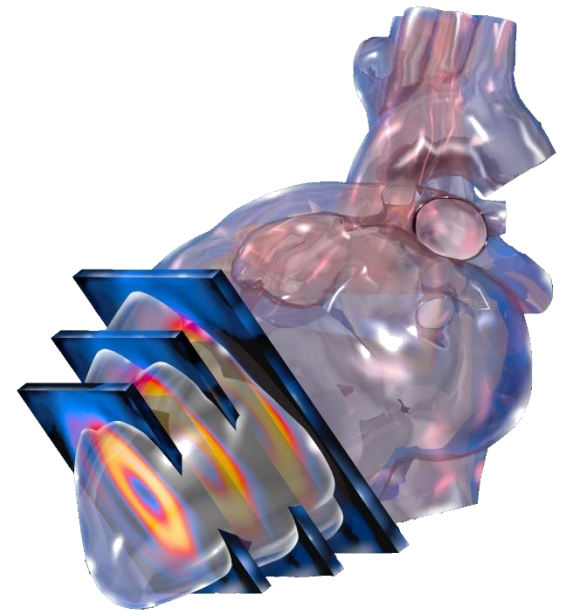


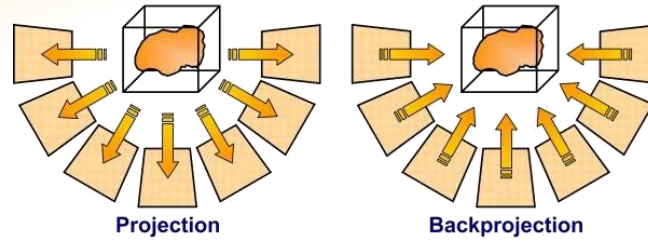
MASTER PhyMed

GMPH308 - Physique de l'imagerie médicale

TOMOGRAPHIE D'EMISSION



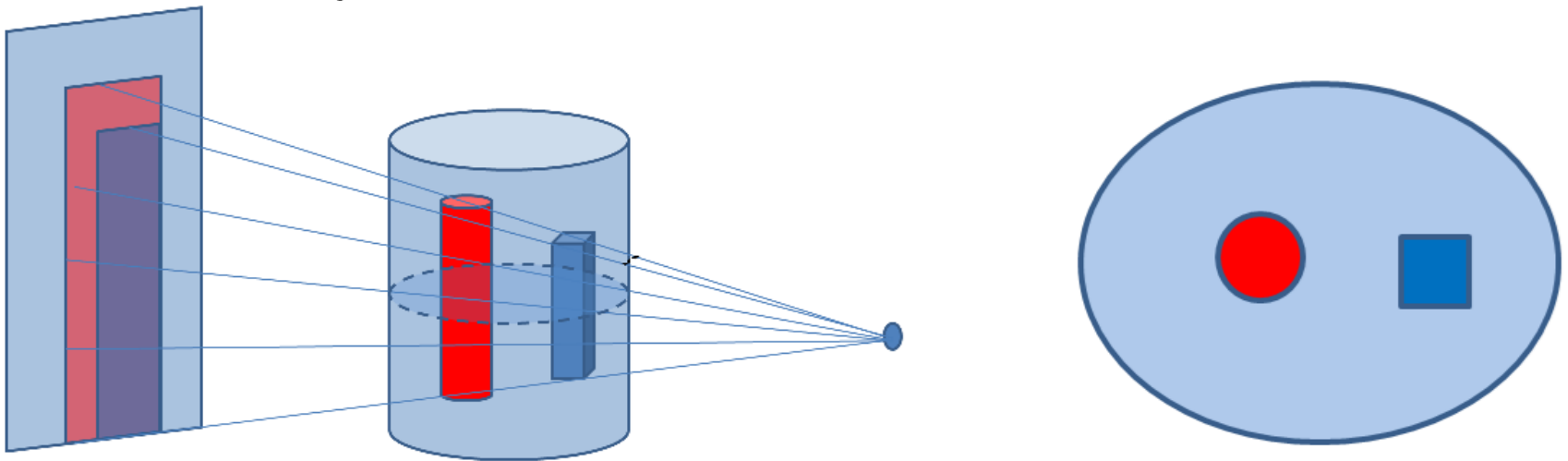
Tomographie



Projections

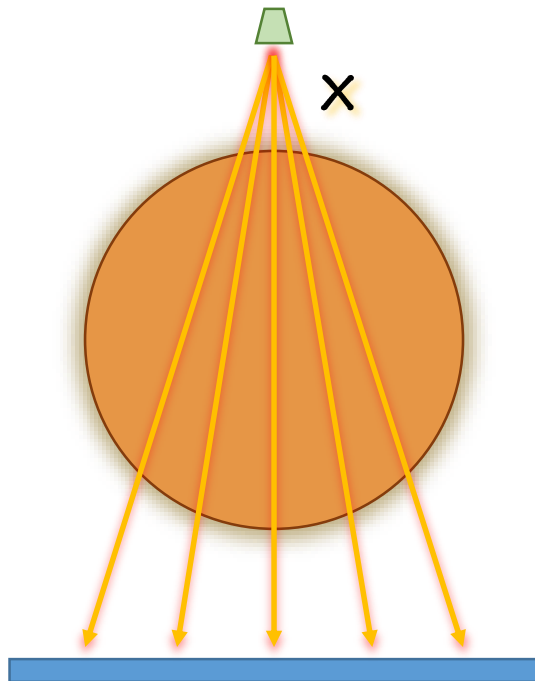


Coupes

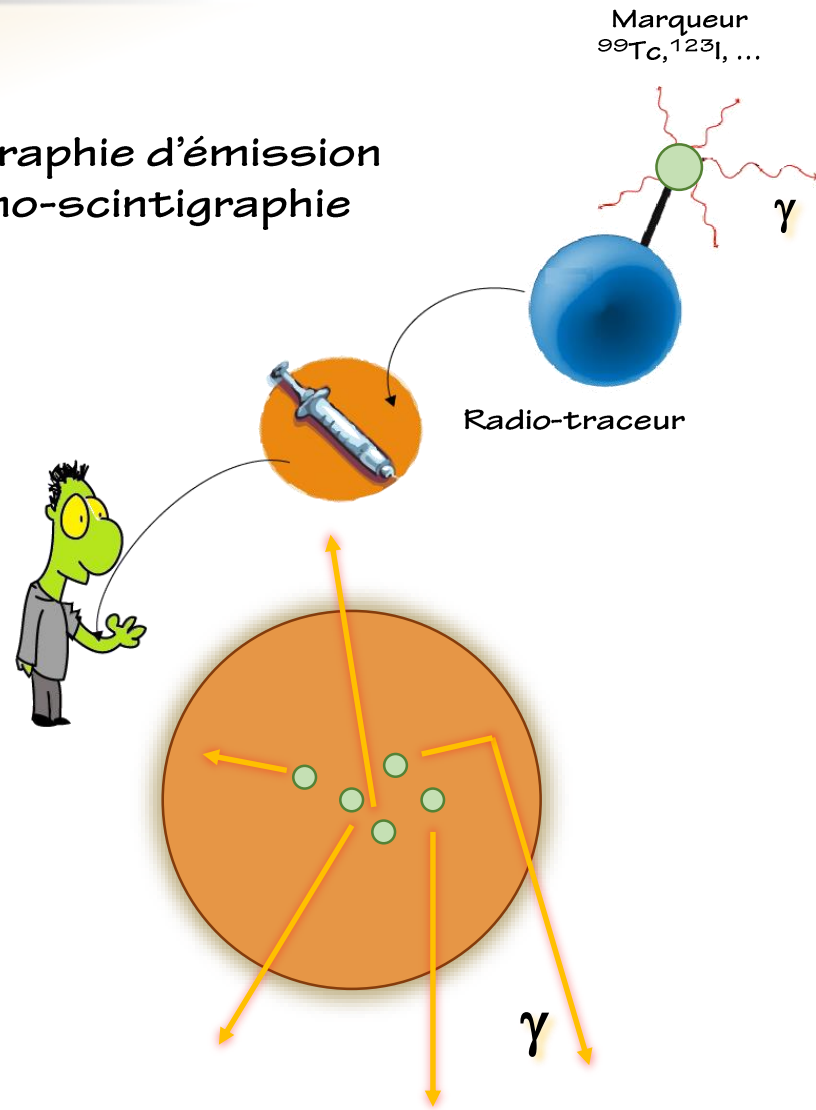


Tomographie

Tomographie de transmission



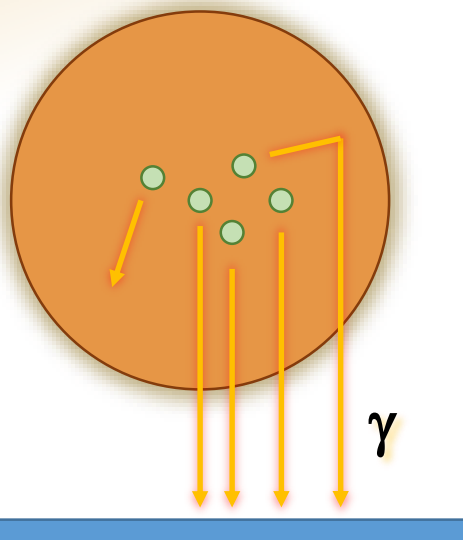
Tomographie d'émission = tomo-scintigraphie



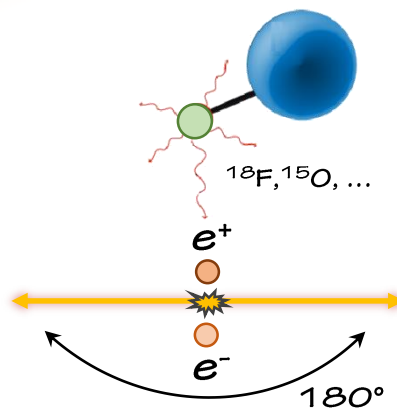
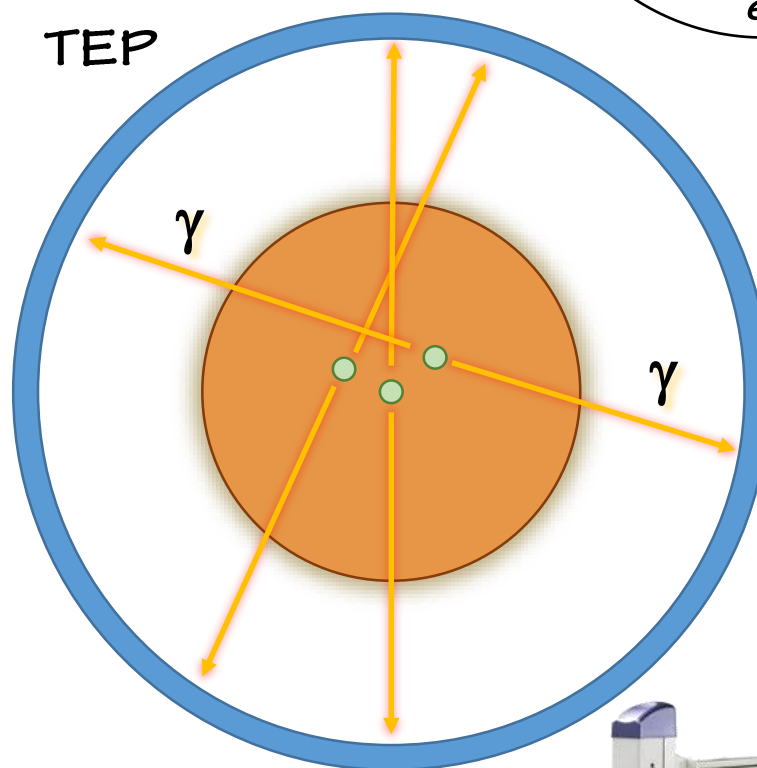
Tomographie

Tomographie d'émission

TEMP = SPECT

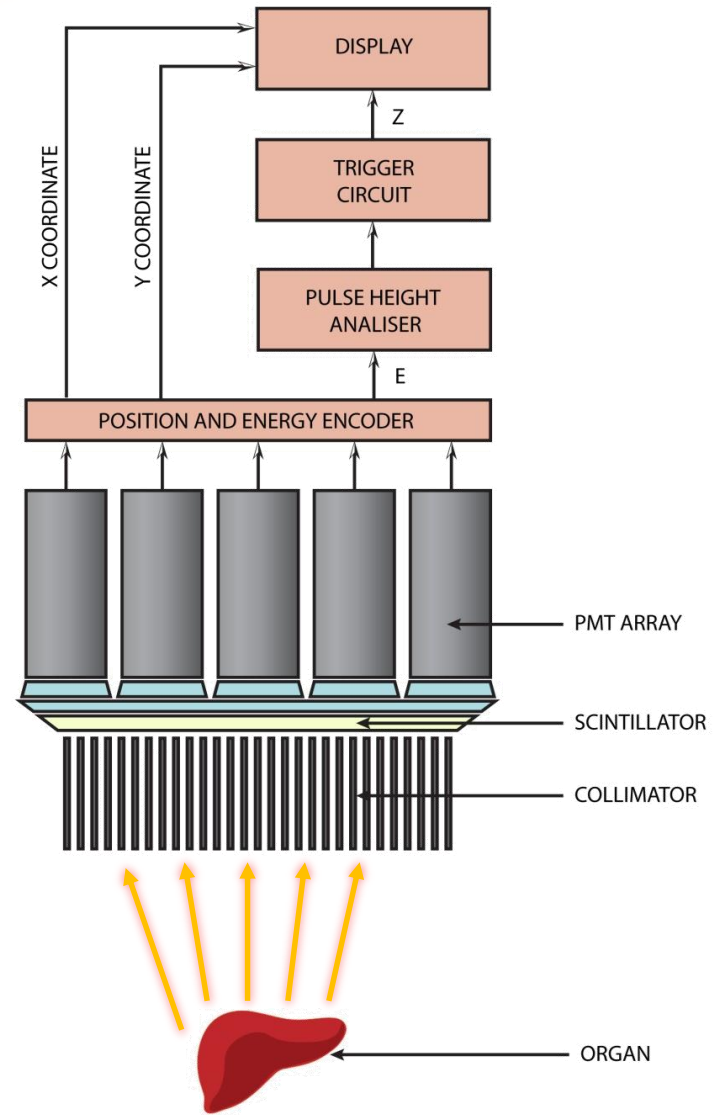
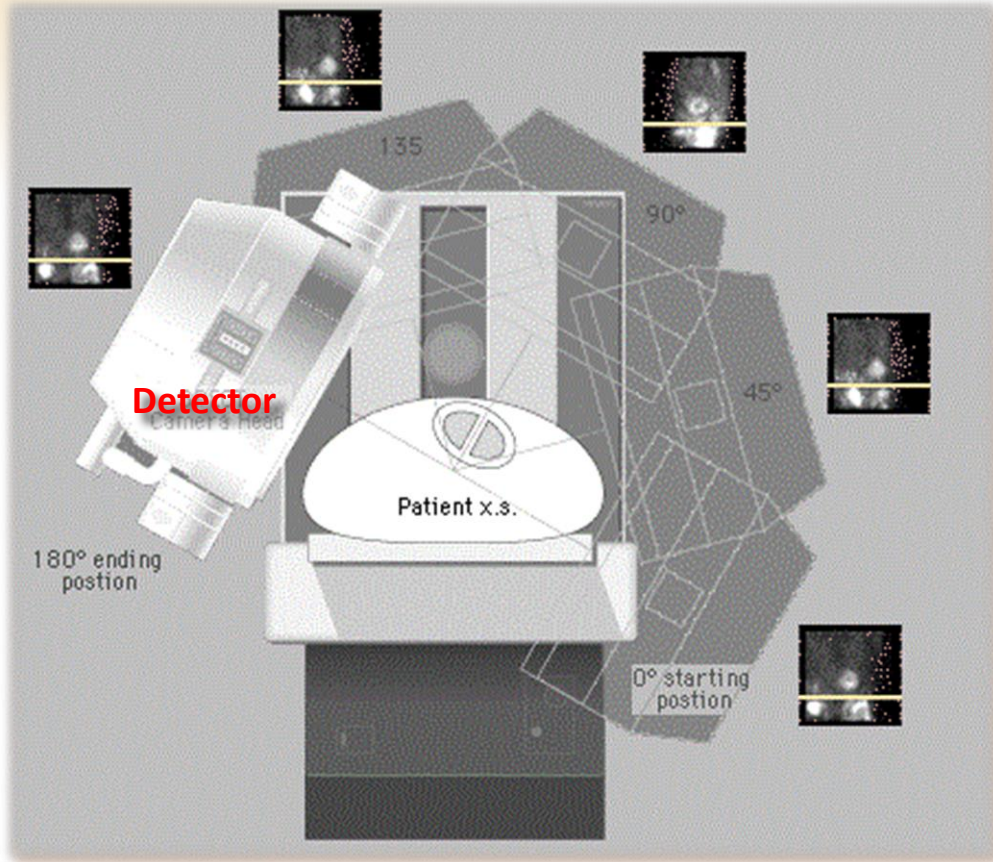


