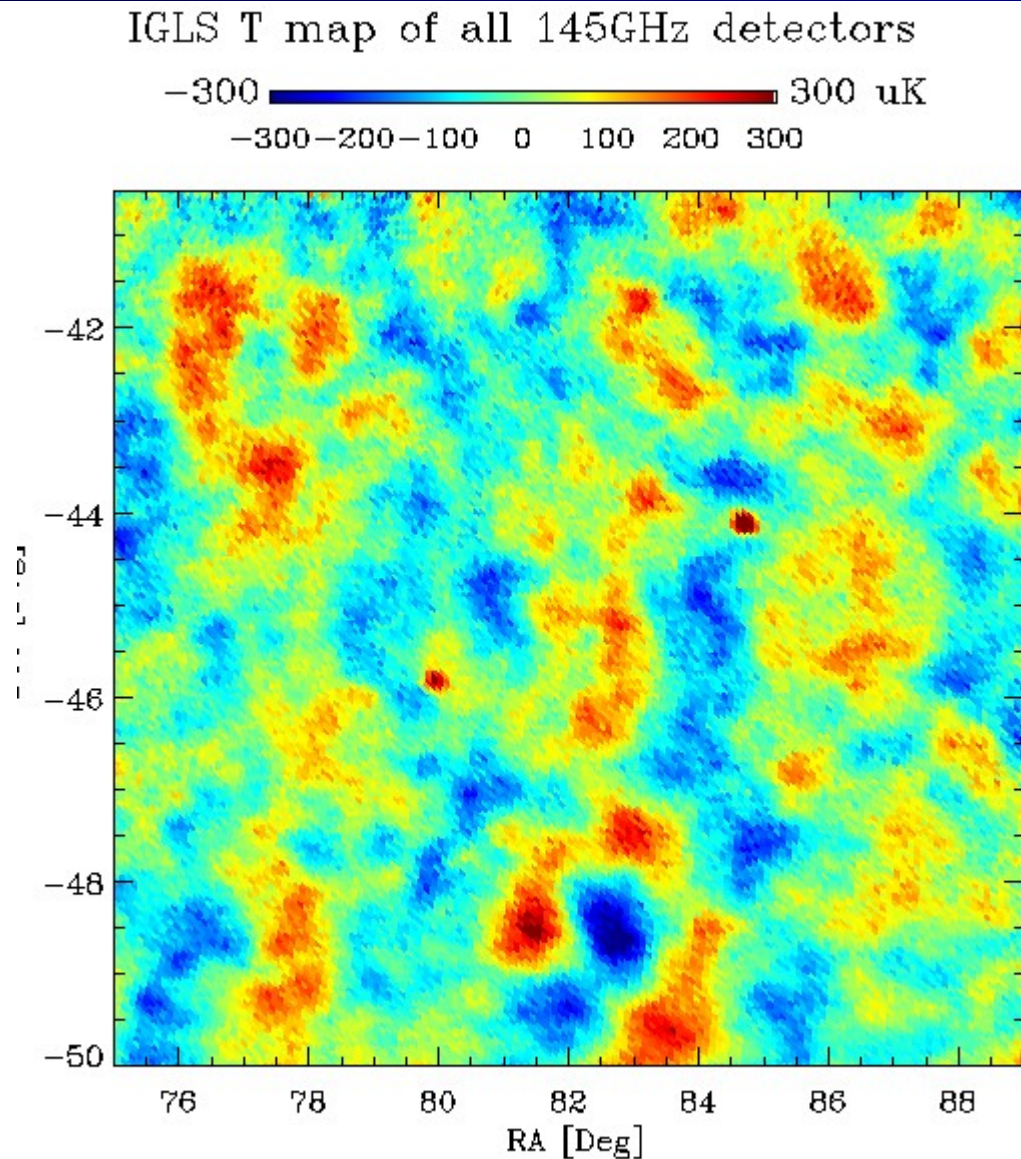
Map-making with several detectors

- We can make 1 map for each detector and then do the average, but this is not optimal because the detectors have different noise properties.

