

# Reference Frames & Reference Systems in Geodesy

F. Sansò, A. Vatalis

Grafarend 75 - Berlin 13/02/2015

# Reference Frames & Reference Systems on Geodesy

Conceptually a central problem in Geodesy where we want to estimate coordinates from observables that are independent of the Coordinate System chosen (what theoretical physicists call *gauge invariance*)

The rank defect of the deterministic model

$$\underline{y} = \begin{vmatrix} y_1 \\ \vdots \\ y_M \end{vmatrix} = F(\underline{x}) \quad \underline{x} = \begin{vmatrix} \underline{x}_1 \\ \underline{x}_2 \\ \vdots \\ \underline{x}_N \end{vmatrix}$$

↑

Observables

↑

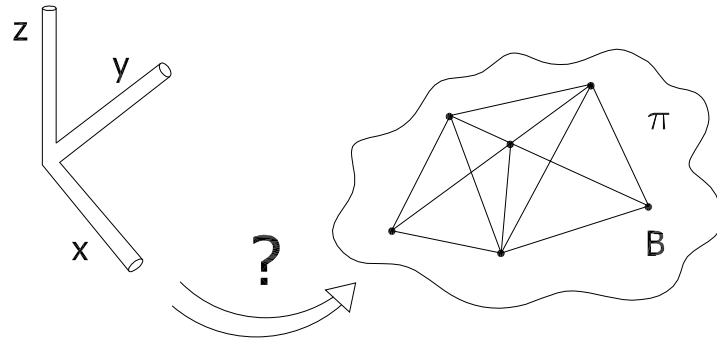
Parameters ( $3N$  coordinates of  $N$  points)

is strictly related to the need of defining a R.S. and a R.F. for further geodetic operations (e.g. networks densification).

Geodetic/Statistical literature on the subject by Bjerhammar, More/Penrose, Rao, Tautenberg, Grafarend...

# Classical definition

- A R.S. is a physical Cartesian triad in a chosen position with respect to a body  $B$  represented by a polyhedron  $\pi$

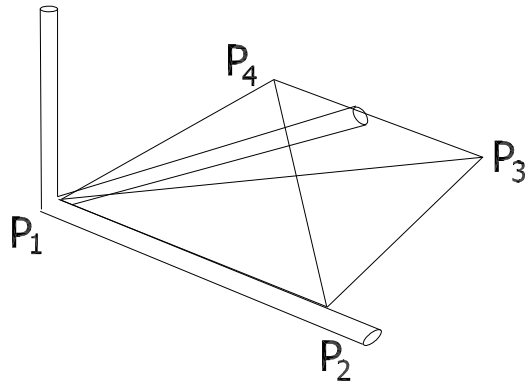


R.S. gives the relative position  
 $(xyz) \Leftrightarrow \pi(\underline{x}_1, \dots, \underline{x}_N)$

- A R.F. is the set of estimated coordinates  $(\hat{\underline{x}}_1, \hat{\underline{x}}_2, \dots, \hat{\underline{x}}_2)$  (the estimated polyhedron  $\hat{\pi}$ ) in the given R.S. i.e. physical Cartesian triad. It is a “frame” in the sense that once  $\{\hat{\underline{x}}_i\}$  are frozen, we can attach to them new points by new observations.

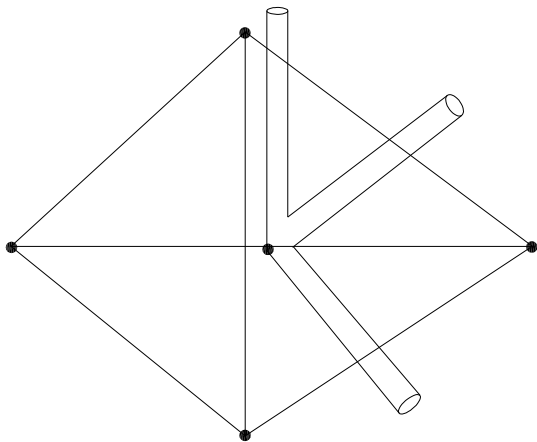
1. Is it correct to try to define first a R.S. and then a R.F.?

No: unless I use 3 points of  $\pi$  to define the R.S.



Classical R.S. attached to  $\pi$

I need first to know the coordinates of  $\underline{x}_i$ , i.e. I need to have estimates  $\hat{\pi}(\hat{\underline{x}})$  to place the coordinate triad



Example: if I want a barycentric triad  
I need first to know the coordinates of  
the points

2. Do we really need to define the triad with respect to  $\pi$  or to  $\hat{\pi}$ ?

No: we could equivalently say that we fix the triad and we try to place  $\pi$  or  $\hat{\pi}$  with respect to it!