TEP



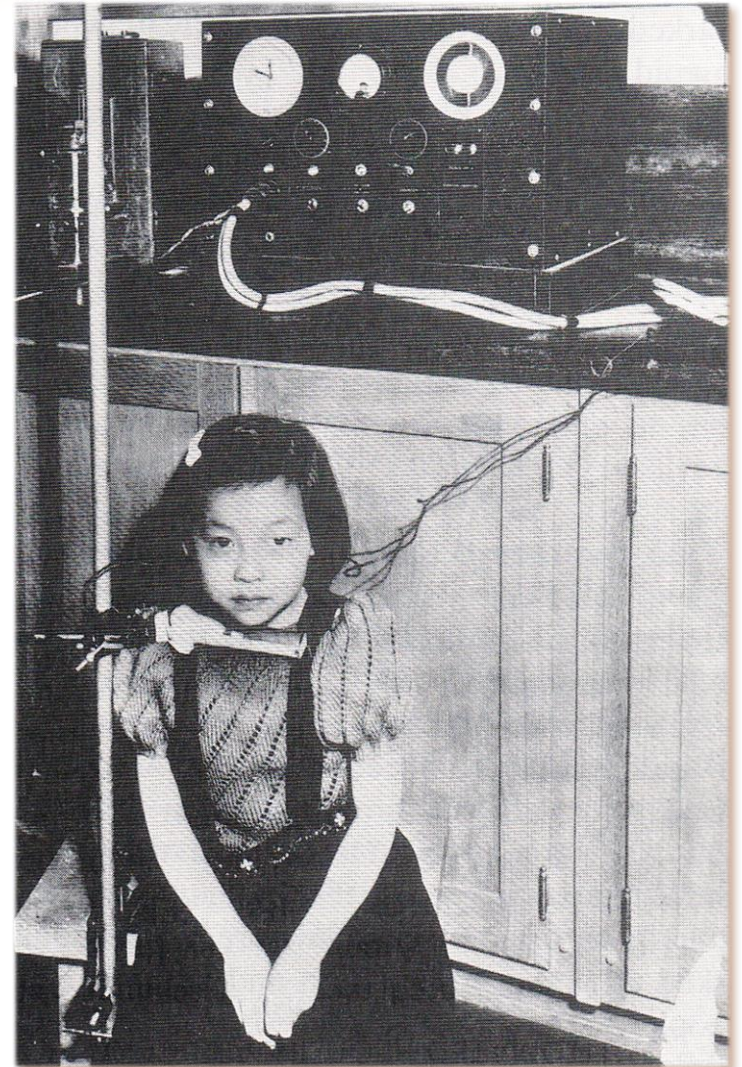
TEMP - acquisition

Gamma camera d'Anger



TEMP - acquisition

*Gamma caméra d'Anger
(1958)*



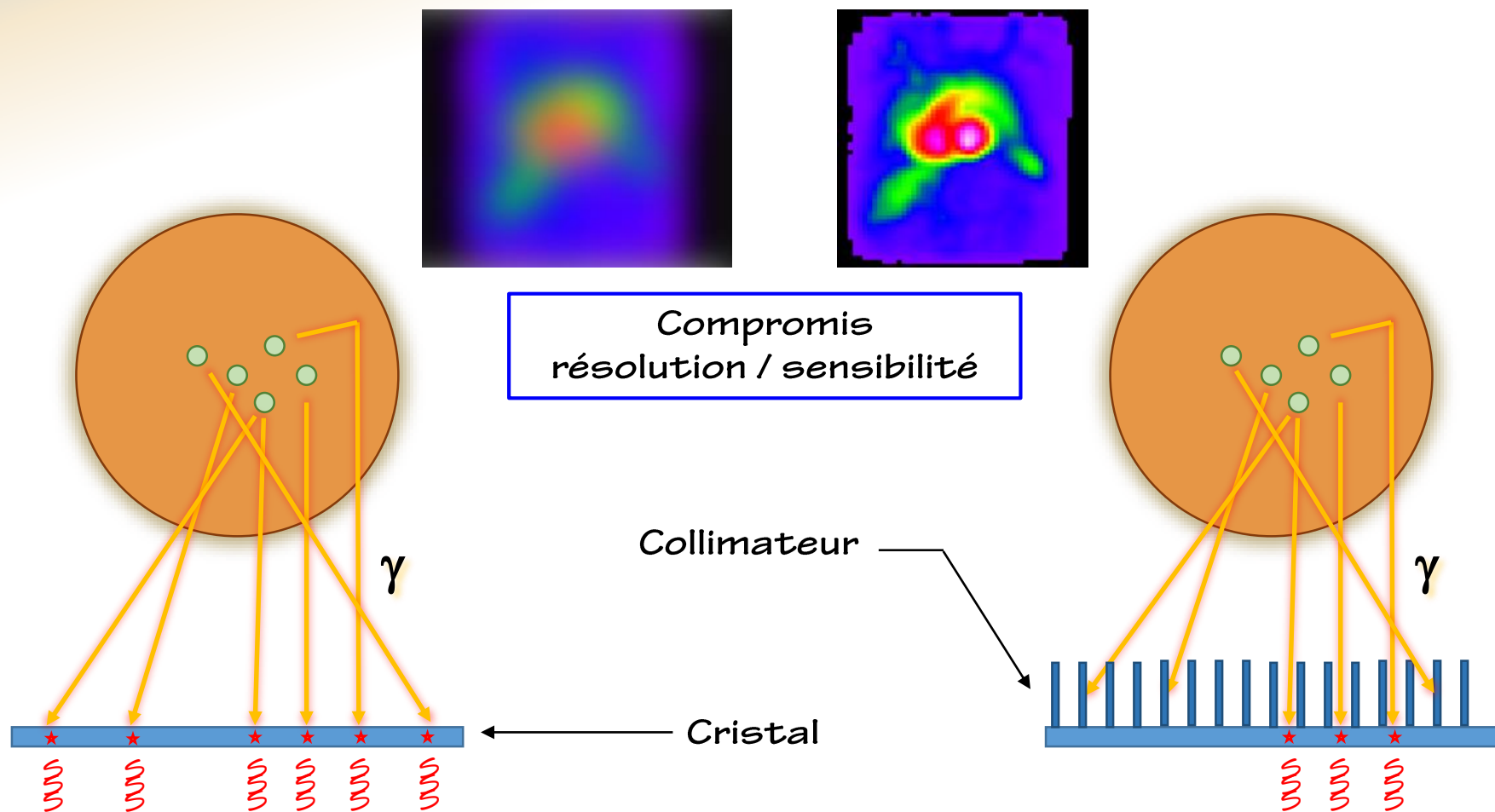
TEMP - acquisition

Gamma caméra d'Anger
(2010)



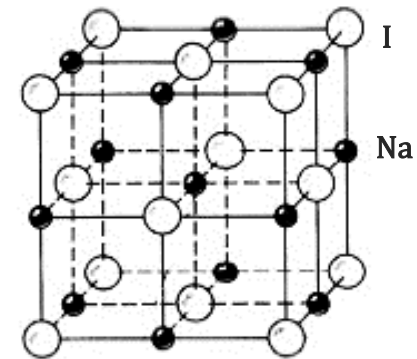
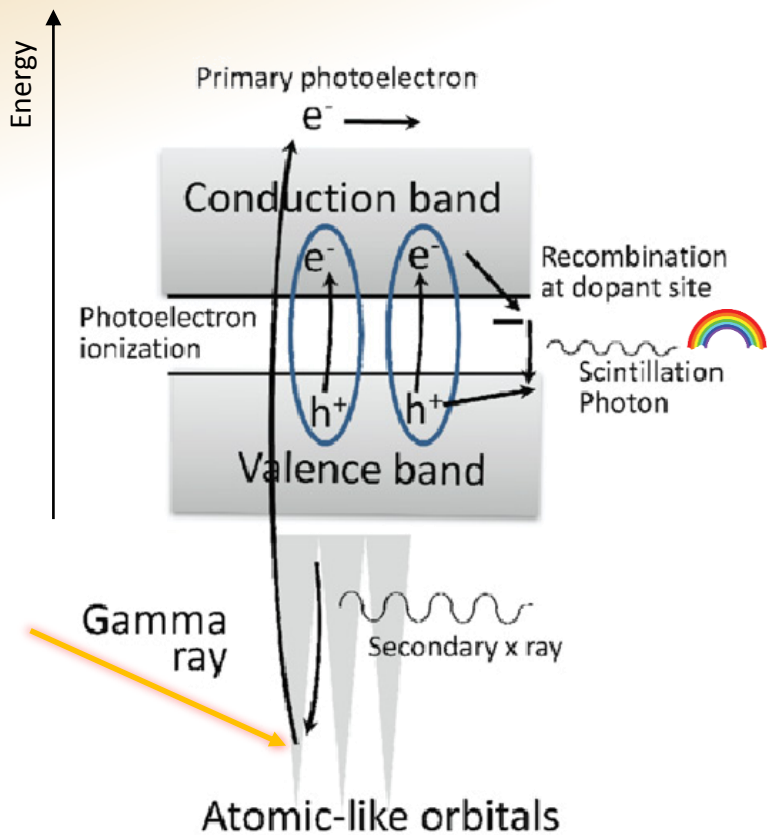
TEMP - acquisition

■ Collimation

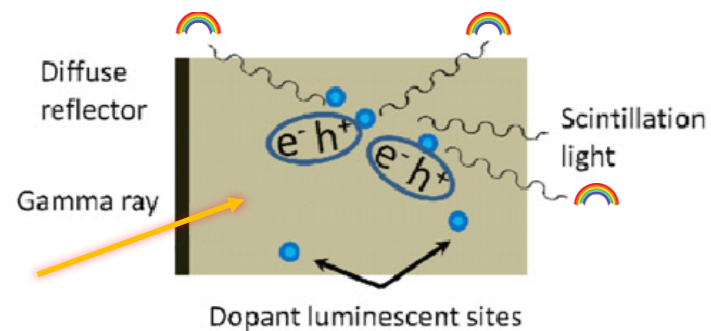


TEMP - acquisition

■ Scintillation

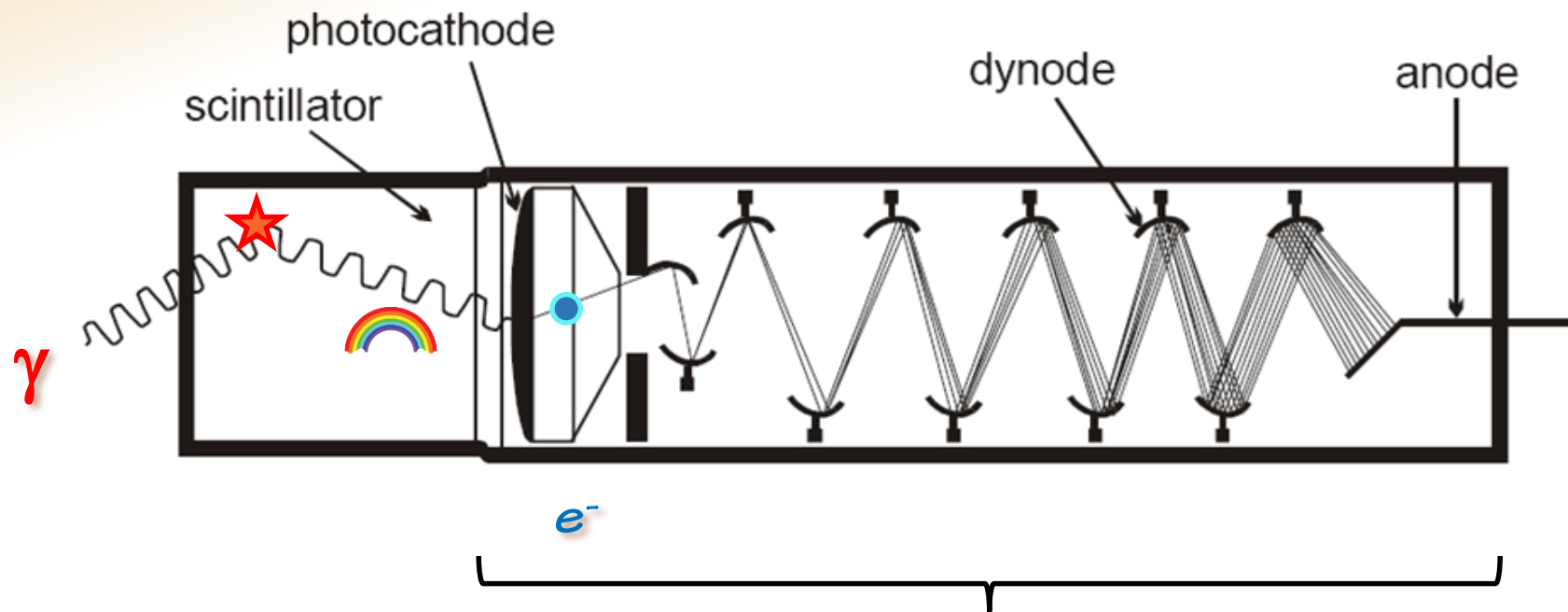


NaI(Tl)



TEMP - acquisition

■ Amplification



photomultiplier tube

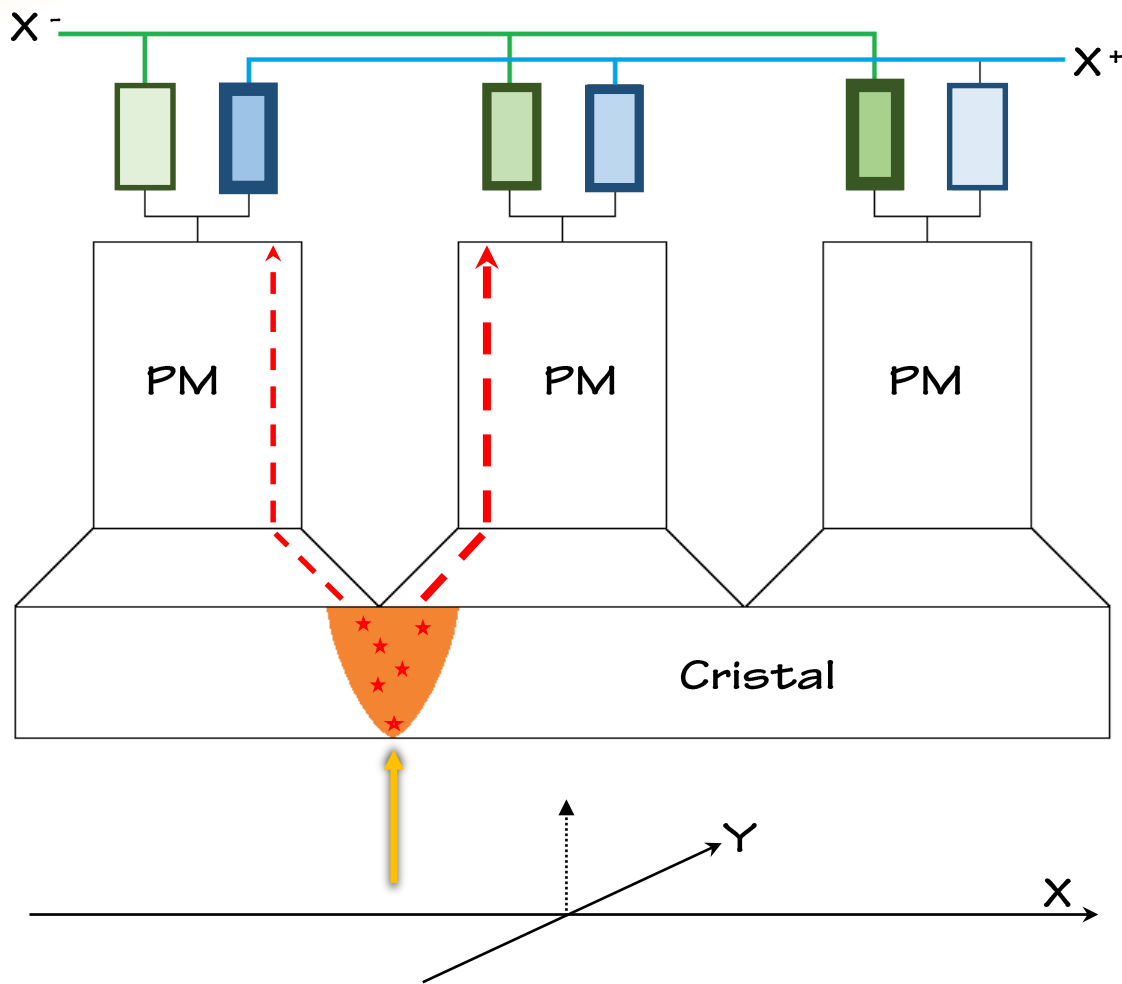
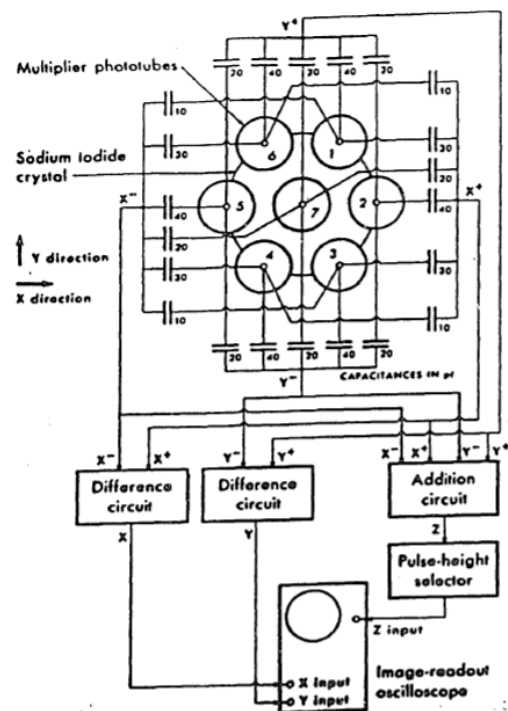
TEMP - acquisition

