

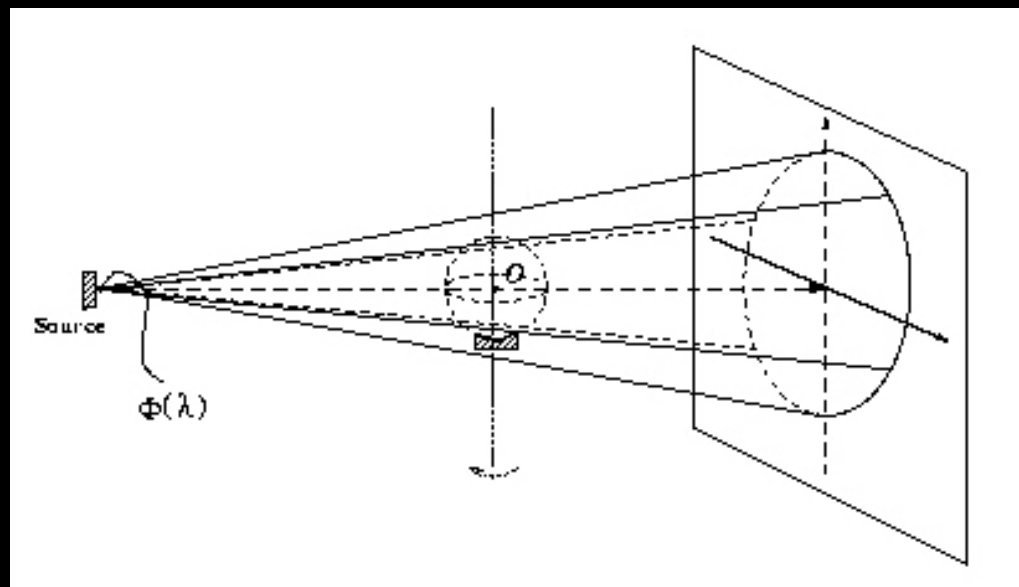
*Cone-beam image reconstruction  
by moving frames*

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## *Cone-beam Imaging Apparatus*

- Radiation source and area detector
- Source-detector rotate around the object
- Collect integrals of density along straight lines when source travels along a 3D curve





## *The problem*

- Cone-beam reconstruction:
  - to recover a 3D density function from its integrals along a set of lines emitting from a 3D curve
- Among the first problems in integral geometry



## *Difficulties*

- Inverse calculation deals with curved spaces
- Algorithmic implementation encounters:
  - Large dataset
  - sophisticated geometric mapping
  - expensive data interpolations

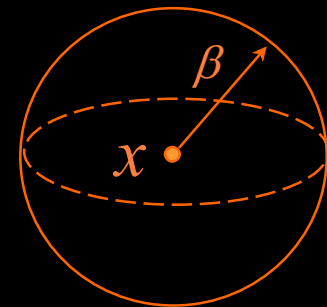
## Radon's formula

- Radon transform: integral over planes

$$Rf(l, \beta) := \int_{x \in \{x | x \cdot \beta = l\}} f(x) dx$$

- 3D Radon inverse (1917):

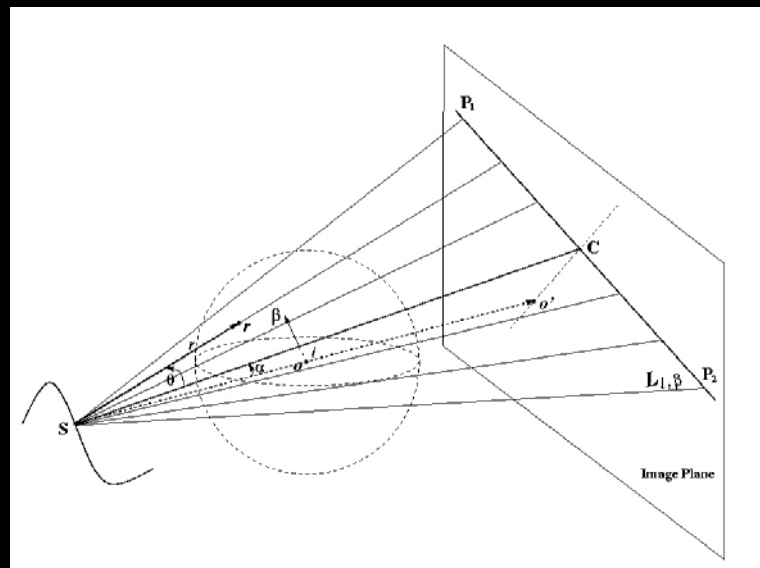
$$f(x) = -\frac{1}{8\pi} \int_{S^2} \left. \frac{\partial^2 Rf(l, \beta)}{\partial l^2} \right|_{l=x \cdot \beta} d\beta$$