- We can build a data vector as:

$$\vec{d} = (d_0^1 \dots d_n^1 d_0^2 \dots d_n^2 \dots d_0^M \dots d_n^M)$$
and use the optimal map-making, but this is not optimal too

- The solution is to build also a full noise covariance matrix which is block diagonal, each block being associated to one detector. This means we are assuming piece-wise noise stationarity, but the properties of the noise matrix are still valid and we can estimate the map-making solution in exactly the same way as for single detector.





Stokes parameters

- La radiazione incidente incontra un filtro che trasmette $\frac{1}{2}$
 - Filtro 1: I_0 isotropico
 - Filtro 2: I_1 polarizzatore lungo l'asse x
 - Filtro 3: I_2 polarizzatore lungo un asse a 45 gradi
 - Filtro 4: I_3 trasmette solo polarizzazione circolare destra
- Definiamo 4 parametri di Stokes come

$$\begin{aligned} I &= 2I_0 \\ Q &= 2I_1 - 2I_0 \\ U &= 2I_2 - 2I_0 \\ V &= 2I_3 - 2I_0 \end{aligned} \quad (1.35)$$

Stokes parameters

It's easier to explicit this definition using the complex form of the electric fields

$$\begin{cases} E_x(t) = E_{0x}(t)e^{i\delta_x(t)}e^{i(kz-\omega t)} \\ E_y(t) = E_{0y}(t)e^{i\delta_y(t)}e^{i(kz-\omega t)} \\ E_z(t) = 0 \end{cases} \quad (1.36)$$

Then

$$\begin{aligned} I_0 &= \frac{1}{2} \langle E_x E_x^* + E_y E_y^* \rangle \\ I_1 &= \langle E_x E_x^* \rangle \\ I_2 &= \langle E'_x E'^*_x \rangle = \langle \frac{1}{\sqrt{2}}(E_x + E_y) \frac{1}{\sqrt{2}}(E_x^* + E_y^*) \rangle \\ I_3 &= \langle E''_x E''^*_x \rangle = \langle \frac{1}{\sqrt{2}}(E_x + iE_y) \frac{1}{\sqrt{2}}(E_x^* - iE_y^*) \rangle \end{aligned} \quad (1.37)$$

where was used $E'_x = \frac{1}{\sqrt{2}}(E_x + E_y)$ (rotation of 45 degrees on the plane $z = 0$ to detect 45 degrees linear polarization) and $E''_x = \frac{1}{\sqrt{2}}(E_x + iE_y)$: 90 degrees phase-shift of the $\hat{y}$ component and rotation of 45 degrees to measure right circular polarization. Plugging the (1.37) into the (1.35) we find

$$\begin{bmatrix} I \\ Q \\ U \\ V \end{bmatrix} = \begin{bmatrix} \langle E_x E_x^* + E_y E_y^* \rangle \\ \langle E_x E_x^* - E_y E_y^* \rangle \\ \langle E_x E_y^* + E_y E_x^* \rangle \\ i \langle E_x E_y^* - E_y E_x^* \rangle \end{bmatrix} = \begin{bmatrix} \langle E_{0x}^2 + E_{0y}^2 \rangle \\ \langle E_{0x}^2 - E_{0y}^2 \rangle \\ \langle 2E_{0x}E_{0y} \cos \delta \rangle \\ \langle 2E_{0x}E_{0y} \sin \delta \rangle \end{bmatrix} \quad (1.38)$$

where $\delta(t) = \delta_x(t) - \delta_y(t)$, $E_{0x}(t)$ and $E_{0y}(t)$ are time dependent and averaging is made on time scales longer than the period T of the wave.