■ Localisation

$$X = X^+ - X^-$$

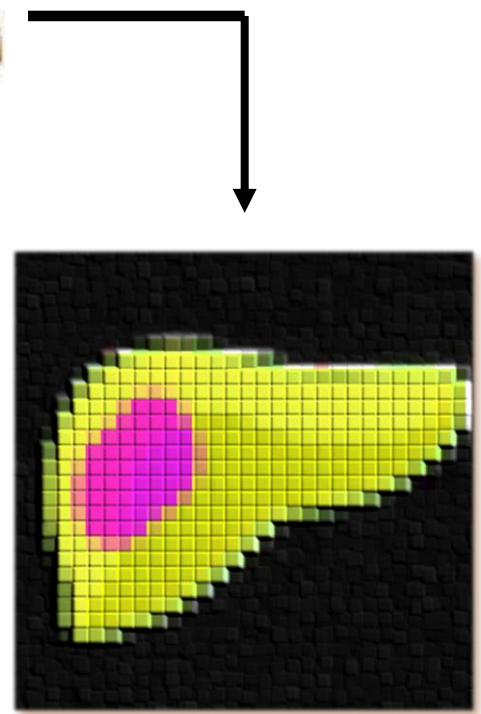
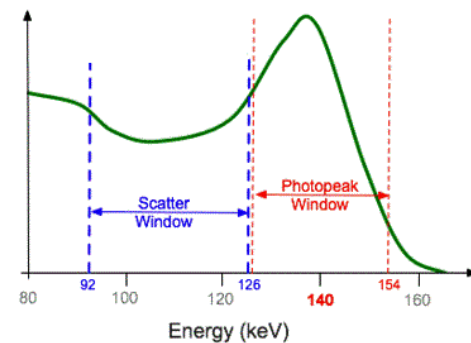
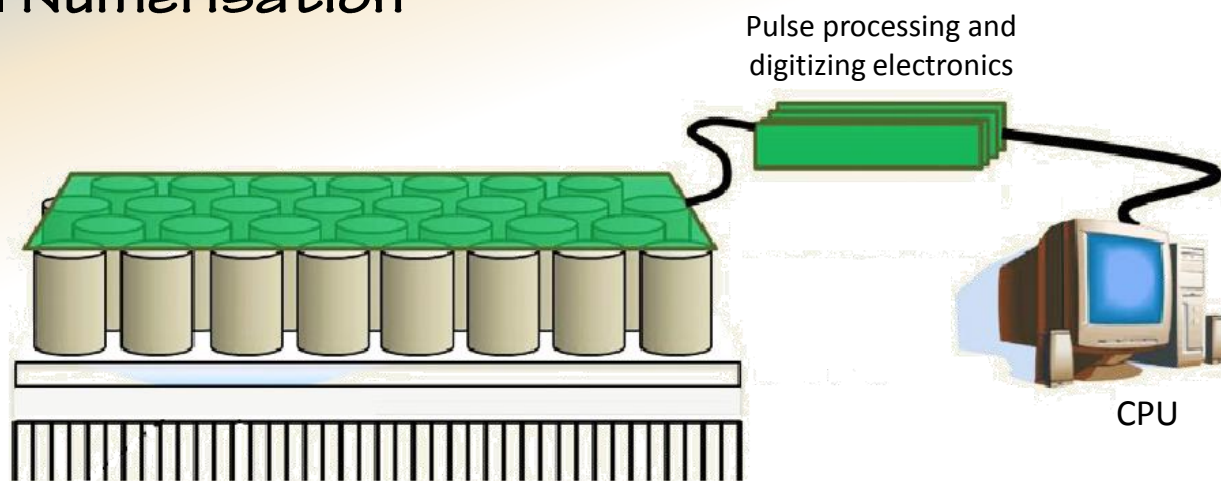
$$Y = Y^+ - Y^-$$

$$E = X^+ + X^- + Y^+ + Y^-$$

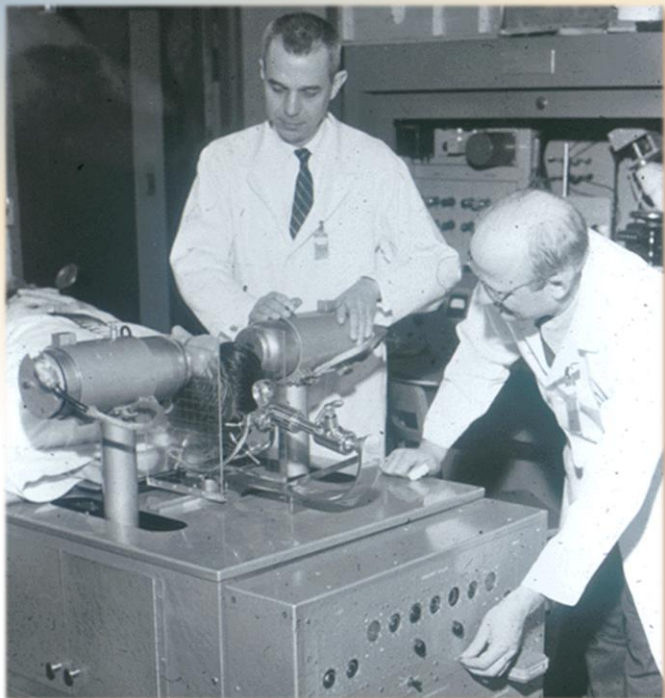


TEMP - acquisition

■ Numérisation



TEP - acquisition



G Brownell & C Burnham
Boston 1952

The New England Journal of Medicine

Copyright, 1951, by the Massachusetts Medical Society

Volume 245

DECEMBER 6, 1951

Number 23

THE USES OF NUCLEAR DISINTEGRATION IN THE DIAGNOSIS AND TREATMENT OF BRAIN TUMOR*

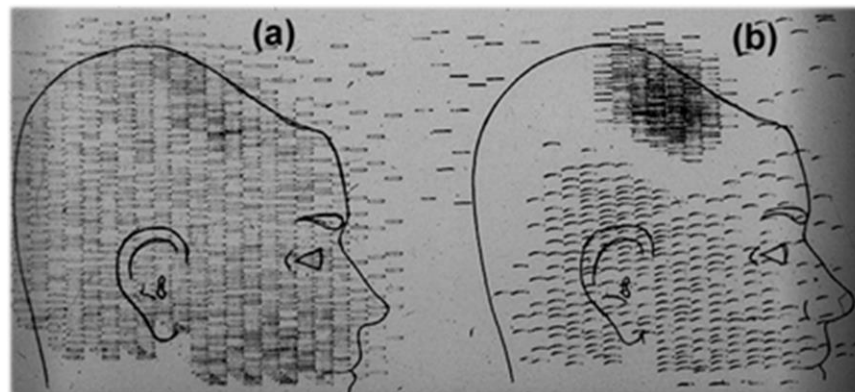
WILLIAM H. SWEET, M.D.†

BOSTON

IN THE utilization of isotopes to aid in the clinical management of intracranial tumors that has been developed at this hospital in the past few years, we have been aided by many workers. The time at which an elaborate addition to the hospital's investigative facilities is brought into action is perhaps a suitable time to draw attention to the diversity of talents that has been required for making some headway in such a limited field as that of brain tumors.

ascertain by the counting rate when neoplasm is entered and where its limits are.

Such a probe counter was first designed by the physicist Dr. Charles Robinson, then of Wisconsin. Dr. Solomon promptly secured the aid of Dr. Robinson and his counter, which device we used in man in time to send a telegram with the data to