Backprojection sphere

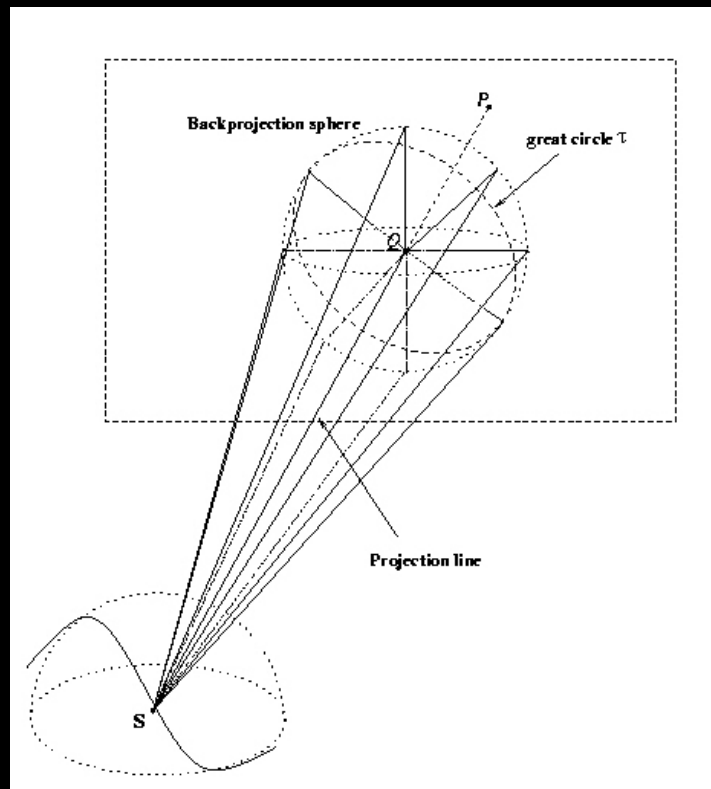
## *Grangeat's "Fundamental Relation"*

- Link cone-beam data to the first-order radial derivative of the Radon transform (1991)

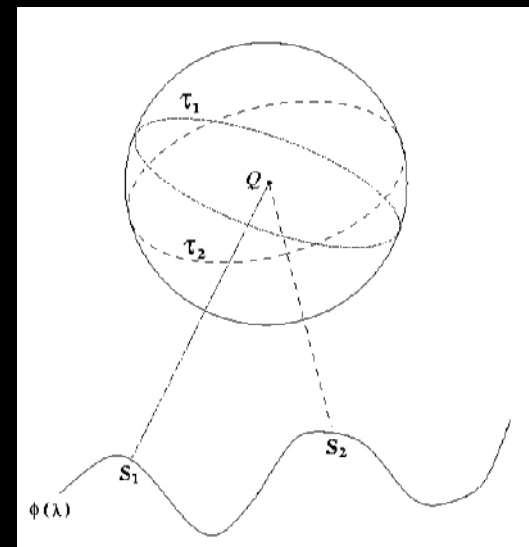


# Geometric constraints: within and across projections

- 3D to 2D reduction



- Great circle under a rigid rotation



# Formulae by Yang and Katsevich

- Yang (2002)

$$f(x) = -\frac{1}{8\pi} \int_{\Lambda} \left\{ \int_{\substack{\beta \in \{x - \Phi(\lambda)\}^\perp \\ \beta \in S^2}} R'' f(\Phi(\lambda) \cdot \beta, \beta) \frac{|\Phi'(\lambda) \cdot \beta|}{M(\lambda, \beta)} d\beta \right\} d\lambda$$

along curve
within projection
across projection

- Katsevich (2003)