Program for a “conceptual suite”

- a Define a R.F. i.e. the route from observations  $Y_0$  to  $\hat{\pi}(\hat{x})$
- b Define a R.S. in relation to a R.F.  
(we need an operative definition according to Einstein)
- c Define a quality index of a R.F. based on its “distance” from the R.S. and the optimal R.F

# Definition of a R.F.

$$RF \equiv \{\mathcal{O}, \mathcal{M}_o, \mathcal{M}_e, \mathcal{C}\} ;$$

- Observations: a vector  $\underline{Y}_0$  considered as a sample from an  $M$ -dimensional R.V.  $\underline{Y}$   
+ metadata on the measurement process
- Model:  $\mathcal{M}_o$ -stochastic

$$\underline{Y} = \underline{y} + \underline{\nu}$$

$f_n(\nu)$  known, or at least  $E\{\nu\} = 0$ ,  $C_\nu$  known

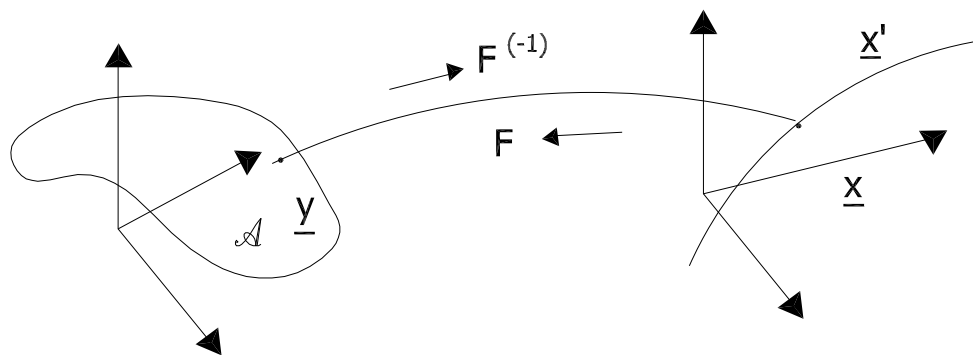
- Model:  $\mathcal{M}o$ -deterministic

$$\underline{y} = F(\underline{x}) \text{ (we ignore ancillary parameters)}$$

$\Downarrow$

- defines the manifold  $\mathcal{A}$  of admissible values
- it is rank deficient i.e.

$$\underline{x} \xrightarrow{F} \underline{y} \text{ (1 value)} , \quad \underline{y} \xrightarrow{F^{-1}} \underline{x} \text{ many values}$$



The fiber  $F$  is defined so that

$$\underline{x}' = G(\underline{x}, \underline{\vartheta}) \quad \underline{\vartheta} \text{ 6 d.f.!$$

$\Downarrow$

$$F(\underline{x}') \equiv F[G(\underline{x}, \underline{\vartheta})] \equiv G(\underline{x}), \quad \forall \underline{\vartheta}$$

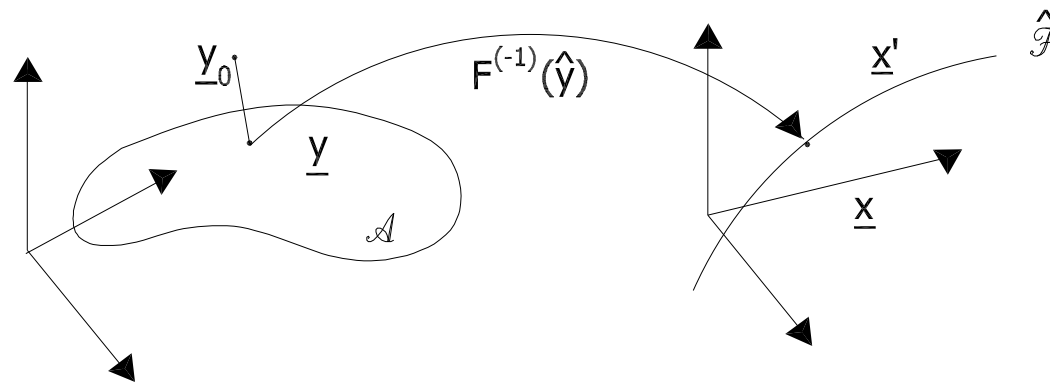
Note:  $G(\underline{x}, \underline{\vartheta}) = H(\underline{\vartheta})\underline{x}$  rototraslation! known form!