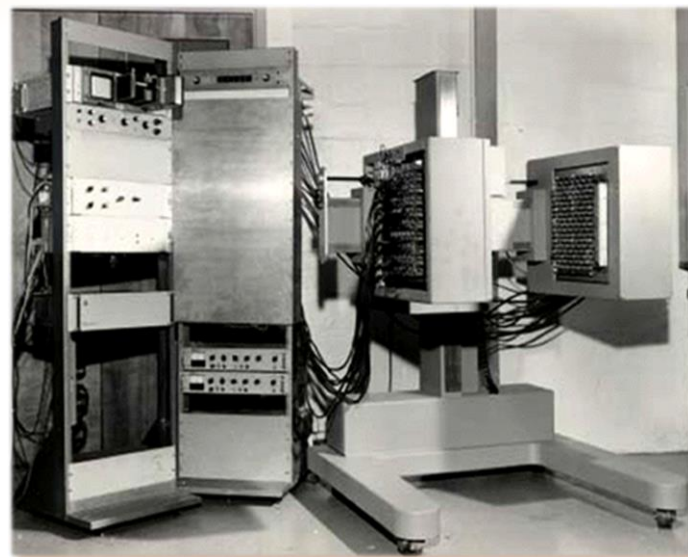


TEP - acquisition

Multidétecteur
Mode tomographique

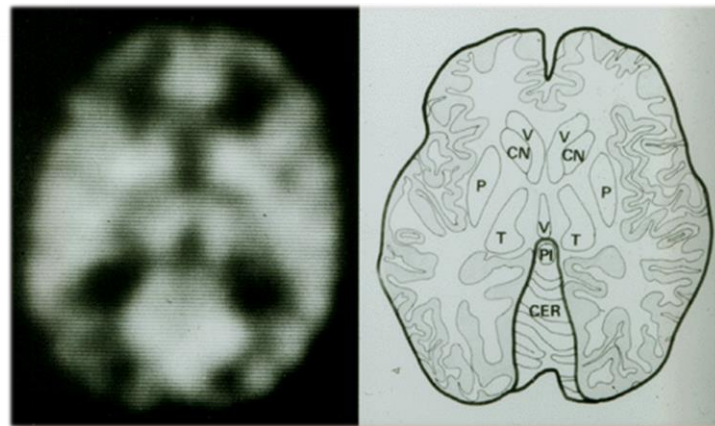


1975



1970

^{18}F FDG
1976



TEP - acquisition



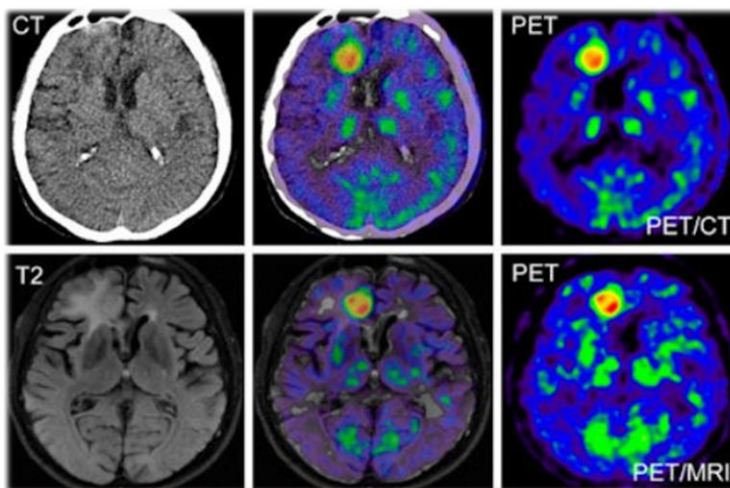
Anneau de détecteurs
1980



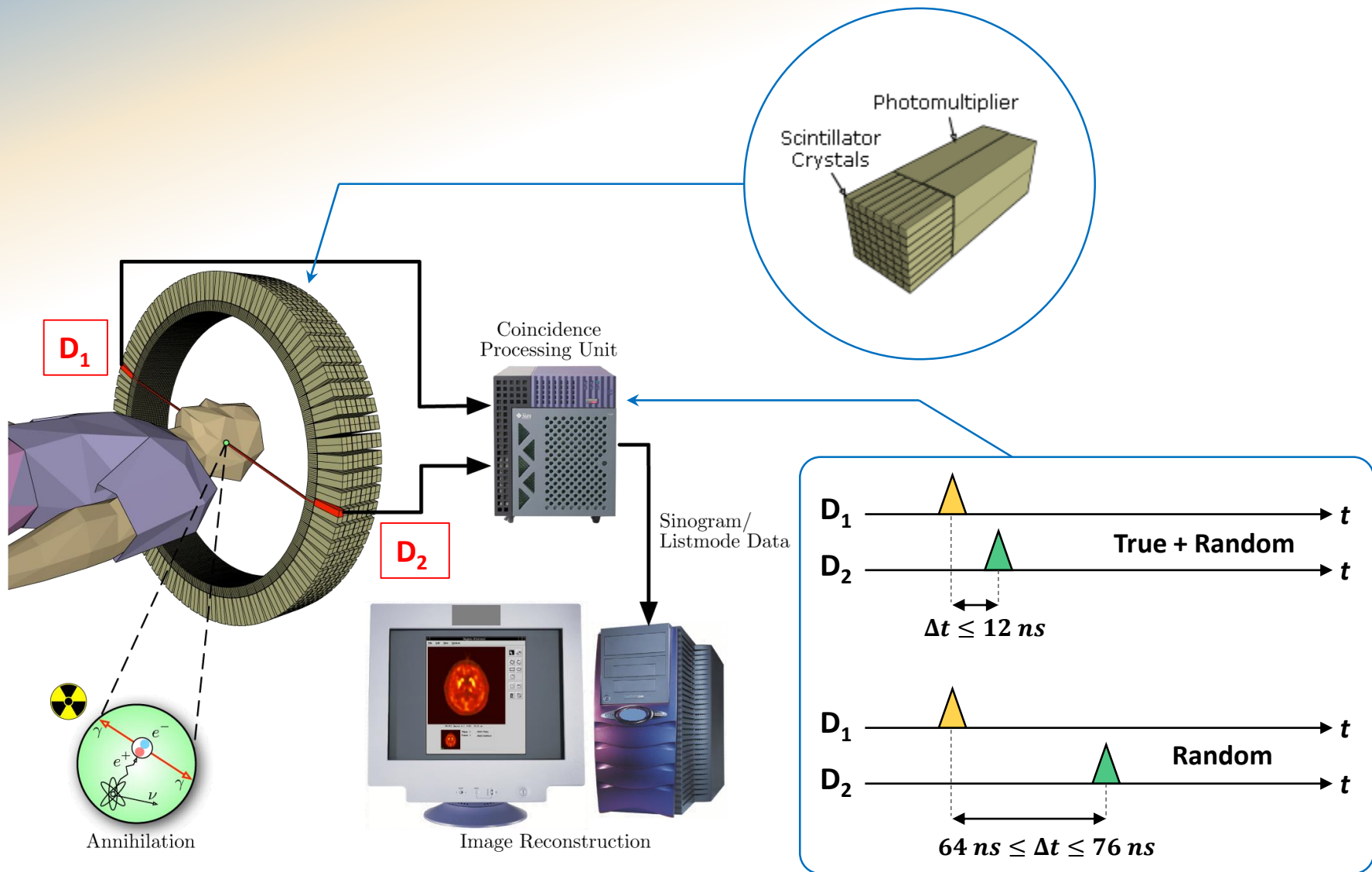
TEP-CT
1998



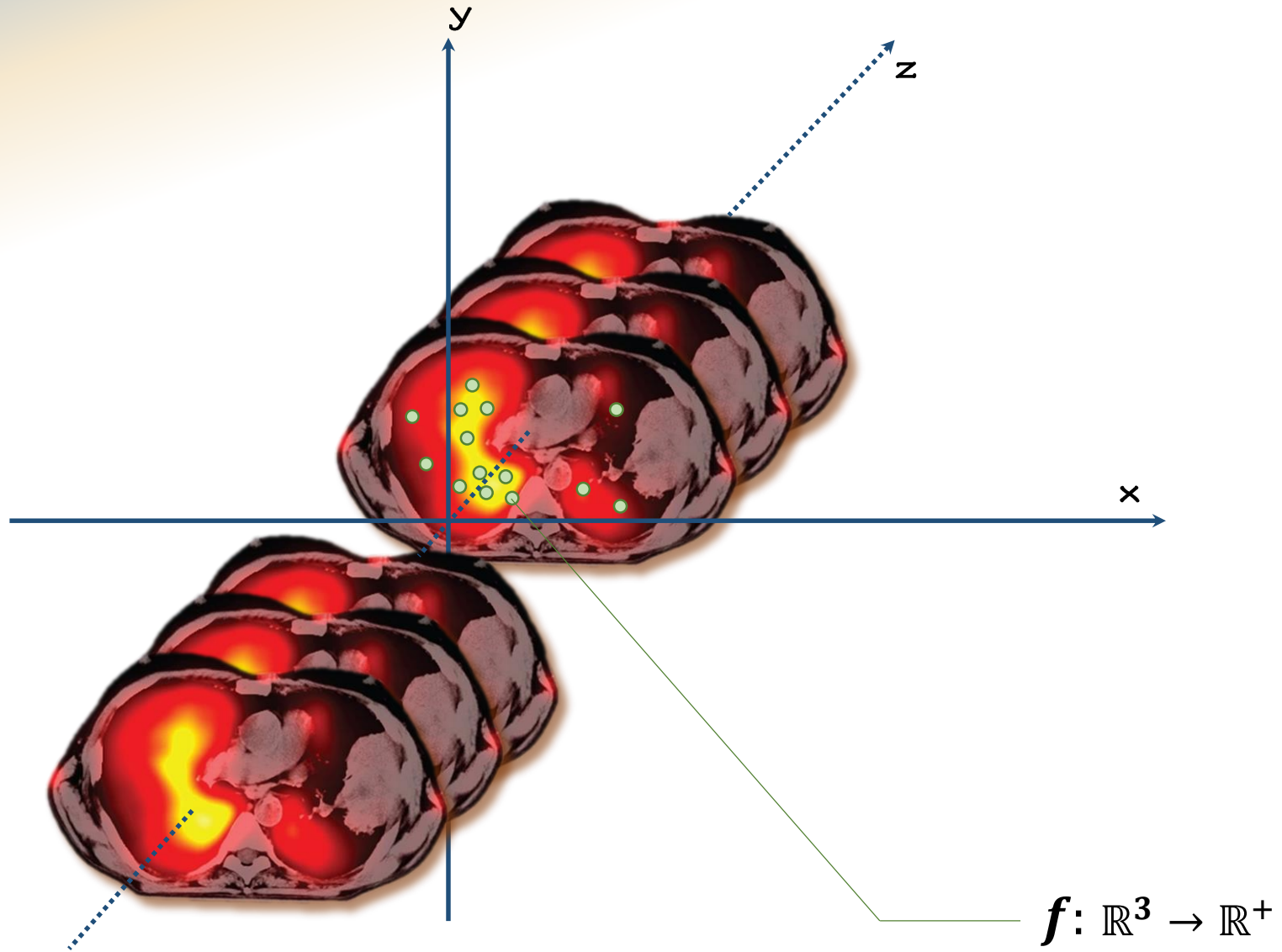
TEP-MR
2010



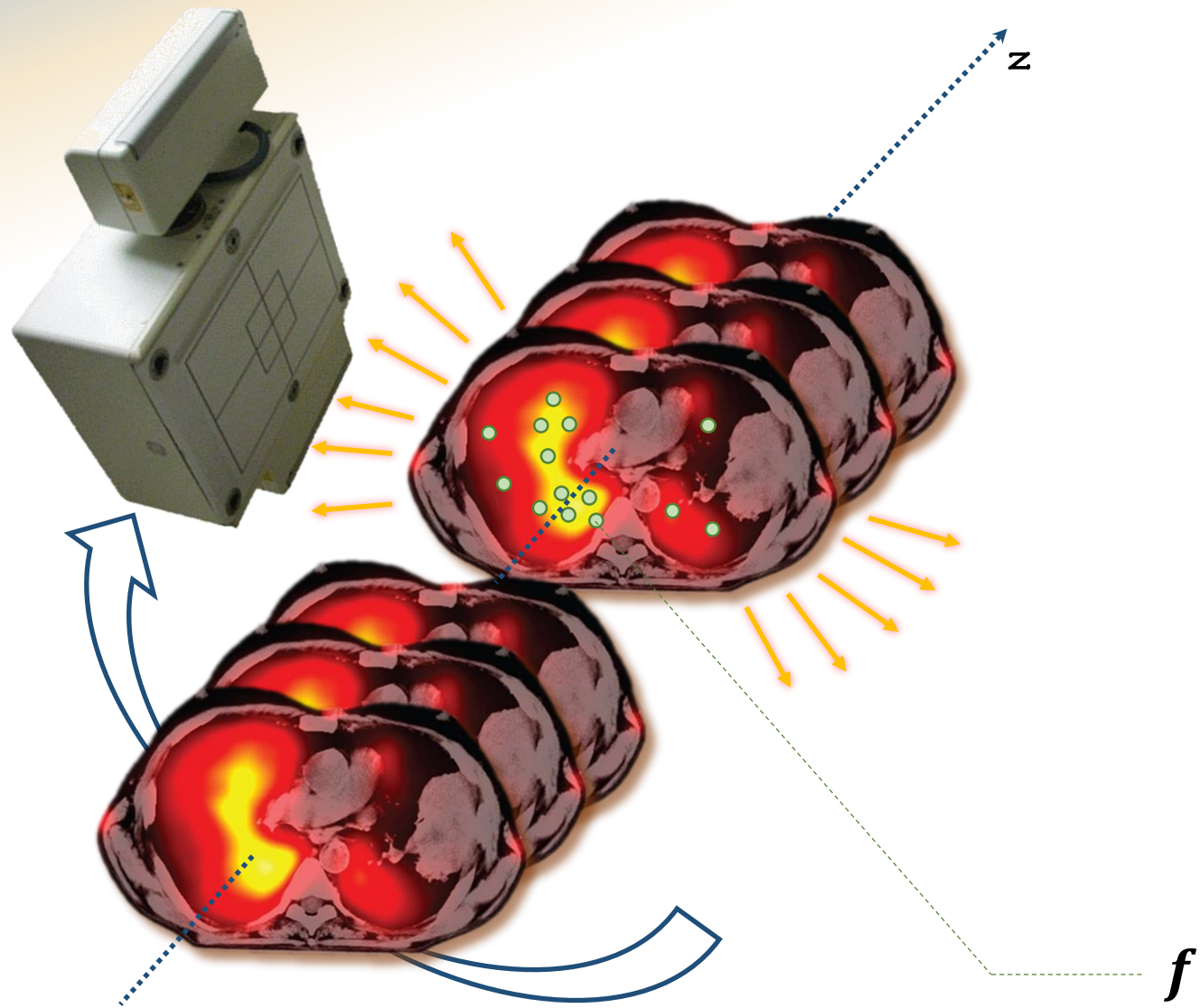
TEP - acquisition



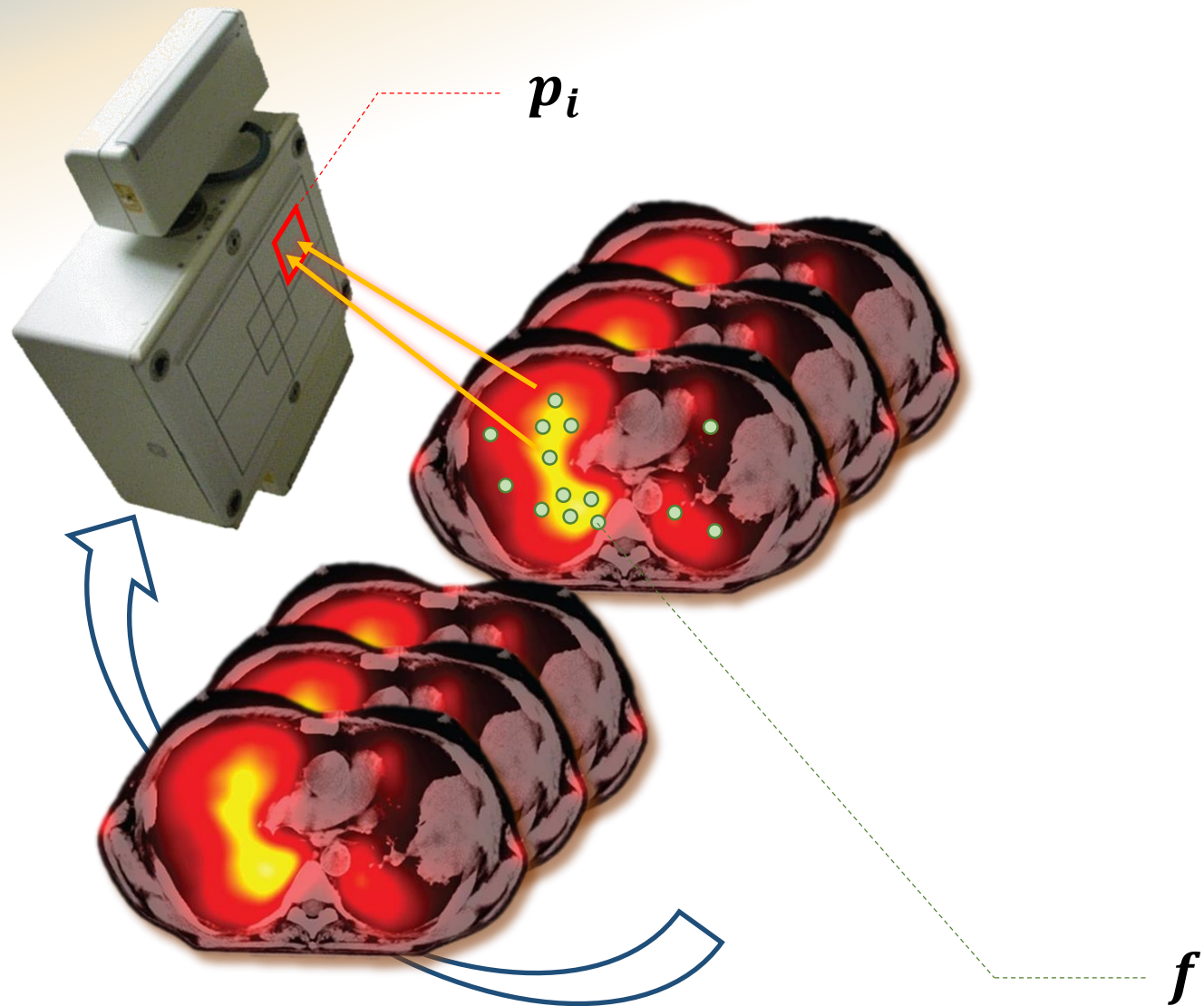
Problème tomographique



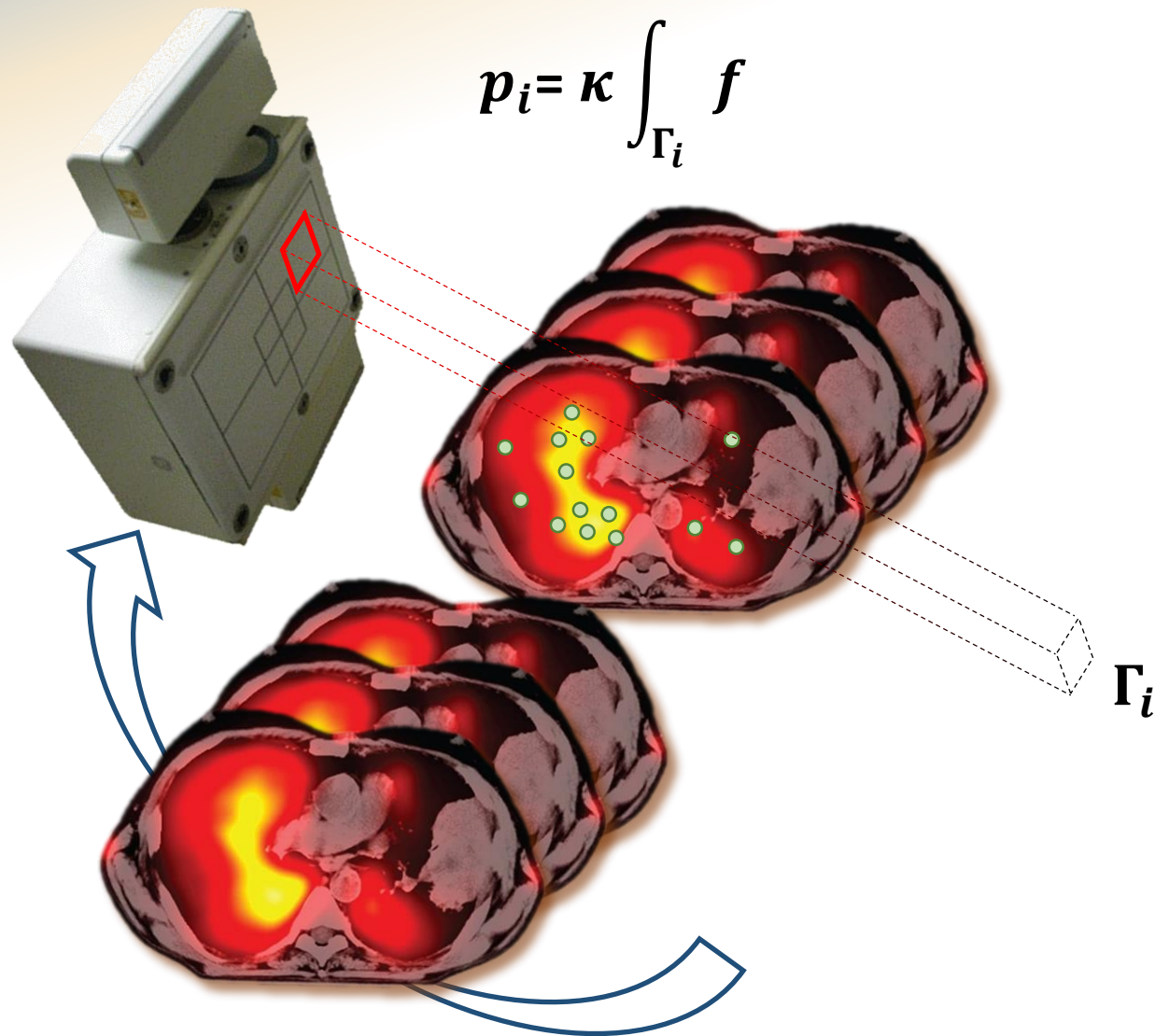
Problème tomographique



Problème tomographique

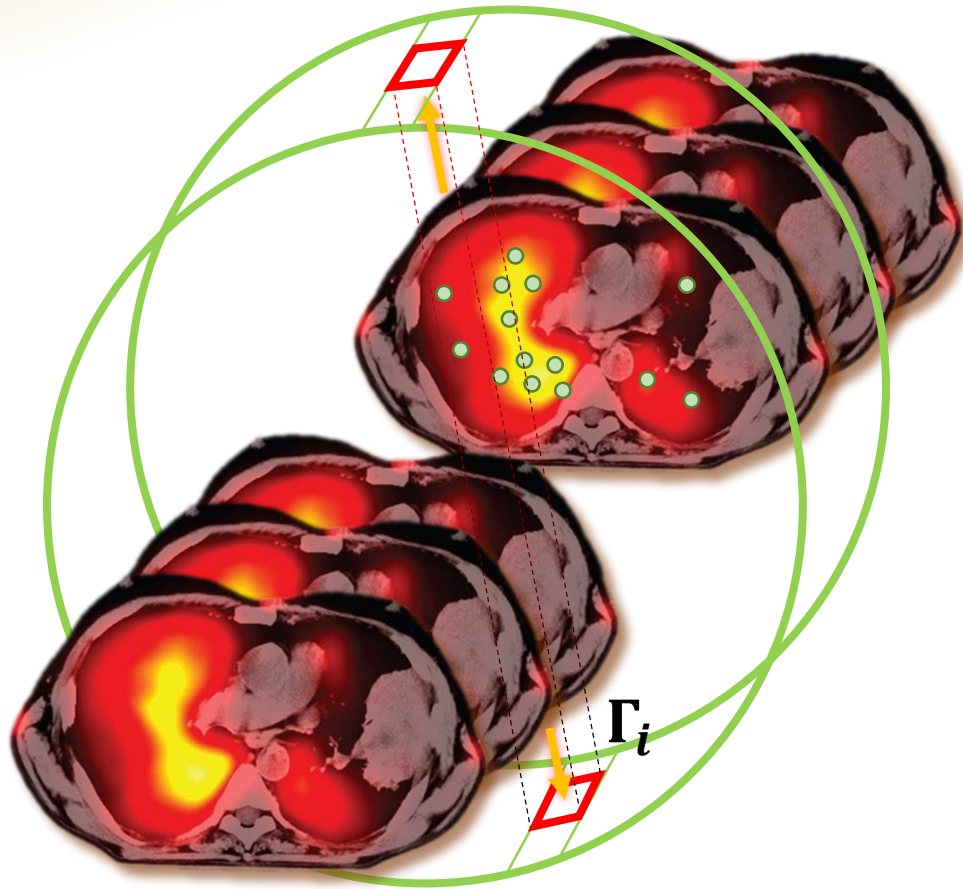


Problème tomographique

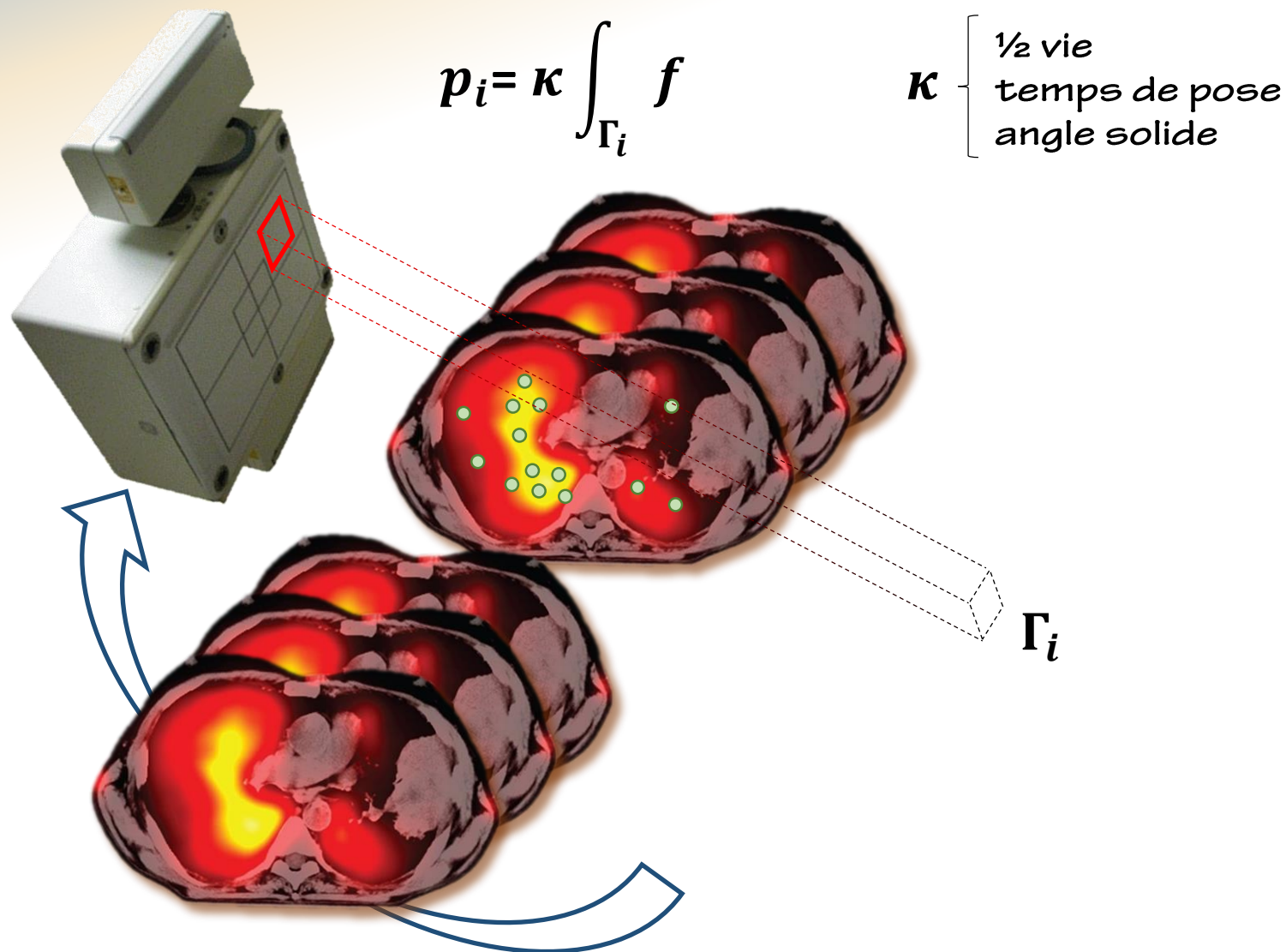


Problème tomographique

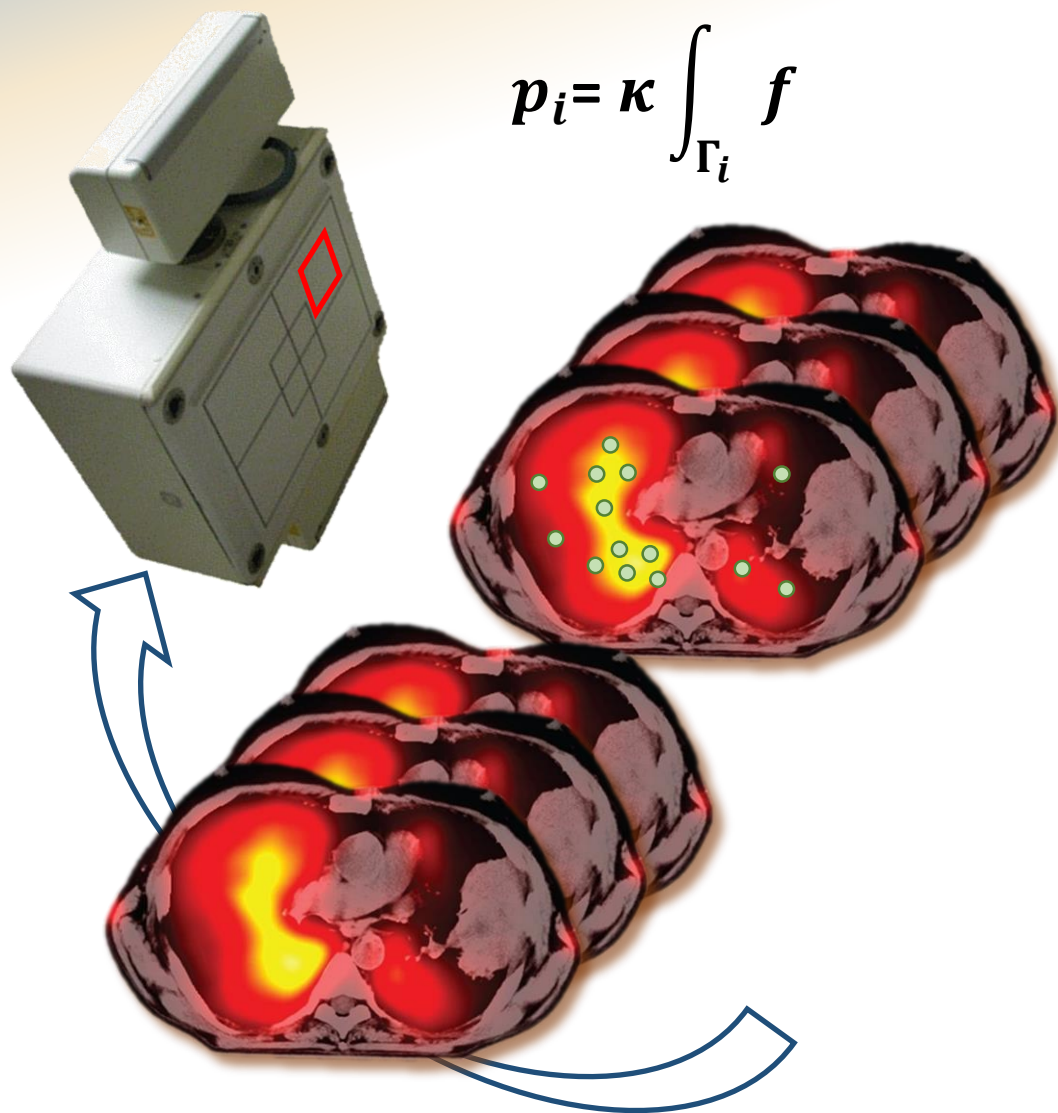