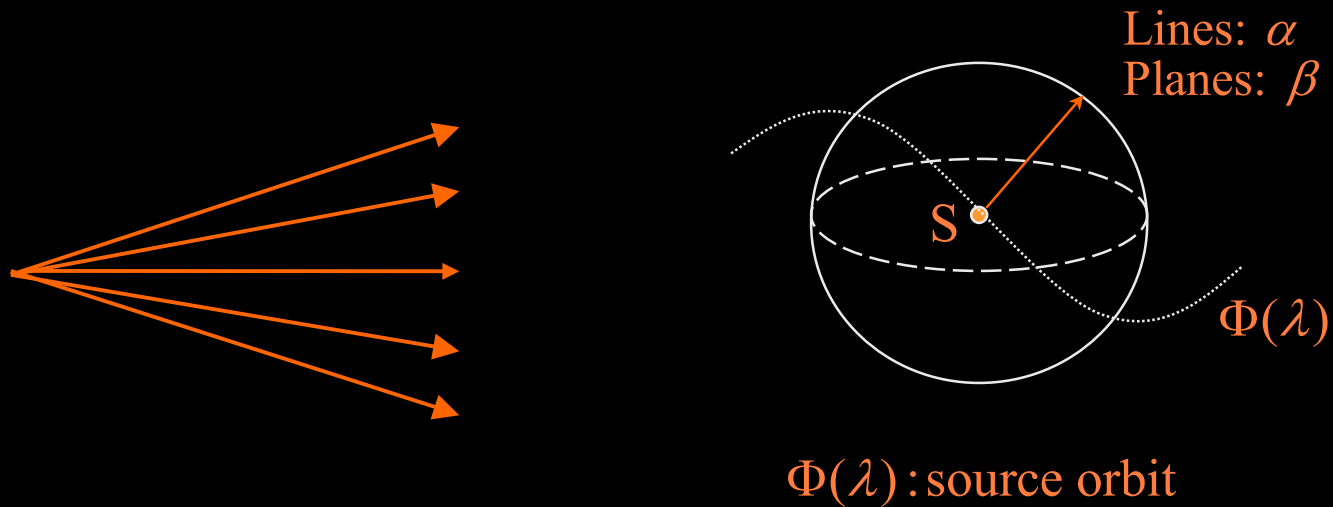
$$f(x) = -\frac{1}{8\pi} \int_{\Lambda} \frac{\sum c_m(\lambda, x)}{|x - \Phi(\lambda)|} \int_0^{2\pi} \frac{\partial}{\partial \lambda'} Df(\Phi(\lambda'), \cos \gamma \alpha(\lambda, x) + \sin \gamma \alpha^\perp(\lambda, x, \theta_m)) \Big|_{\lambda'=\lambda} \frac{d\gamma}{\sin \gamma} d\lambda$$