- Method: several possibilities but in Geodesy

$$\mathcal{M}e \equiv \text{L.S.}$$

Up to here we have only a path

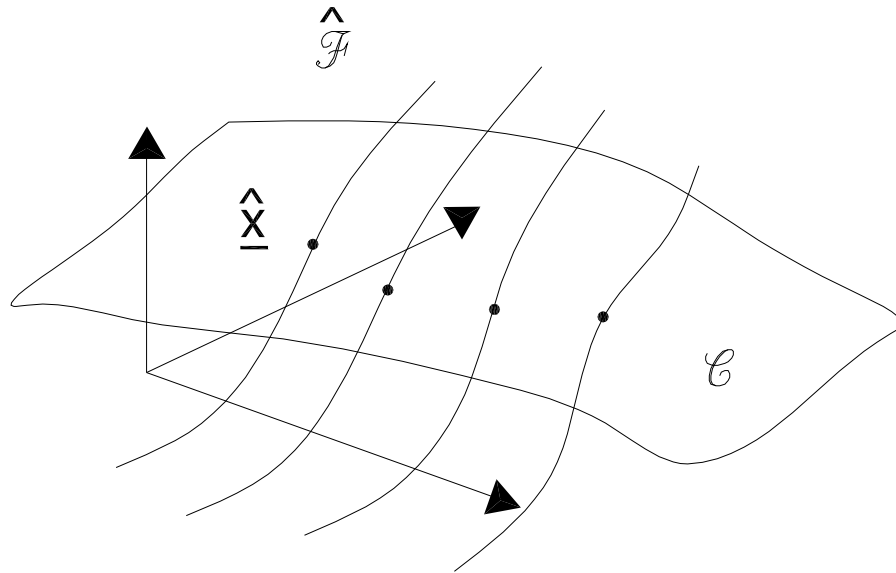


$$\min |Y_0 - \underline{\hat{y}}|_{C_v^{-1}}^2$$

$$\underline{\hat{y}} \in \mathcal{A}$$

Note:  $F^{(-1)}(S)$  is the fiber  $\hat{\mathcal{F}}$  and not a specific  $\underline{\hat{x}}$

- Constraints:  $\mathcal{C}$  to go from  $\hat{\mathcal{F}}$  to  $\hat{\underline{x}}$  by crossing all the fibers with a constraining manifold



$$\mathcal{C} \sim R^6$$

$$\mathcal{C}^{comp} \sim R^{n-6} = R^{3N-6}$$

$\mathcal{C} \equiv \{C(\underline{x}) = 0\}$  it can be put into an explicit form by splitting

$$\underline{x} = \begin{array}{l} \left| \begin{array}{l} \underline{x}_0 \\ \underline{x}_c \end{array} \right| \begin{array}{l} \rightarrow 3N - 6 \\ \rightarrow 6 \text{ suitably chosen coordinates} \end{array} \end{array}$$

$$\underline{x}_c = K(\underline{x}_0)$$

Note: the correspondence

$$\mathcal{A} = \underline{y} = F[\underline{x}_0, K(\underline{x}_0)] \quad \underline{x}_0 \in R^{n-6}$$

is now 1 to 1, i.e. the matrix

$$J = F_{\underline{x}_0} + F_{\underline{x}_c} K_{x_0} \text{ is of full rank}$$

- the choice  $K_{\underline{x}_0} \equiv 0$  is always possible
- the matrix  $F_{\underline{x}_0}$  alone spans the tangent manifold to  $\mathcal{A}$  so  $F_{\underline{x}_0}$  is of full rank and

$$F_{\underline{x}_c} = F_{\underline{x}_0} \Gamma \text{ for some } \Gamma .$$

i. e.

$$J = F_{\underline{x}_0} (I + \Gamma K_{x_0}) \quad ((I + \Gamma K_{\underline{x}_0}) \text{ regular: transversality condition})$$

So with our definition of R.F. we are able now to solve the L.S. problem

$$\hat{\underline{x}}_0 = \arg \min \| Y_{-0} - F[\underline{x}_0, K(\underline{x}_0)] \|_{C_\nu^{-1}}$$

and compute

$$\hat{\underline{x}}_c = K(\hat{\underline{x}}_0) .$$

Note that if  $\hat{\underline{x}}_0$  has covariance matrix

$$C_0 = C_{\hat{\underline{x}}_0} = (J^T C_\nu^{-1} J)^{-1}$$

then the full vector  $\hat{\underline{x}}$  has covariance matrix

$$C_{\hat{\underline{x}}} = \left| \begin{array}{cc} C_0 & C_0 K_{\underline{x}_0}^T \\ K_{\underline{x}_0} C_0 & K_{\underline{x}_0} C_0 K_{\underline{x}_0}^T \end{array} \right| \quad (\text{Note: this is singular})$$

Note: we could define a reduced  $RF \equiv \{\hat{\underline{x}}, C_{\hat{\underline{x}}}\}$  but never  $\hat{\underline{x}}$  alone!