$$p_i = \kappa \int_{\Gamma_i} f$$



Problème tomographique



Problème tomographique

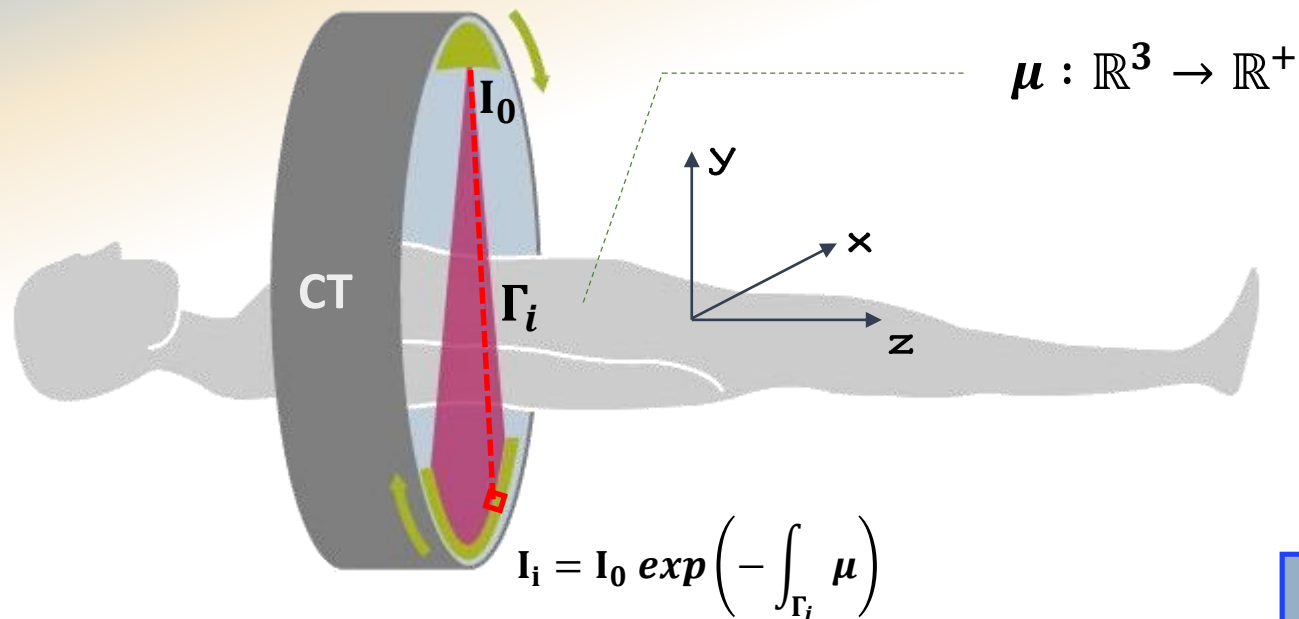


$p_i \quad (i = 1 \dots N)$



f

Problème tomographique



$$p_i = -\ln\left(\frac{I_i}{I_0}\right) = \int_{\Gamma_i} \mu$$

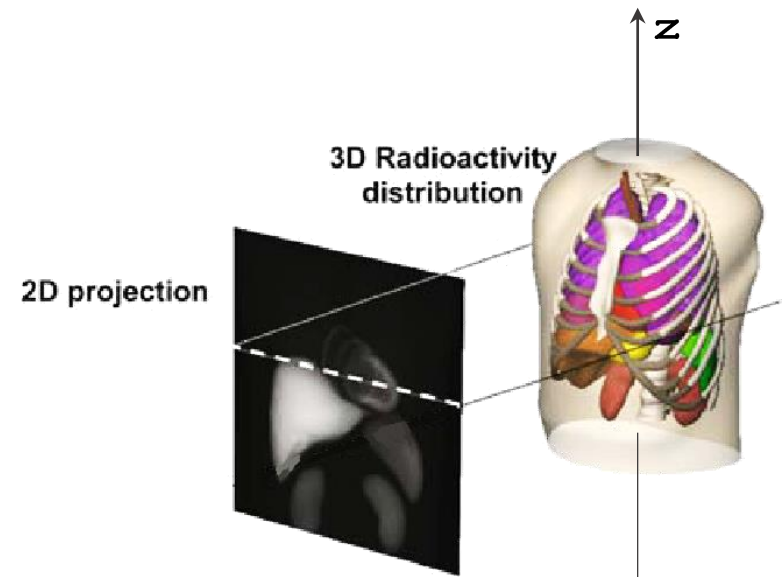
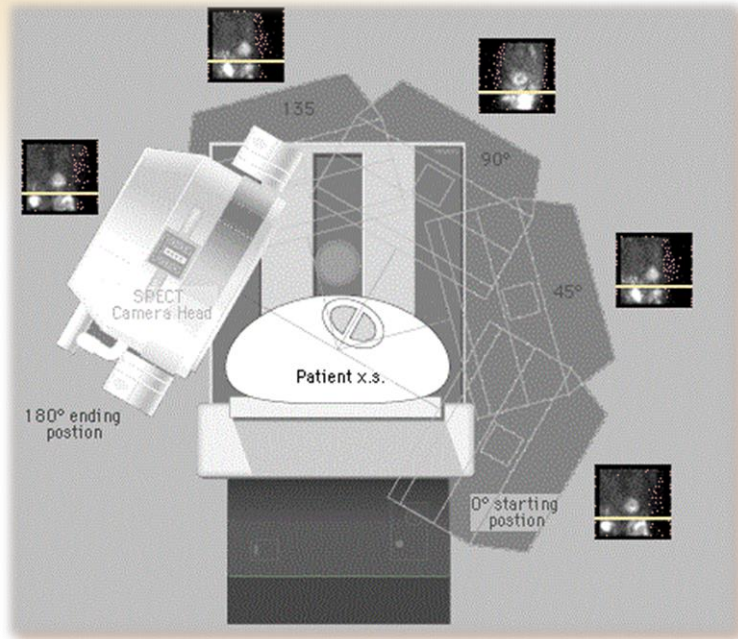
$p_i \quad (i = 1 \dots N)$



μ

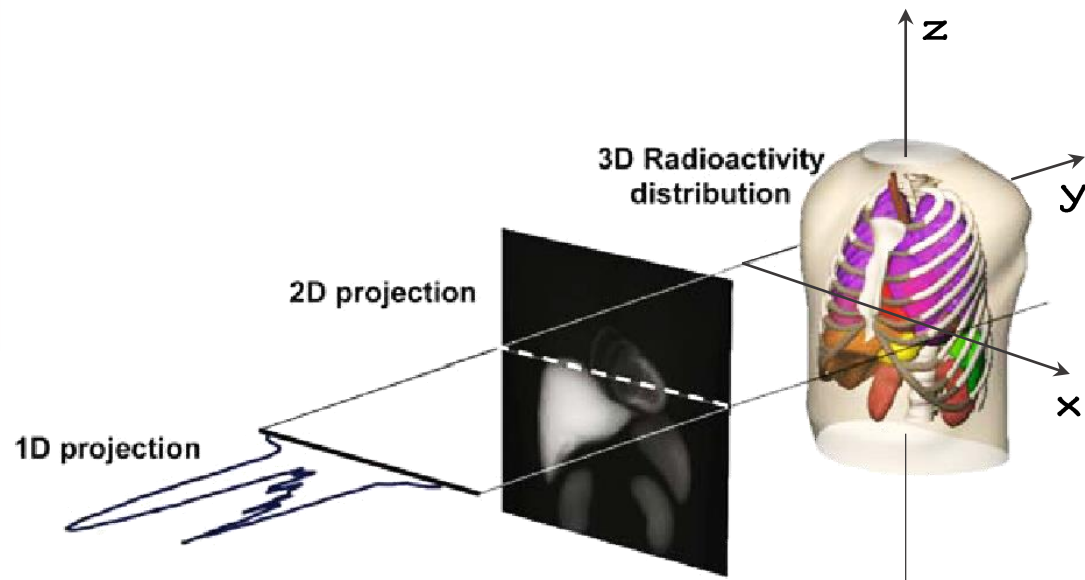
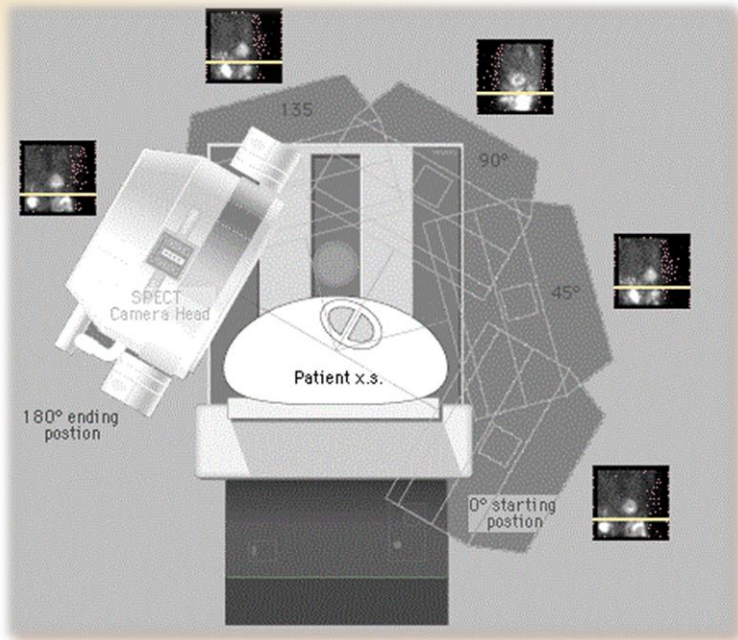
Reconstruction

■ Modèle analytique



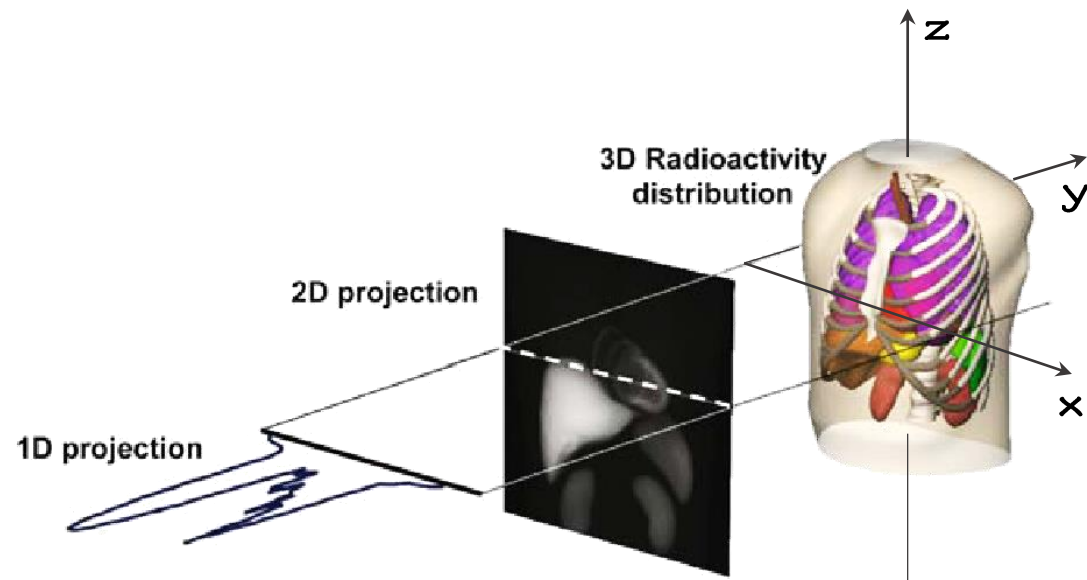
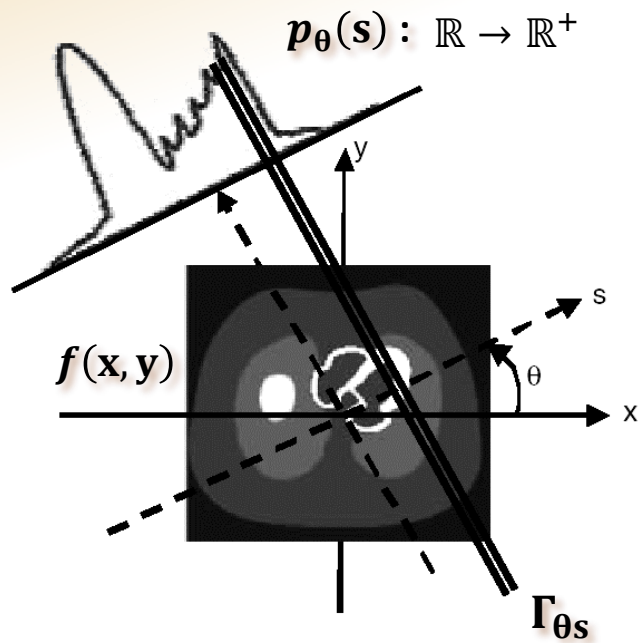
Reconstruction

■ Modèle analytique



Reconstruction

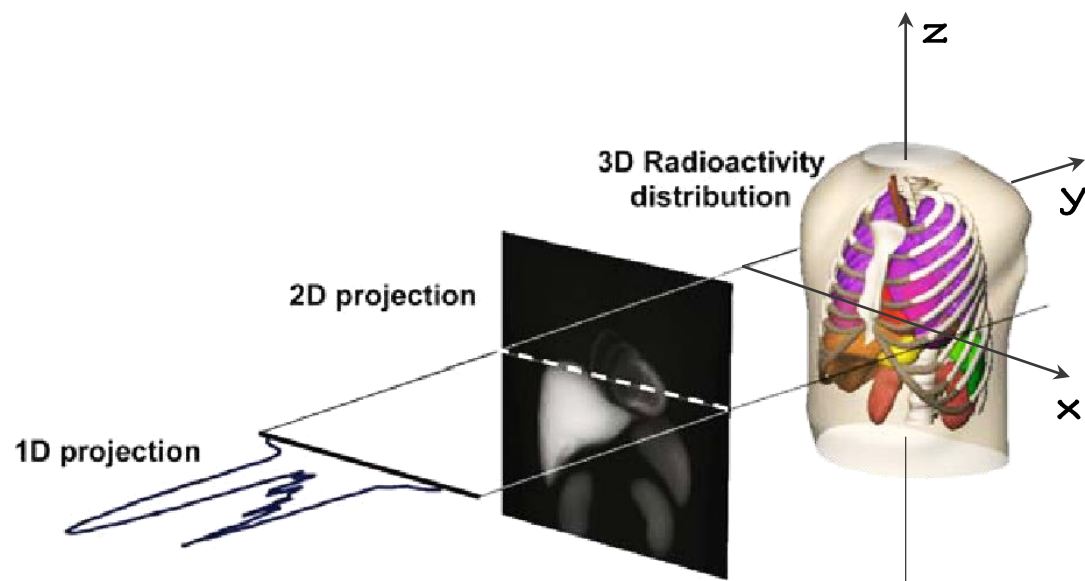
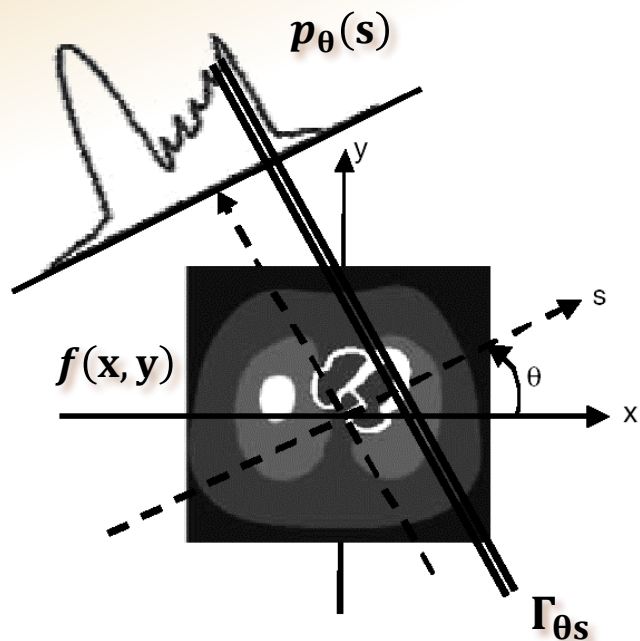
■ Modèle analytique