1.5.2 Polarized and unpolarized wave

A monochromatic wave is always polarized. If we define the Stokes parameters for a monochromatic wave, we can remove the time averages in (1.38) and find

$$Q^2 + U^2 + V^2 = I^2 \quad (1.42)$$

A non-monochromatic wave is not necessarily polarized. The degree of polarization is defined

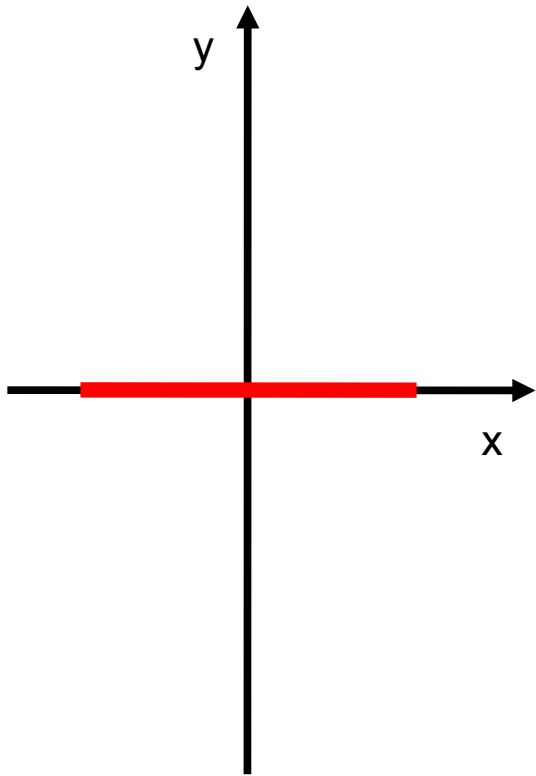
$$\Pi = \frac{\sqrt{Q^2 + U^2 + V^2}}{I} \quad (1.43)$$

and is unity for a polarized wave. Linear polarization may be represented by rods

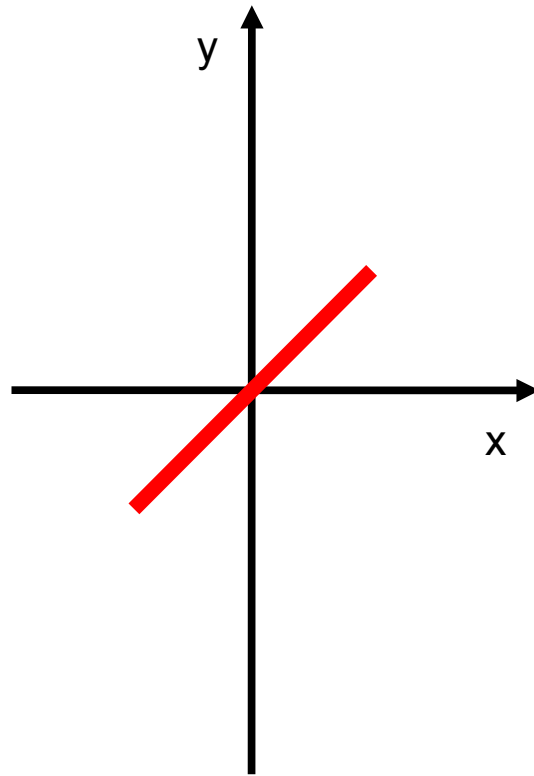
$$\bar{P} = \begin{bmatrix} P \cos \alpha \\ P \sin \alpha \end{bmatrix} \quad (1.44)$$

of magnitude $P = \sqrt{(Q^2 + U^2)}/I$, rotationally invariant, and angle α with respect to the chosen frame: $\alpha = \frac{1}{2} \arctan(U/Q)$. The use of rods instead of vectors reflects the spin-2 nature of linear polarization: 180 degrees rotations return to the original state. To have an unpolarized wave, all the time dependent quantities in the (1.38) δ , E_{0x} and E_{0y} must be random distributions and $\langle E_{0x}(t) \rangle = \langle E_{0y}(t) \rangle$. This is verified, for example, in radiation of thermal origin.