along curve
within projection
across projection



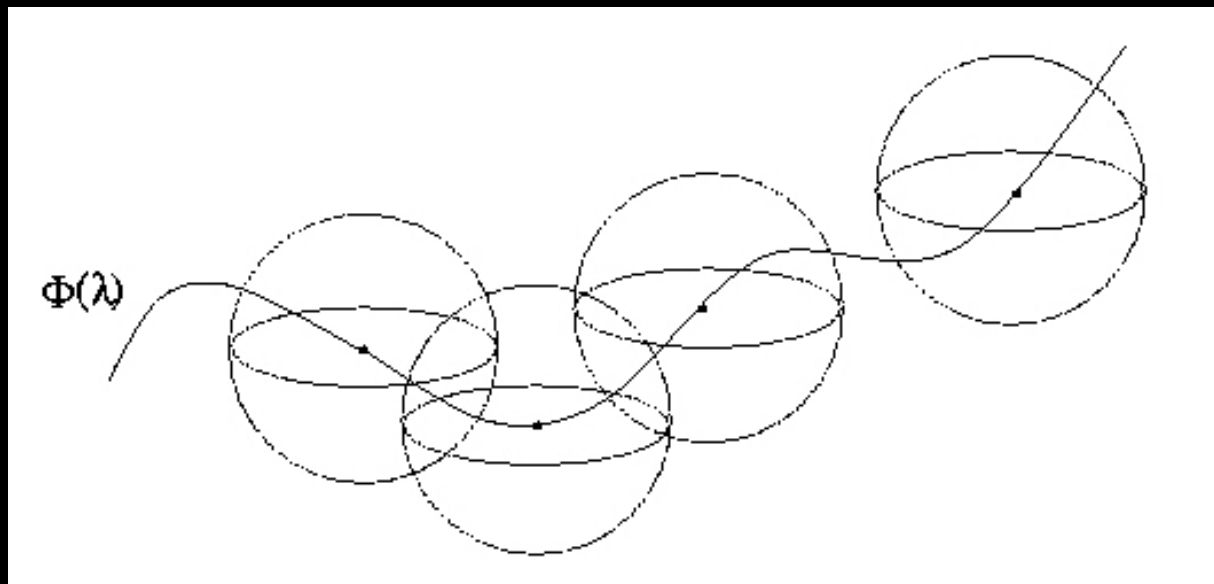
# *Forward projection and spherical space*

- Forward projection maps  $\mathbb{E}^3$  to  $\mathbb{P}^2$



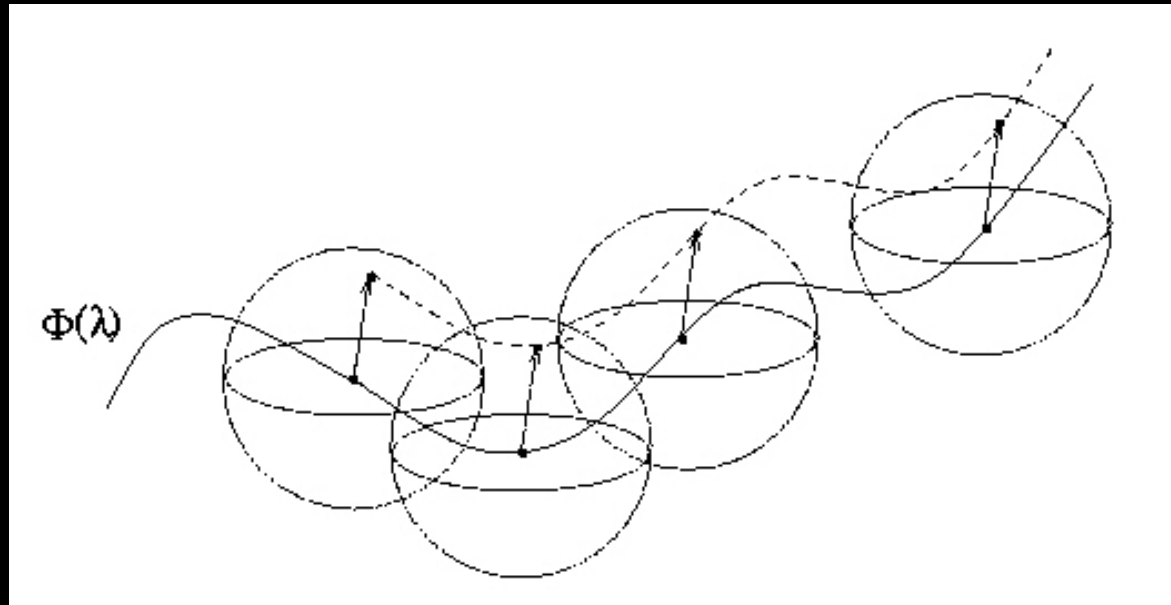
## *Fiber bundle structure*

- Attach to each source point a unit sphere which represents a local fiber
- Fiber bundle: the union of all the spheres