Reconstruction

■ Modèle analytique

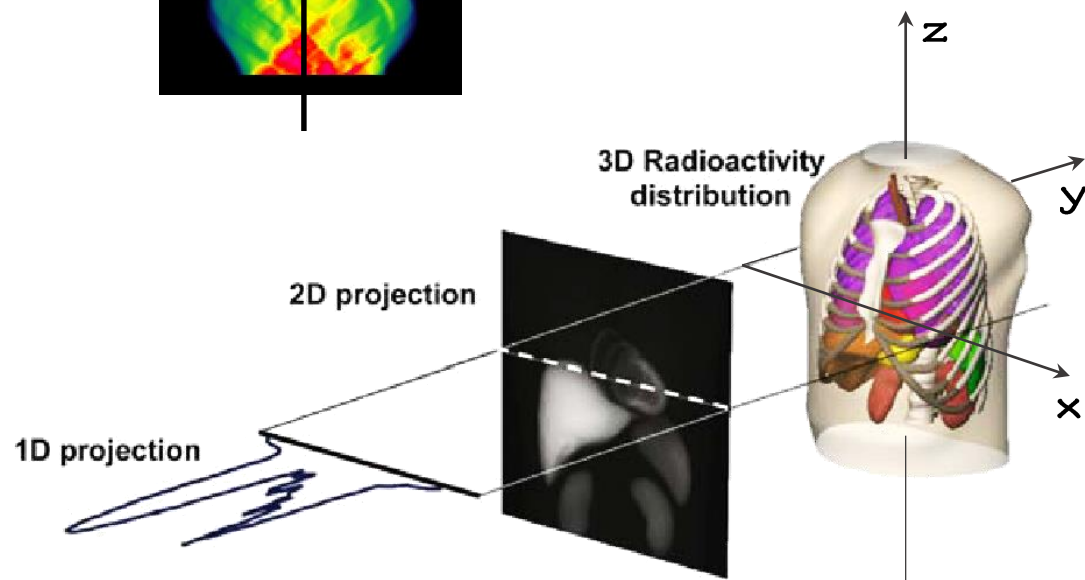
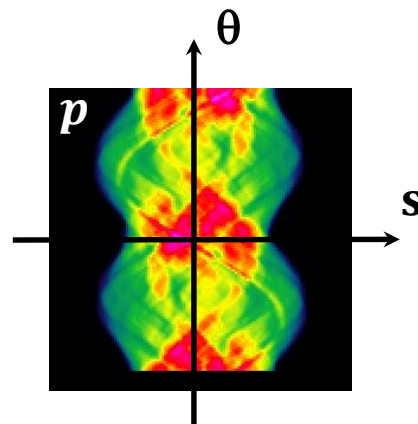
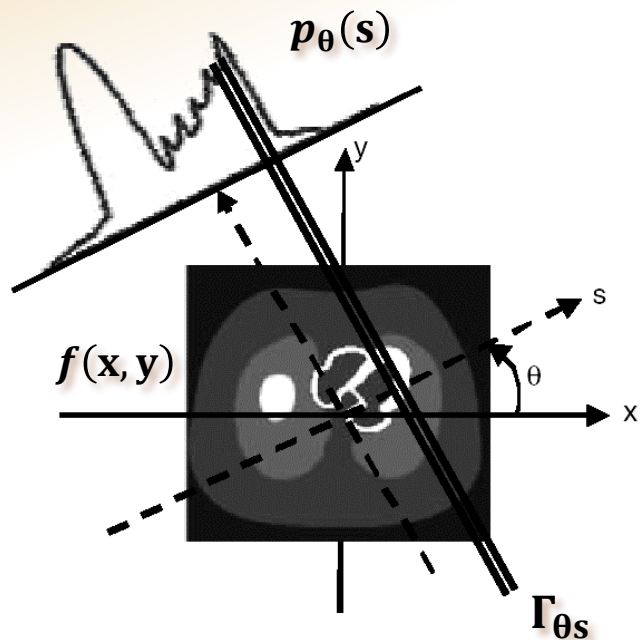
$$p_{\theta}(s) = \int_{\Gamma_{\theta s}} f(x, y)$$



Reconstruction

■ Modèle analytique

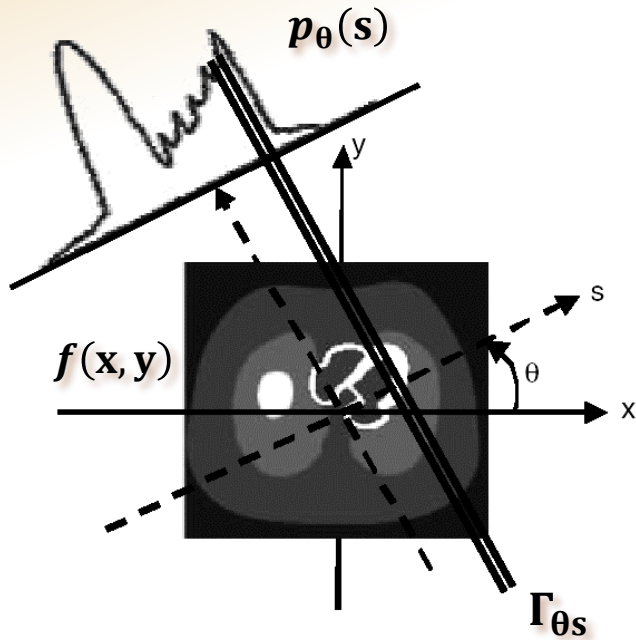
$$p_{\theta}(s) = \int_{\Gamma_{\theta s}} f(x, y)$$



Reconstruction

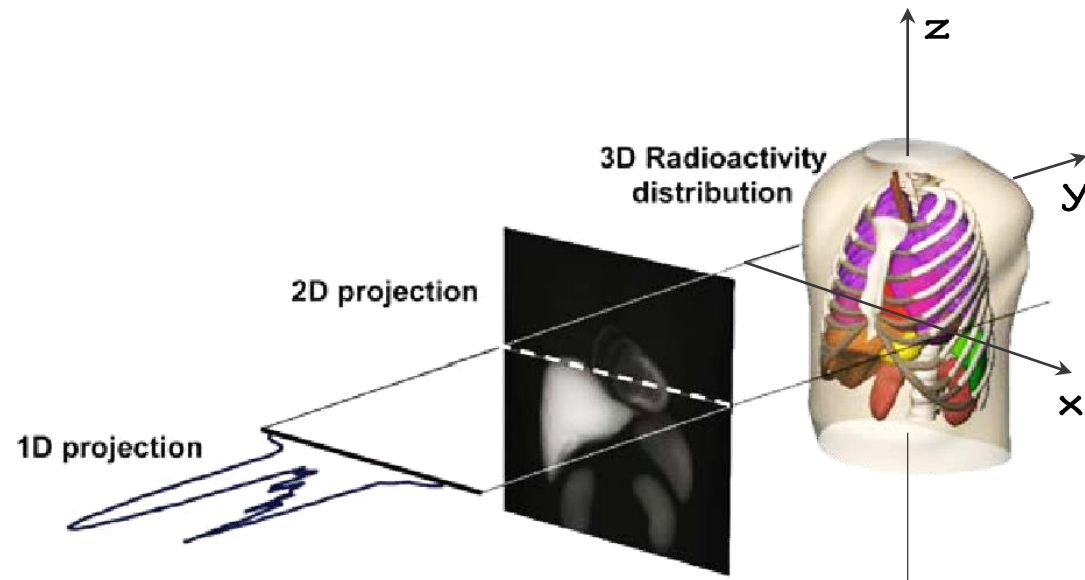
■ Modèle analytique

$$p_{\theta}(s) = \int_{\Gamma_{\theta s}} f(x, y)$$



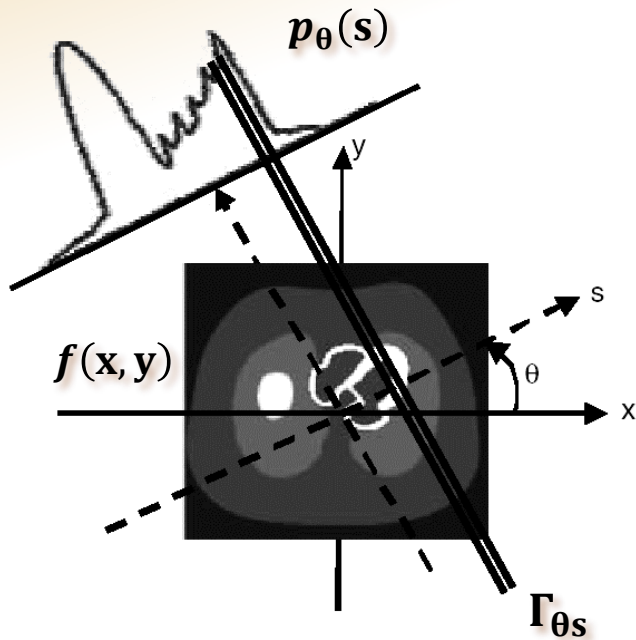
Projection (Radon): $p = R f$

Rétro-projection: R^*



Reconstruction

■ Modèle analytique



Projection

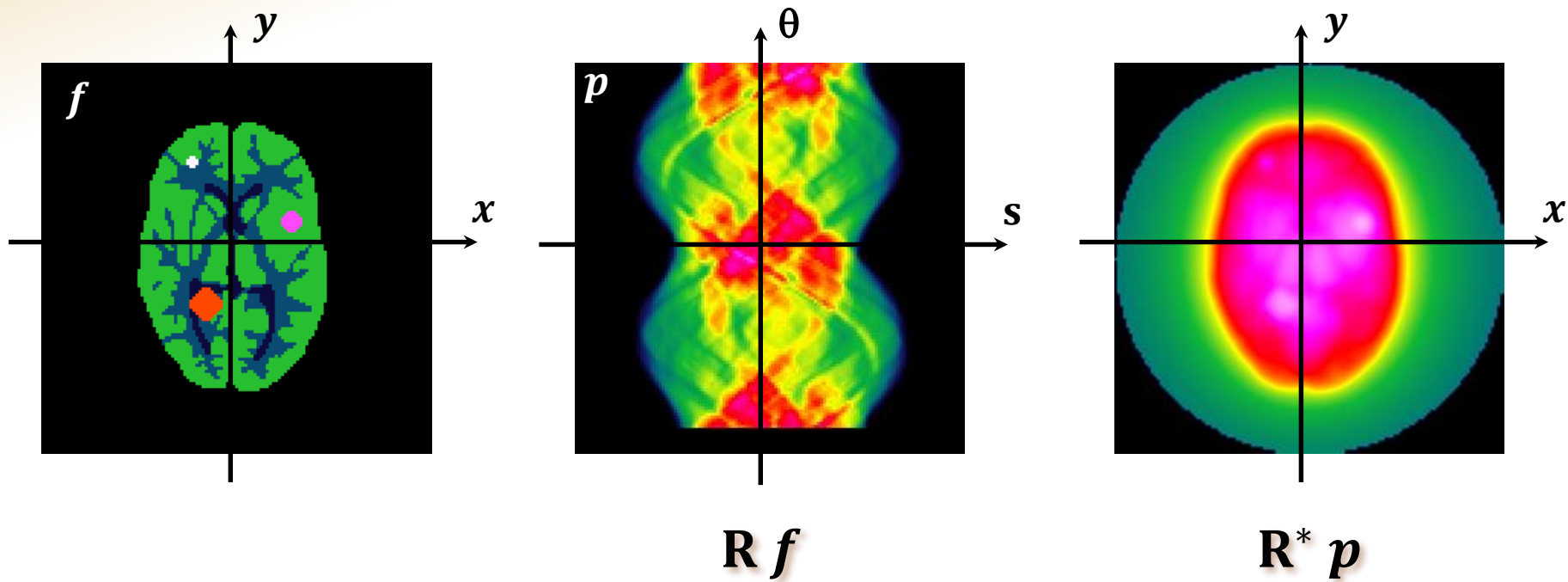
$$Rf_{\theta}(s) = \int_{\mathbb{R}} f(s \cos \theta - t \sin \theta, s \sin \theta + t \cos \theta) dt$$