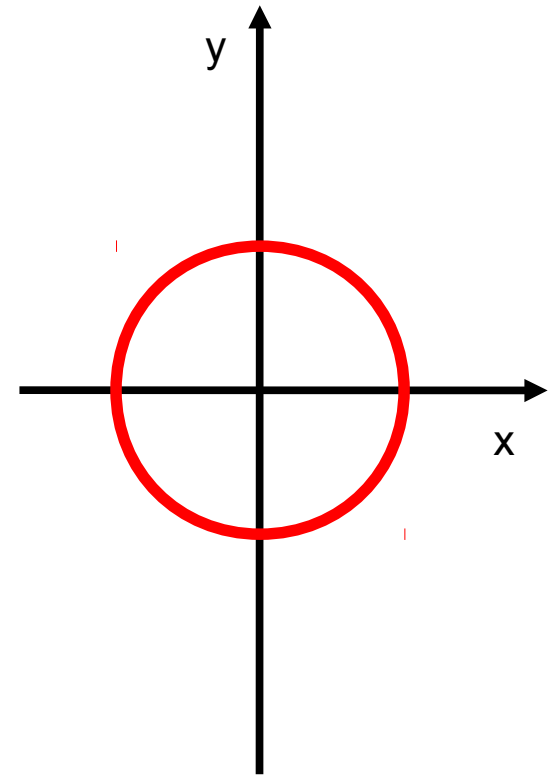
Stokes parameters



$$\begin{aligned} Q &= I \\ U &= 0 \\ V &= 0 \end{aligned}$$



$$\begin{aligned} Q &= 0 \\ U &= I \\ V &= 0 \end{aligned}$$



$$\begin{aligned} Q &= 0 \\ U &= 0 \\ V &= I \end{aligned}$$

Stokes parameters

Rotation of the reference frame by an angle θ around the propagation vector acts on the Stokes parameters as

$$\begin{bmatrix} I' \\ Q' \\ U' \\ V' \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos(2\theta) & \sin(2\theta) & 0 \\ 0 & -\sin(2\theta) & \cos(2\theta) & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} I \\ Q \\ U \\ V \end{bmatrix} \quad (1.41)$$

I and V are invariant, Q and U are said to be spin 2 because they are invariant for 180 degrees rotations.



From equation (1.35) we have

$$\begin{aligned} I_x &= (I + Q)/2 \\ I_y &= (I - Q)/2 \end{aligned} \quad (1.49)$$

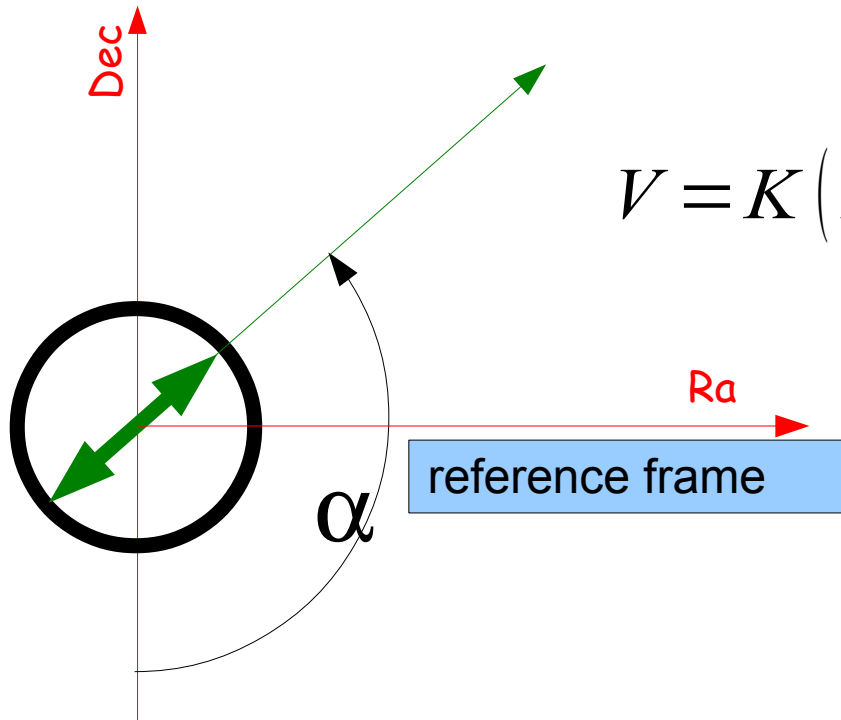
A detector with a polarizer along x measures

$$I_x = (I + Q)/2$$

A polarization sensitive detector in a generic direction measures

$$(T = I) \quad V = \mathcal{K} [T(\hat{n}) + Q(\hat{n}) \cos(2\alpha)] + U(\hat{n}) \sin(2\alpha) + n \quad (3.1)$$