## Definition of R.S.

Assume you have  $R$  repetitions  $\underline{Y}_{or}$  ,  $r = 1 \dots R$ , of the same observations. Then by using the strong Law of Large Numbers maintaining that

$$\frac{1}{R} \sum_{r=1}^R \underline{Y}_{or} \rightarrow \underline{y} \in \mathcal{A} \text{ almost surely (a.s.),}$$

we easily prove that for the L.S. solution  $\hat{\underline{x}} = \arg \min \sum_{r=1}^R |\underline{Y}_{or} - F(\underline{x})|_{C_\nu}^{-2}$

the limit  $\lim_{R \rightarrow \infty} \hat{\underline{x}} = \underline{x}$  holds a.s. even for non-linear models. So we can define a R.S. as the body  $\pi = \lim_{R \rightarrow \infty} \hat{\pi}$  which attains now a precise position in our coordinate triad!

Note: restriction to linear models/linear constraints.

In this case

$$F(\underline{x}) = A\underline{x} = A_0\underline{x}_0 + A_c\underline{x}_c \quad A_c = A_0\Gamma$$

$$\underline{x}_c = K(\underline{x}_0) = K\underline{x}_0 \quad K_{\underline{x}_0} = K$$

$$J = F_{\underline{x}_0} + F_{\underline{x}_c}K_{x_0} = A_0 + A_cK$$

$$C_0 = [(A_0 + A_cK)^T C_\nu^{-1} (A_0 + A_cK)]^{-1}$$

Moreover, in this case L.S. give an unbiased estimate

$$E\{\hat{\underline{x}}\} = \underline{x}$$

$$(\underline{x}_c = K\underline{x})$$

# Quality index

Restricted to linear models, an obvious quality index is

$$\begin{aligned} Q &= E\{|\hat{\underline{x}} - \underline{x}|^2\} = \text{Tr}C_{\hat{\underline{x}}} = \\ &= \text{Tr}C_0 + \text{Tr}KC_0K^T = \\ &= \text{Tr}C_0(I + K^TK) \end{aligned}$$

so we can search for the best R.F. as the one satisfying constraints (i.e.  $K$ ) such that  $Q$  is minimized.

Optimal R.F.: the variational equation is

$$\frac{1}{2}\delta Q = \text{Tr}\delta_{K^T}C_0(I + K^T K) + \text{Tr}C_0\delta K^T K$$

now

$$\delta_{K^T}C_0 = -C_0\delta K^T A_c^T C_\nu^{-1}(A_0 + A_c K)C_0$$

Putting to 0 the first variation we get

$$-A_c^T C_\nu^{-1}(A_0 + A_c K)C_0(I + K^T K)C_0 + KC_0 = 0;$$

even simplifying  $C_0$  from the right a really horrible question!

However

$$A_c = A_0\Gamma, \quad J = A_0 + A_c K = A_0(I + \Gamma K)$$

$$C_0 = (I + \Gamma K)^{-1}(A_0^T C_\nu^{-1} A_0)^{-1}(I + K^T \Gamma^T)^{-1}$$



Substituting and simplifying we get the notable equation

$$K = \Gamma^T (I + K^T \Gamma^T)^{-1} (I + K^T K)$$

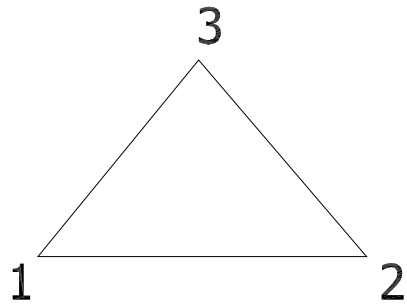
I don't know its general solution, however one solution is certainly

$$\boxed{K = \Gamma^T}$$

Is this the wanted minimum?

I cannot answer but I can show by an elementary example that this is what we are after!

## Example



The leveling triangle with

$$\underline{Y} = \begin{vmatrix} Y_1 \\ Y_2 \\ Y_3 \end{vmatrix} = \begin{vmatrix} q_2 - q_1 \\ q_3 - q_2 \\ q_1 - q_3 \end{vmatrix} = \begin{bmatrix} -1 & 1 & 0 \\ 0 & -1 & 1 \\ 1 & 0 & -1 \end{bmatrix} \underline{x}; \quad C_\nu = I$$

Put  $A_0 = \begin{bmatrix} -1 & 1 \\ 0 & -1 \\ 1 & 0 \end{bmatrix} \quad A_c \left| \begin{array}{c} 0 \\ 1 \\ -1 \end{array} \right|$  so that

$$A_c = -A_0 \left| \begin{array}{c} 1 \\ 1 \end{array} \right| \text{ i.e. } \Gamma = - \left| \begin{array}{c} 1 \\ 1 \end{array} \right|, \Gamma^T = -[1 \ 1]$$

But then our optimal frame is

$$x_c = q_3 = \Gamma^T \underline{x}_0 = -(q_1 + q_2) \Rightarrow q_1 + q_2 + q_3 = 0$$

the barycentric frame as everybody would expect!