Rétro-projection

$$R^*p(x, y) = \int_0^{\pi} p_{\theta}(x \cos \theta + y \sin \theta) d\theta$$

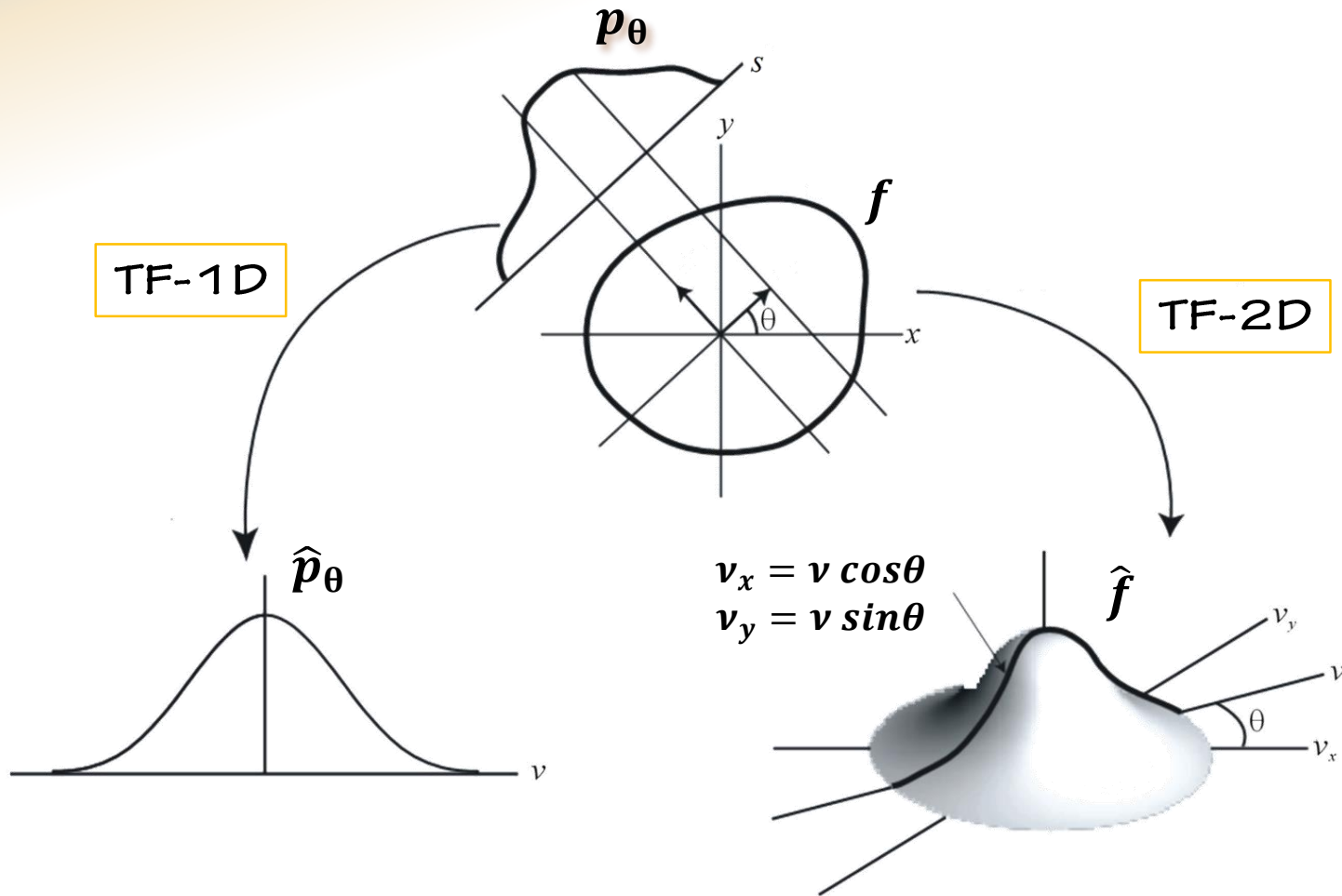
Reconstruction

■ Modèle analytique



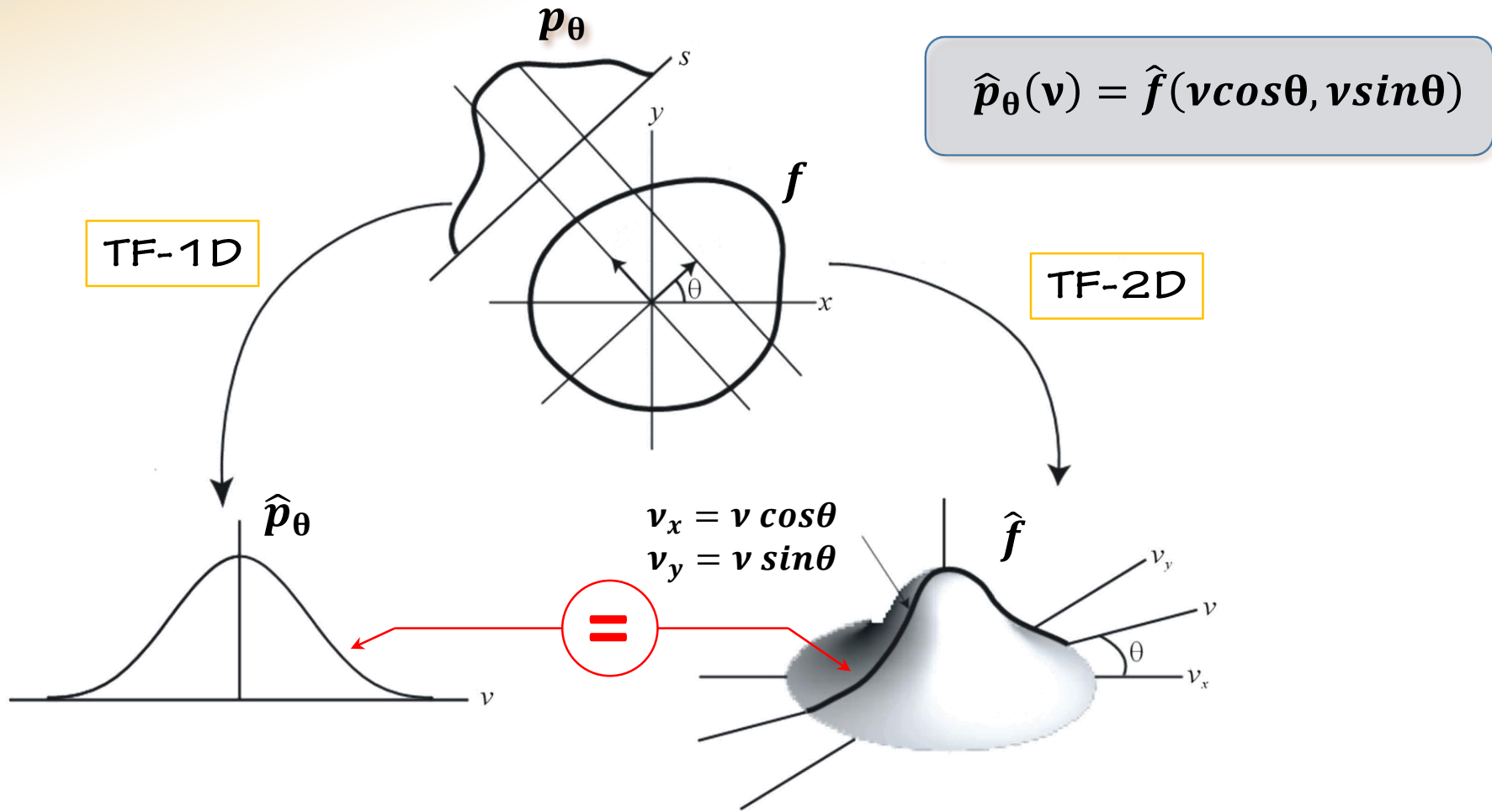
Reconstruction

■ Modèle analytique — Synthèse de Fourier



Reconstruction

■ Modèle analytique — Synthèse de Fourier



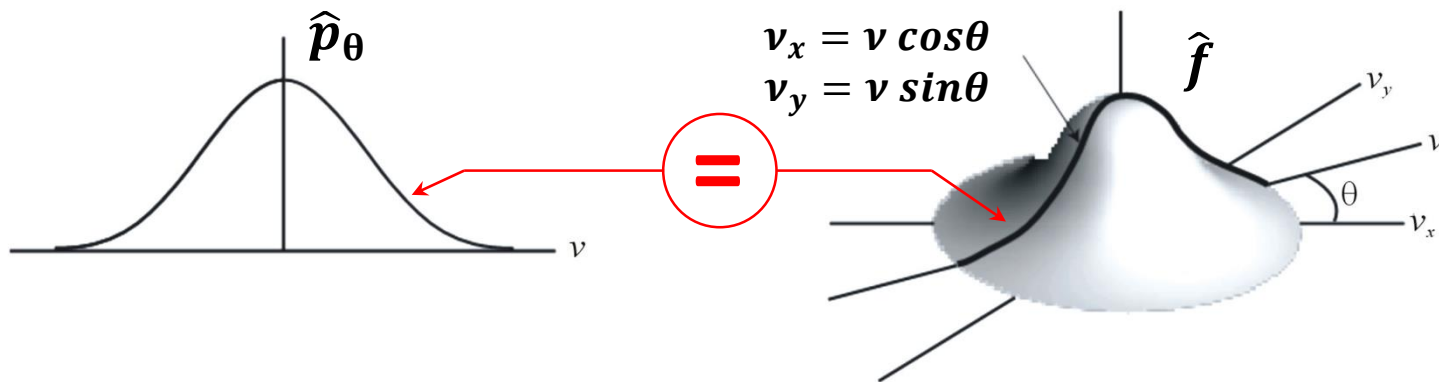
Reconstruction

■ Modèle analytique — Synthèse de Fourier

$$\hat{p}_\theta(v) = \int_{\mathbb{R}} p_\theta(s) e^{-ivs} ds = \iint_{\mathbb{R}^2} f(s \cos \theta - t \sin \theta, s \sin \theta + t \cos \theta) e^{-ivs} ds dt$$

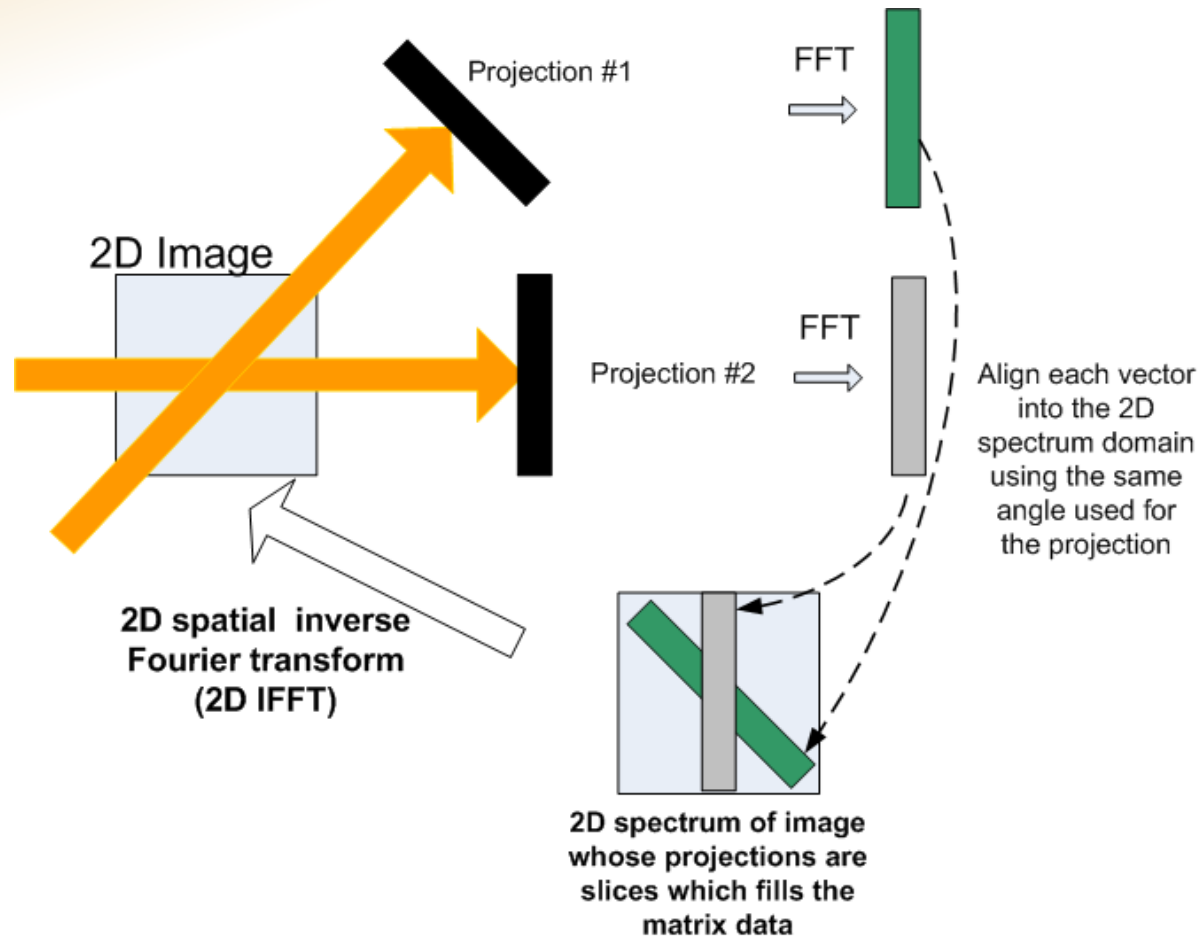
$$\begin{cases} x = s \cos \theta - t \sin \theta \\ y = s \sin \theta + t \cos \theta \end{cases} \quad \begin{cases} s = x \cos \theta + y \sin \theta \\ t = x \sin \theta - y \cos \theta \end{cases}$$

$$= \iint_{\mathbb{R}^2} f(x, y) e^{-iv(x \cos \theta + y \sin \theta)} dx dy = \hat{f}(v \cos \theta, v \sin \theta)$$



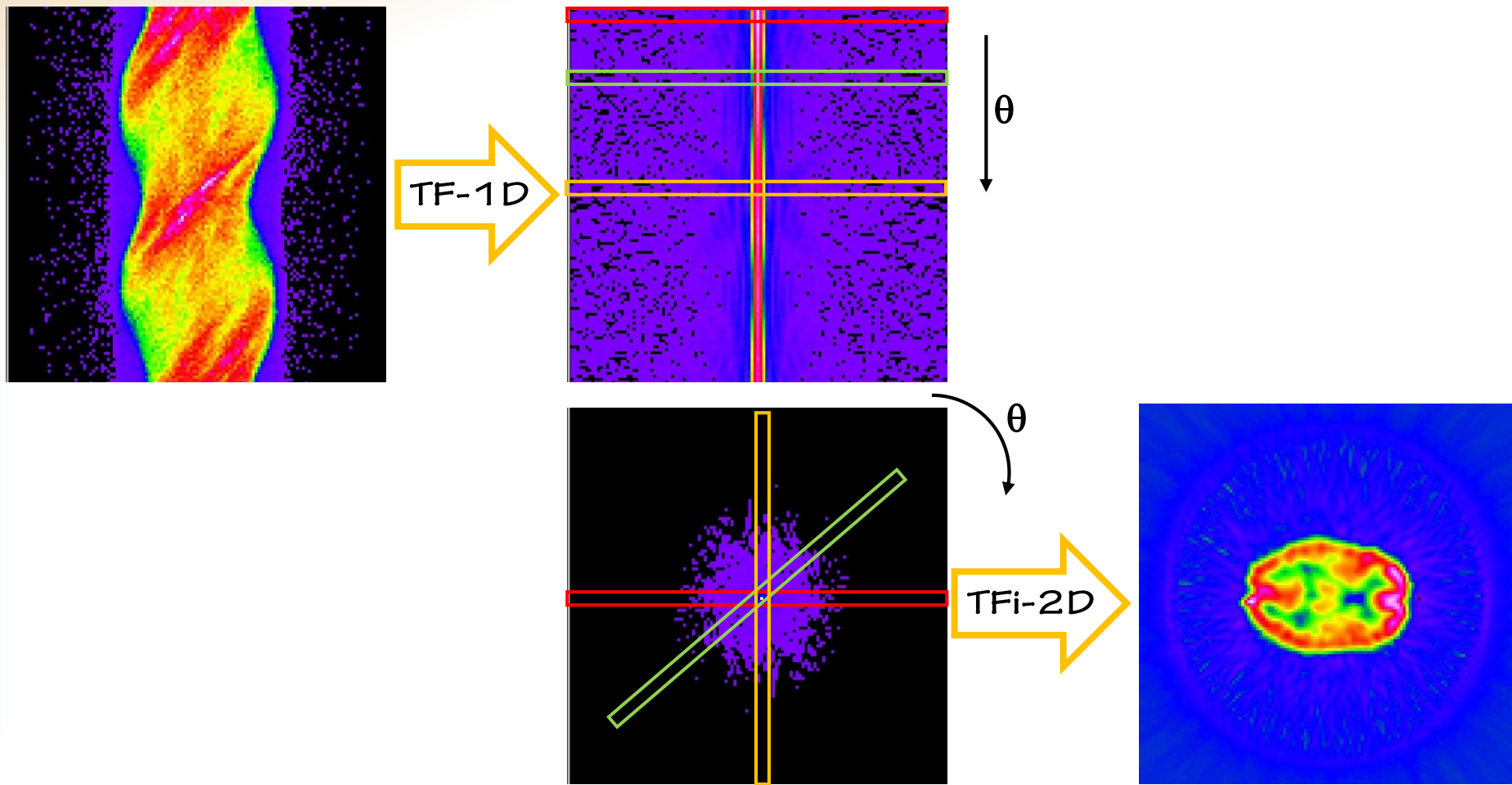
Reconstruction

■ Modèle analytique — Synthèse de Fourier



Reconstruction

■ Modèle analytique — Synthèse de Fourier



Reconstruction

- Modèle analytique — Rétro-projection filtrée

$$f = \iint_{\mathbb{R}^2} \hat{f} e^{i(xv_x + yv_y)} dv_x dv_y$$

$$\begin{aligned} v_x &= v \cos\theta \\ v_y &= v \sin\theta \end{aligned}$$

Reconstruction

- Modèle analytique — Rétro-projection filtrée

$$f = \iint_{\mathbb{R}^2} \hat{f} e^{i(xv_x + yv_y)} dv_x dv_y$$

$$\begin{aligned} v_x &= v \cos\theta \\ v_y &= v \sin\theta \end{aligned}$$

$$f = \int d\theta \int \hat{f} e^{iv(x\cos\theta + y\sin\theta)} |v| dv$$

Reconstruction

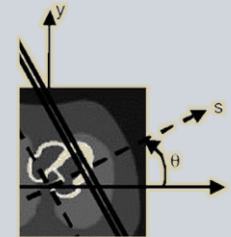
■ Modèle analytique — Rétro-projection filtrée

$$f = \iint_{\mathbb{R}^2} \hat{f} e^{i(xv_x + yv_y)} dv_x dv_y$$

$$\begin{aligned} v_x &= v \cos\theta \\ v_y &= v \sin\theta \end{aligned}$$

$$f = \int d\theta \int \hat{f} e^{iv(x\cos\theta + y\sin\theta)} |v| dv$$

$= s$



Reconstruction

■ Modèle analytique — Rétro-projection filtrée

$$f = \iint_{\mathbb{R}^2} \hat{f} e^{i(xv_x + yv_y)} dv_x dv_y$$

$$\begin{aligned} v_x &= v \cos\theta \\ v_y &= v \sin\theta \end{aligned}$$

$$f = \int d\theta \int \hat{f} e^{iv(x\cos\theta + y\sin\theta)} |v| dv$$

$$\hat{f}(v\cos\theta, v\sin\theta) = \hat{p}_\theta(v)$$

Reconstruction

■ Modèle analytique — Rétro-projection filtrée

$$f = \iint_{\mathbb{R}^2} \hat{f} e^{i(xv_x + yv_y)} dv_x dv_y$$

$$\begin{aligned} v_x &= v \cos\theta \\ v_y &= v \sin\theta \end{aligned}$$

$$f = \int d\theta \int \hat{f} e^{iv(x\cos\theta + y\sin\theta)} |v| dv$$

$$\hat{f}(v\cos\theta, v\sin\theta) = \hat{p}_\theta(v)$$

$$f = \int d\theta \int dv e^{isv} |v| \hat{p}$$

Reconstruction

■ Modèle analytique — Rétro-projection filtrée

$$f = \iint_{\mathbb{R}^2} \hat{f} e^{i(xv_x + yv_y)} dv_x dv_y$$

$$\begin{aligned} v_x &= v \cos\theta \\ v_y &= v \sin\theta \end{aligned}$$

$$f = \int d\theta \int \hat{f} e^{iv(x\cos\theta + y\sin\theta)} |v| dv$$

$$\hat{f}(v\cos\theta, v\sin\theta) = \hat{p}_\theta(v)$$

$$f = \int d\theta \int dv e^{isv} |v| \hat{p}$$

$$\hat{H}p = |v| \hat{p}$$

Reconstruction

■ Modèle analytique — Rétro-projection filtrée

$$f = \iint_{\mathbb{R}^2} \hat{f} e^{i(xv_x + yv_y)} dv_x dv_y$$

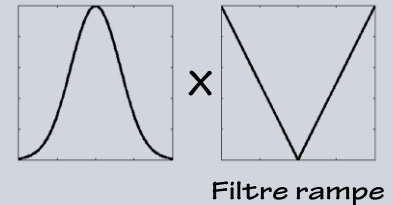
$$\begin{aligned} v_x &= v \cos\theta \\ v_y &= v \sin\theta \end{aligned}$$

$$f = \int d\theta \int \hat{f} e^{iv(x\cos\theta + y\sin\theta)} |v| dv$$

$$\hat{f}(v\cos\theta, v\sin\theta) = \hat{p}_\theta(v)$$

$$f = \int d\theta \int dv e^{isv} |v| \hat{p}$$

$$\widehat{H}p = |v| \hat{p}$$



Reconstruction

■ Modèle analytique — Rétro-projection filtrée

$$f = \iint_{\mathbb{R}^2} \hat{f} e^{i(xv_x + yv_y)} dv_x dv_y$$

$$\begin{aligned} v_x &= v \cos\theta \\ v_y &= v \sin\theta \end{aligned}$$

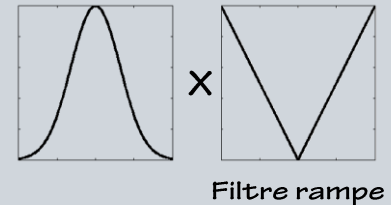
$$f = \int d\theta \int \hat{f} e^{iv(x\cos\theta + y\sin\theta)} |v| dv$$