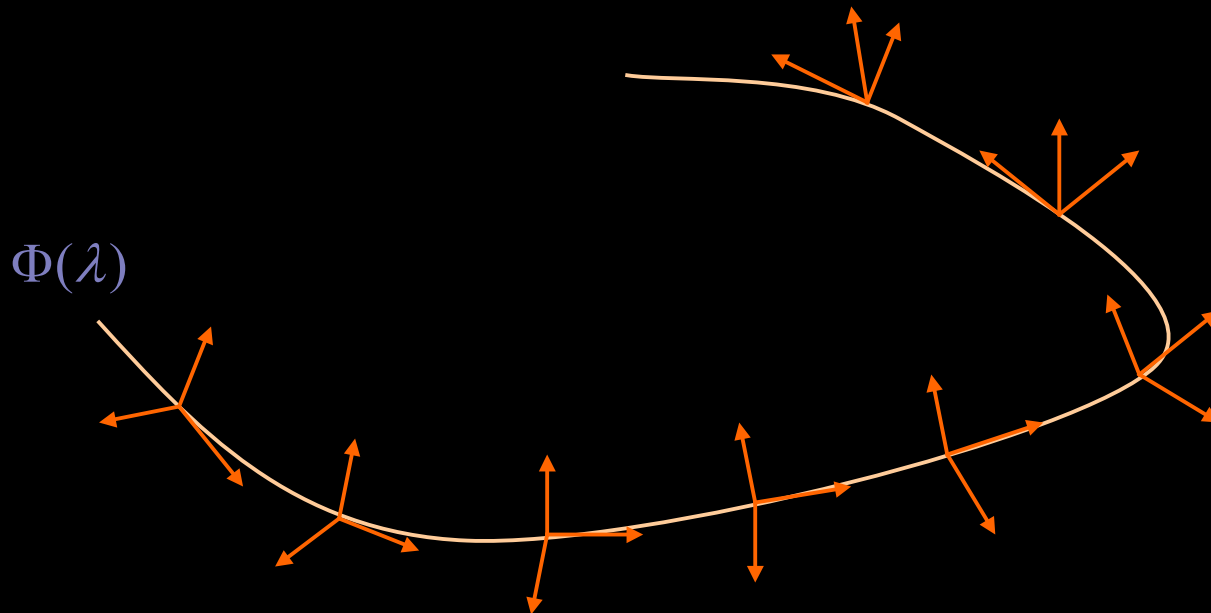
## *Calculations within and across projections*

- Within projection: calculations within each fiber require only local coordinates
- Across projections: differentiation along curves on the fiber bundle requires global coordinates