With polarization the detector measures



$$V = K \left(T_p + Q_p \cos(2\alpha) + U_p \sin(2\alpha) \right) + n$$

matrix form

$$V = K \begin{bmatrix} 1 & \cos(2\alpha) & \sin(2\alpha) \end{bmatrix} \begin{bmatrix} T_p \\ Q_p \\ U_p \end{bmatrix} + n$$

Stokes parameters Q , U
are defined in a reference frame

Map-making problem



Inverse problem: from sky to TOD

$$\begin{bmatrix} V_0 \\ V_1 \\ \dots \\ V_i \\ \dots \\ V_{Nd} \end{bmatrix} = K \begin{bmatrix} 000100 & 000\cos(2\alpha)00 & 000\sin(2\alpha)00 \\ 010000 & 0\cos(2\alpha)0000 & 0\sin(2\alpha)0000 \\ \dots & \dots & \dots \\ 001000 & 00\cos(2\alpha)000 & 00\sin(2\alpha)000 \\ \dots & \dots & \dots \\ 100000 & \cos(2\alpha)00000 & \sin(2\alpha)00000 \end{bmatrix} \begin{bmatrix} T_0 \\ T_1 \\ \dots \\ T_p \\ \dots \\ T_{Np} \\ Q_0 \\ Q_1 \\ \dots \\ Q_p \\ \dots \\ Q_{Np} \\ U_0 \\ U_1 \\ \dots \\ U_p \\ \dots \\ U_{Np} \end{bmatrix} + \begin{bmatrix} n_0 \\ n_1 \\ \dots \\ n_i \\ \dots \\ n_{Nd} \end{bmatrix}$$

TOD pointing matrix ($N_d \times 3N_p$) noise

New pointing matrix, same algebra, same solutions pixellized sky

Naive polarization map-making

- In the polarization mapmaking, the naïve solution is not as simple as average per pixel
- In each pixel there are 3 information to recover (T, Q, U)
- For each pixel, we can write a ($n_d \times 3$) system, where n_d is the number of data in that pixel

$$d_i = I_p + Q_p \cos(2\alpha_i) + U_p \sin(2\alpha_i)$$

$$A = \begin{pmatrix} 1 & \cos(2\alpha_1) & \sin(2\alpha_1) \\ 1 & \cos(2\alpha_2) & \sin(2\alpha_2) \\ \dots & \dots & \dots \\ 1 & \cos(2\alpha_{n_d}) & \sin(2\alpha_{n_d}) \end{pmatrix}$$



Naive polarization map-making

- For each pixel

$$\tilde{m} = (A^T A)^{-1} A^T d$$

$$A^T A = \begin{pmatrix} \sum_{n_d} 1 & \sum_{n_d} \cos(2\alpha_i) & \sum_{n_d} \sin(2\alpha_i) \\ \sum_{n_d} \cos(2\alpha_i) & \sum_{n_d} \cos^2(2\alpha_i) & \sum_{n_d} \cos(2\alpha_i) \sin(2\alpha_i) \\ \sum_{n_d} \sin(2\alpha_i) & \sum_{n_d} \cos(2\alpha_i) \sin(2\alpha_i) & \sum_{n_d} \sin^2(2\alpha_i) \end{pmatrix}$$

$$A^T d = \begin{pmatrix} \sum_{n_d} d_i \\ \sum_{n_d} \cos(2\alpha_i) d_i \\ \sum_{n_d} \sin(2\alpha_i) d_i \end{pmatrix}$$



Calibration of the polarimeters

- This is a very simplified statement of the problem

$$V = K [T_p + Q_p \cos(2\alpha) + U_p \sin(2\alpha)] + n$$

- To be realistic we have to introduce at least the polarimeter efficiency, and the polarimeter orientation angle (in the telescope reference frame)

$$V = K [T_p + \rho_{\text{det}} (Q_p \cos(2(\alpha + \psi_{\text{det}})) + U_p \sin(2\alpha + \psi_{\text{det}}))] + n$$

- Moreover, the different detectors must be well matched to be combined, in particular in terms of Optical response (beam)