## *Euclidean moving frames*

- At each source point attach an orthonormal Euclidean basis, i.e.,

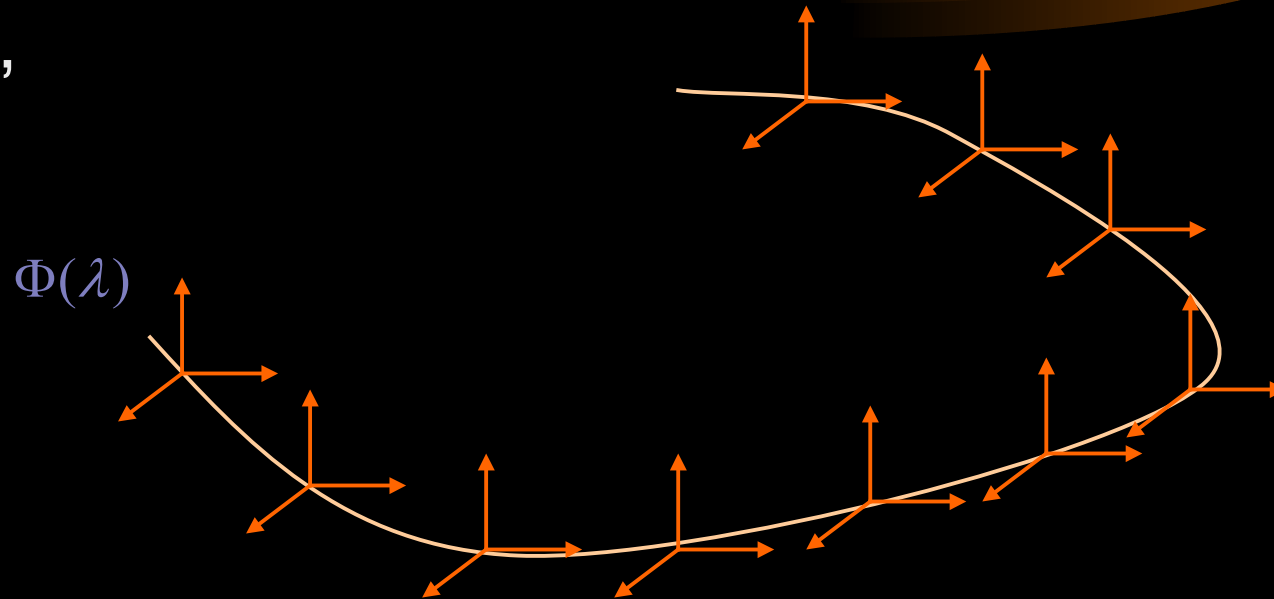


## *Euclidean moving frames (Cont'd)*

- Allows easy exchange between local and global coordinates of points, lines and planes
- Allows treating the non-Euclidean space, “*locally*”, as an Euclidean space equipped with Euclidean-like coordinates

## *Selection of moving frame basis*

- Many choices in selecting moving frame bases, i.e.,

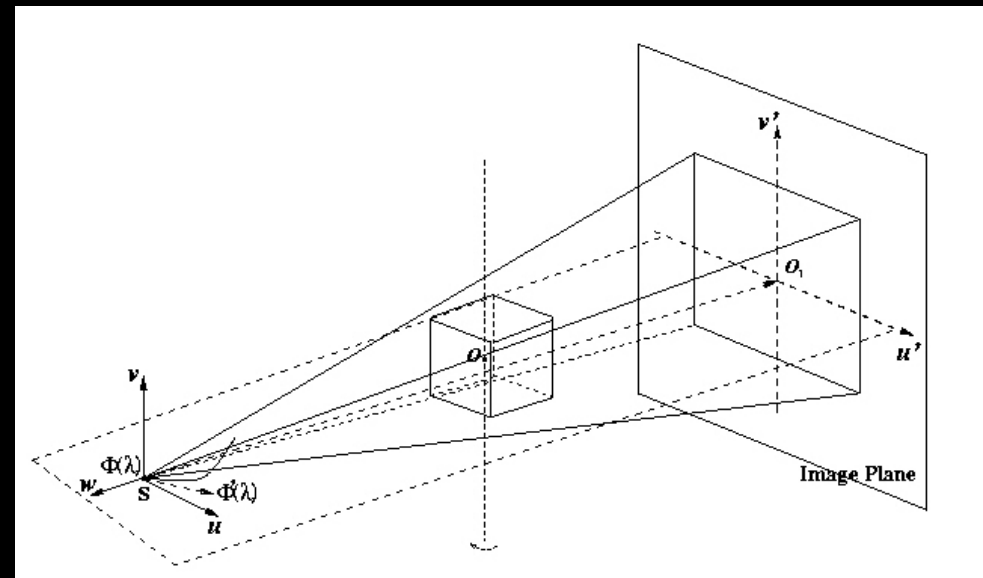


- Main considerations:  
simplify coordinate computation, ease system alignment

## Moving frame basis (I)

- Under cylindrical symmetry:

$$\left\{ \begin{array}{l} u = \frac{(-\phi_2(\lambda), \phi_1(\lambda), 0)}{\sqrt{\phi_1^2(\lambda) + \phi_2^2(\lambda)}} \\ v = (0, 0, 1) \\ w = \frac{(\phi_1(\lambda), \phi_2(\lambda), 0)}{\sqrt{\phi_1^2(\lambda) + \phi_2^2(\lambda)}} \end{array} \right.$$

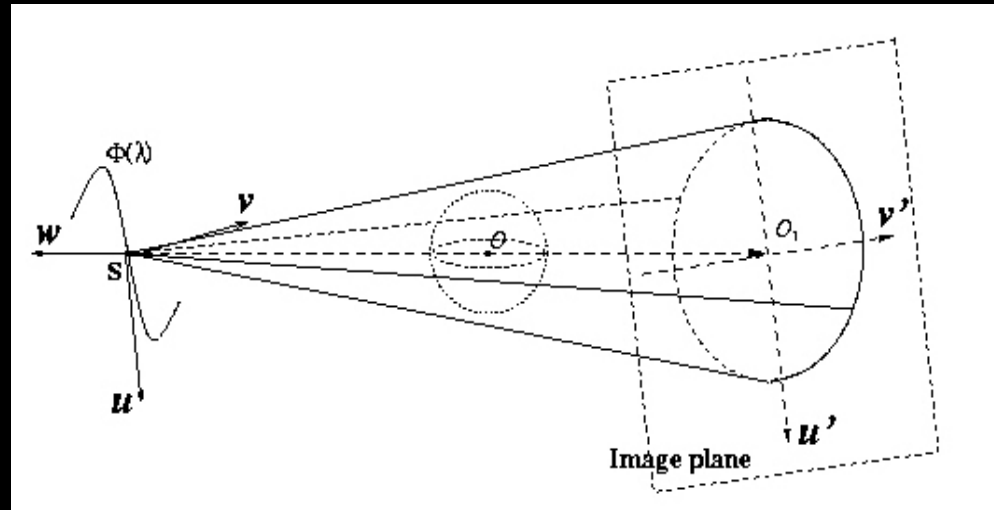


Good selection of moving frames simplifies the geometric computation as well as system alignment.

## Moving frame basis (II)

- Under spherical symmetry:

$$\left\{ \begin{array}{l} w = \frac{\Phi(\lambda)}{|\Phi(\lambda)|} \\ v = \frac{w \times \Phi'(\lambda)}{|w \times \Phi'(\lambda)|} \\ u = v \times w \end{array} \right.$$

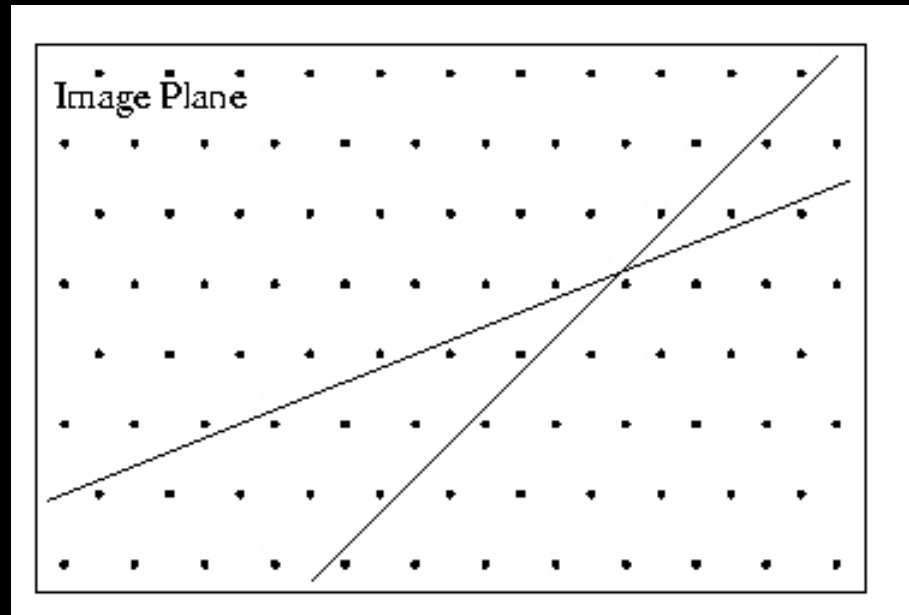


There is an alignment step in system design to align the axes of the detectors to two of the axes of the moving frames.



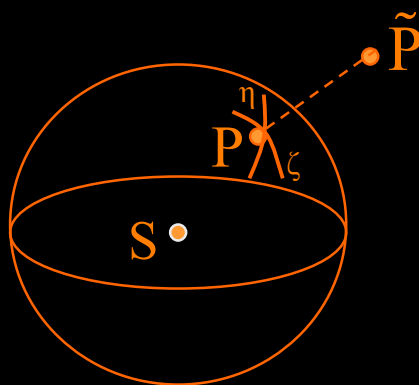
## *Integral within projection*

- Integration in the fiber space
- Local coordinate and discretization
- Irregularity in sampling



## *Exterior differentiation: differentiation across projections*

- Two fibers are disjoint excepts at a zero measure set
- Differentiation across fibers can't be replaced by differentiation within the closed fiber



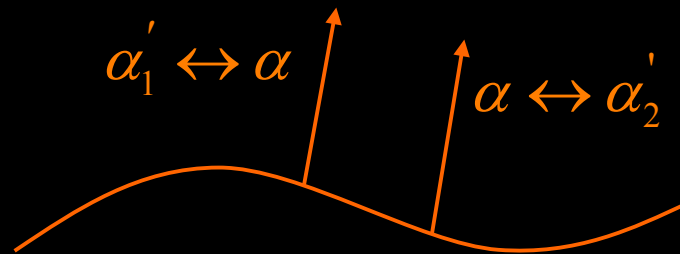
## Exterior derivative (Cont'd)

- Parallel lines: having the same  $\alpha$
- The **rotational matrix** is made up of the three orthonormal basis vectors:

$$R(\lambda) = (u(\lambda), v(\lambda), w(\lambda))$$

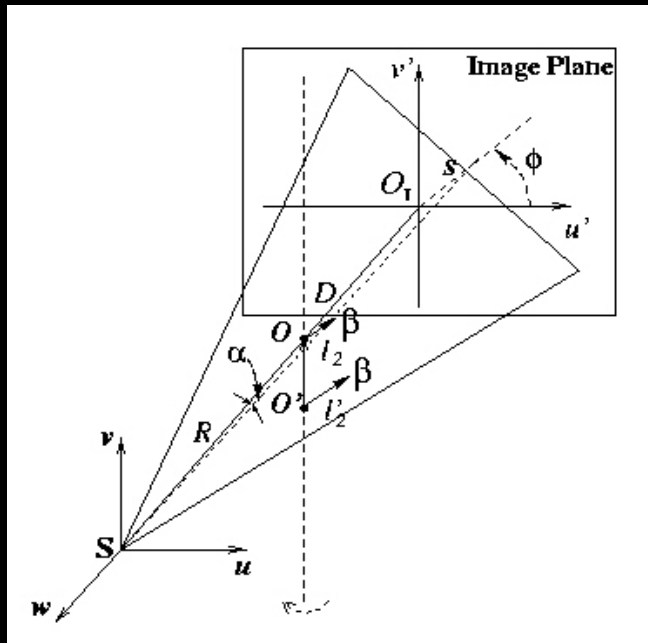
- Local  $\rightarrow$  global  $\rightarrow$  local coordinate transforms:

$$\alpha = R(\lambda_1)\alpha'_1$$
$$\alpha'_2 = R(\lambda_2)^{-1}\alpha$$



## Exterior derivative (Cont'd)

- Parallel planes: having the same  $\beta$
- Local  $\rightarrow$  global  $\rightarrow$  local coordinate transforms
- Local coordinate of the intersection line  $(s, \phi)$



$$\begin{cases} s = D \frac{l_2}{\sqrt{R^2 - l_2^2}} \\ \phi = \arctan \left( \frac{\beta \cdot v}{\beta \cdot u} \right) \end{cases}$$



## *Summary*

- New geometric representations
- A new “method of moving frames” applied to cone-beam reconstruction (a new computational framework)
- 3D discretization of the curved transform space, applicable to all cone-beam geometries
- Methods to compute the exterior derivatives
- Further study on sampling/interpolation schemes to improve algorithm efficiency