$$\hat{f}(v\cos\theta, v\sin\theta) = \hat{p}_\theta(v)$$

$$f = \int d\theta \int dv e^{isv} |v| \hat{p}$$

$$\hat{H}p = |v| \hat{p}$$

TF inv - 1D



Reconstruction

■ Modèle analytique — Rétro-projection filtrée

$$f = \iint_{\mathbb{R}^2} \hat{f} e^{i(xv_x + yv_y)} dv_x dv_y$$

$$\begin{aligned} v_x &= v \cos\theta \\ v_y &= v \sin\theta \end{aligned}$$

$$f = \int d\theta \int \hat{f} e^{iv(x\cos\theta + y\sin\theta)} |v| dv$$

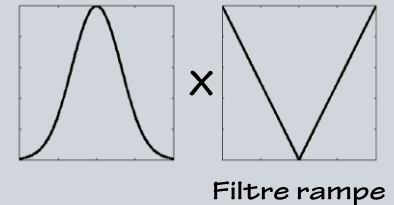
$$\hat{f}(v\cos\theta, v\sin\theta) = \hat{p}_\theta(v)$$

$$f = \int d\theta \int dv e^{isv} |v| \hat{p}$$

$$\hat{H}p = |v| \hat{p}$$

TF inv - 1D

Rétro-projection



Reconstruction

■ Modèle analytique — Rétro-projection filtrée

$$f = \iint_{\mathbb{R}^2} \hat{f} e^{i(xv_x + yv_y)} dv_x dv_y$$

$$\begin{aligned} v_x &= v \cos\theta \\ v_y &= v \sin\theta \end{aligned}$$

$$f = \int d\theta \int \hat{f} e^{iv(x\cos\theta + y\sin\theta)} |v| dv$$

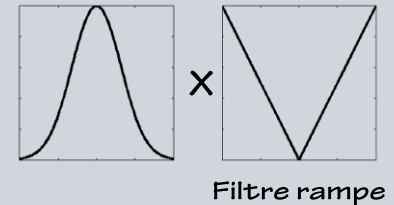
$$\hat{f}(v\cos\theta, v\sin\theta) = \hat{p}_\theta(v)$$

$$f = \int d\theta \int dv e^{isv} |v| \hat{p}$$

$$\hat{H}p = |v| \hat{p}$$

TF inv -1D

Rétro-projection



$$f = R^* Hp$$

Reconstruction

- *Modèle analytique — Rétro-projection filtrée*

$$\mathbf{H}p = \text{TF}^{-1}(\hat{h}\hat{p})$$

$$\hat{h}(v) = |v|$$

$$\mathbf{H}p = h * p$$

Reconstruction

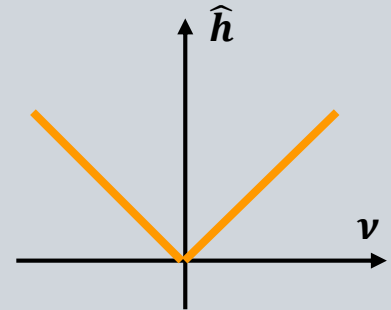
■ Modèle analytique — Rétro-projection filtrée

$$\mathbf{H}p = \text{TF}^{-1}(\hat{h}\hat{p})$$

$$\hat{h}(v) = |v|$$

$$\mathbf{H}p = h * p$$

$$h(s) = \int_{\mathbb{R}} \hat{h}(v) e^{ivs} dv$$



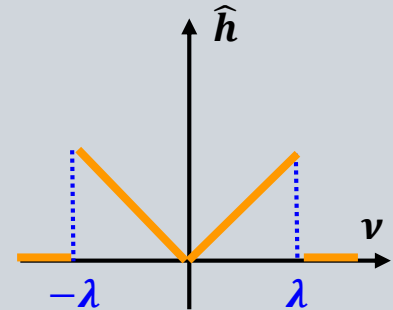
Reconstruction

- Modèle analytique — Rétro-projection filtrée

$$\mathbf{H}p = \text{TF}^{-1}(\widehat{h} \widehat{p})$$

$$\mathbf{H}p = h * p$$

$$h(s) = \int_{-\lambda}^{\lambda} \widehat{h}(v) e^{ivs} dv$$



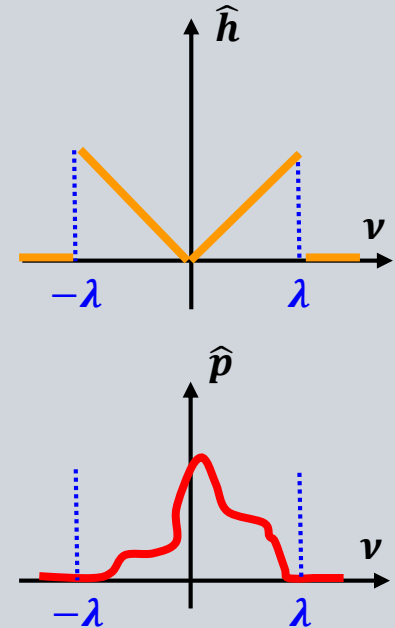
Reconstruction

■ Modèle analytique — Rétro-projection filtrée

$$\mathbf{H}p = \text{TF}^{-1}(\widehat{h} \widehat{p})$$

$$\mathbf{H}p = h * p$$

$$h(s) = \int_{-\lambda}^{\lambda} \widehat{h}(v) e^{ivs} dv$$



$$\begin{aligned} \lambda &\geq v_{max} \\ &= 2\pi F_{max} = \pi \end{aligned}$$

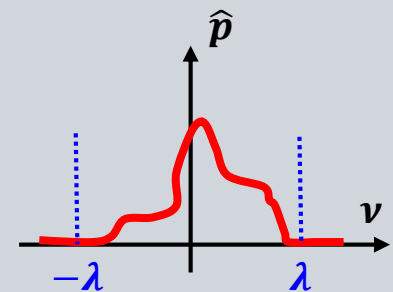
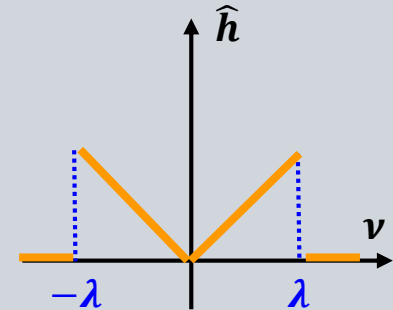
Reconstruction

- Modèle analytique — Rétro-projection filtrée

$$\mathbf{H}p = \text{TF}^{-1}(\hat{h}\hat{p})$$

$$\mathbf{H}p = h * p$$

$$\begin{aligned}h(s) &= \int_{-\lambda}^{\lambda} \hat{h}(v) e^{ivs} dv \\&= -\int_{-\pi}^0 v e^{ivs} dv + \int_0^{\pi} v e^{ivs} dv \\&= \frac{2\pi}{s} \sin(\pi s) + \frac{2}{s^2} (\cos(\pi s) - 1)\end{aligned}$$



$$\begin{aligned}\lambda &\geq v_{max} \\&= 2\pi F_{max} = \pi\end{aligned}$$

Reconstruction

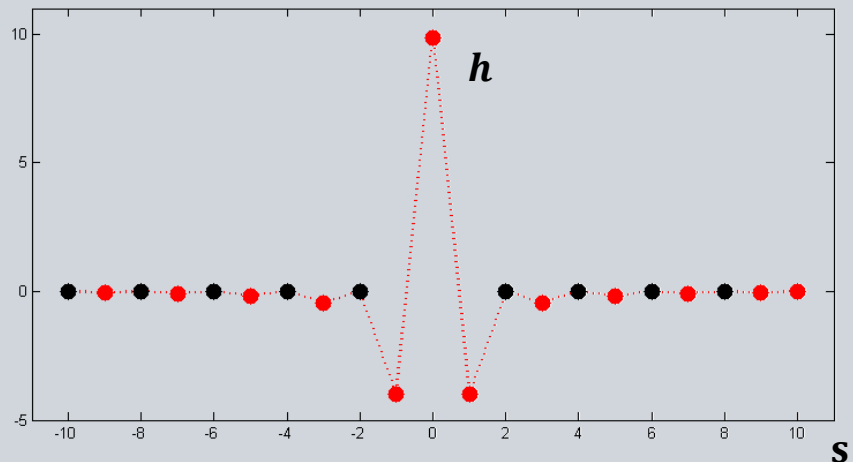
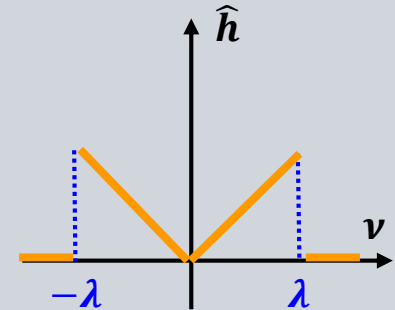
■ Modèle analytique — Rétro-projection filtrée

$$Hp = \text{TF}^{-1}(\hat{h} \hat{p})$$

$$Hp = h * p$$

$$h(s) = \int_{-\lambda}^{\lambda} \hat{h}(v) e^{ivs} dv$$

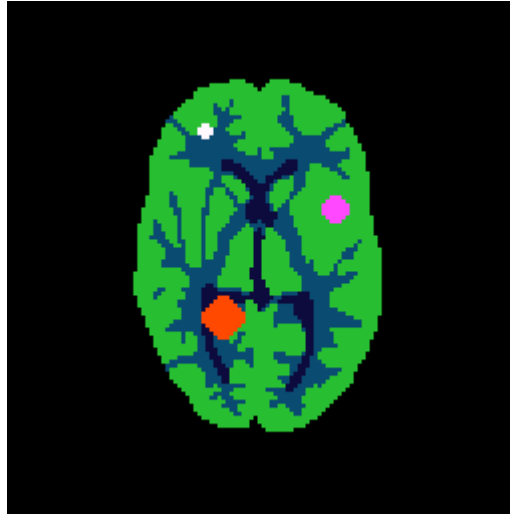
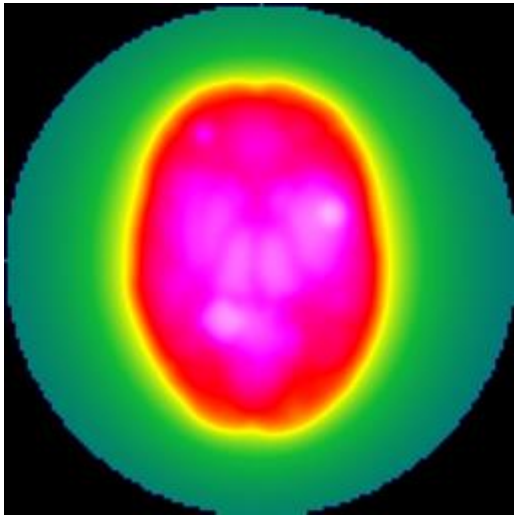
$$= \begin{cases} \pi^2 & \text{en } 0 \\ 0 & \text{pour } s \text{ pair} \\ -\frac{4}{s^2} & \text{pour } s \text{ impair} \end{cases}$$



Reconstruction

- Modèle analytique — Rétro-projection filtrée

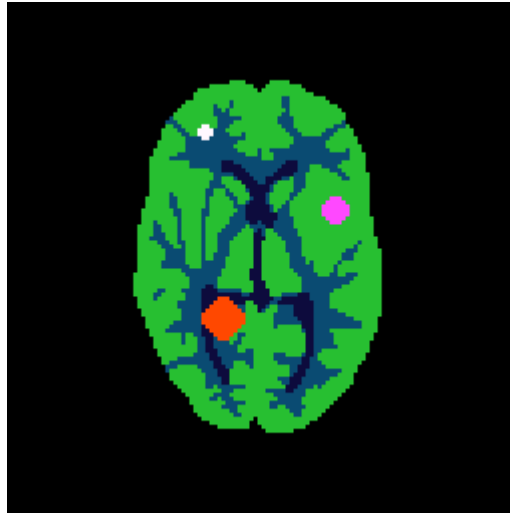
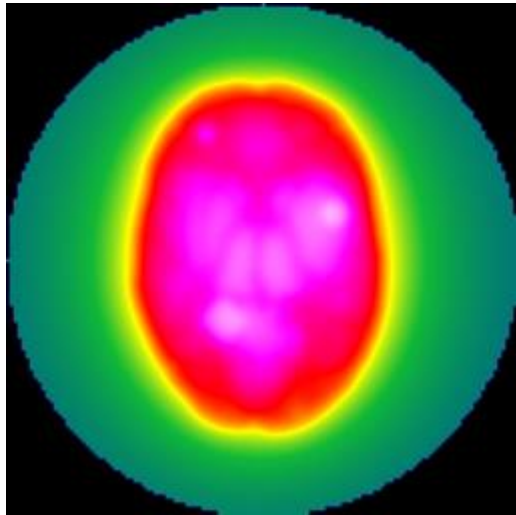
$$f = R^* p$$



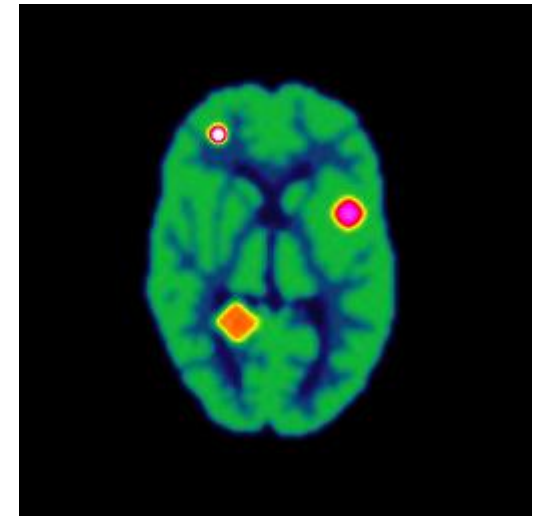
Reconstruction

- Modèle analytique — Rétro-projection filtrée

$$f = R^* p$$



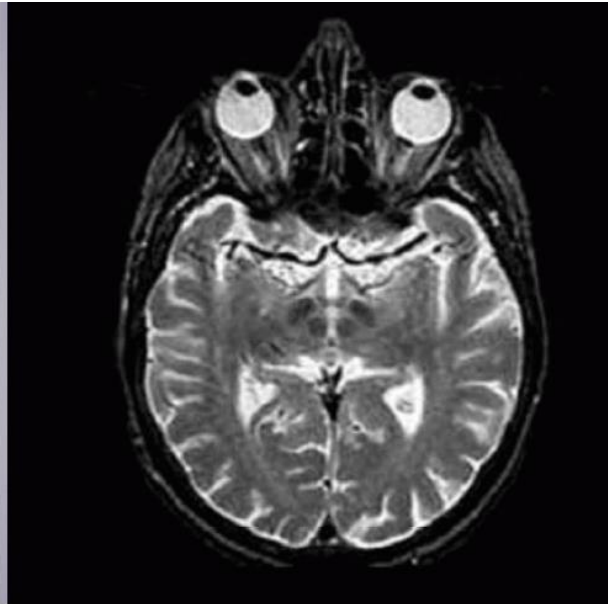
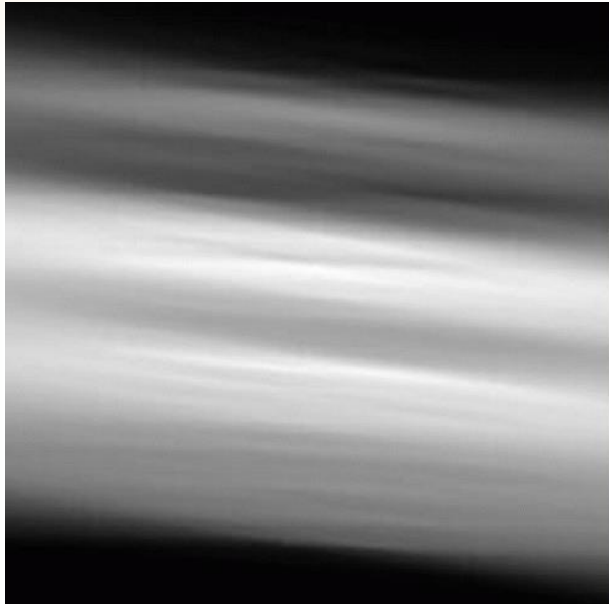
$$f = R^* H p$$



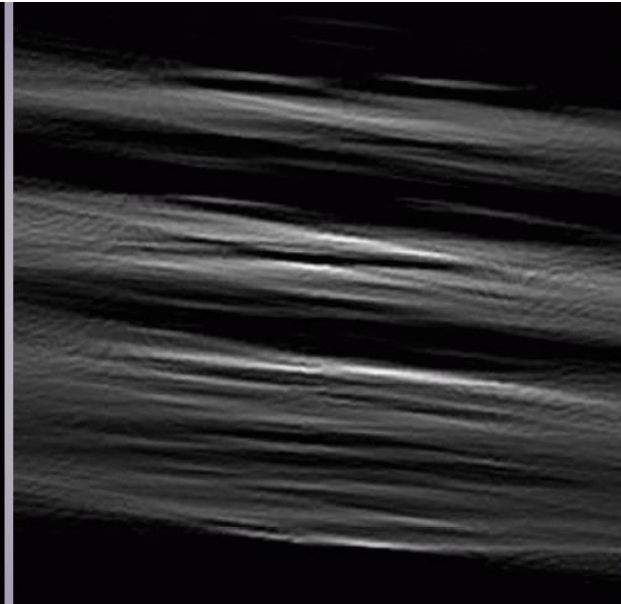
Reconstruction

- Modèle analytique — Rétro-projection filtrée

$$f = R^* p$$



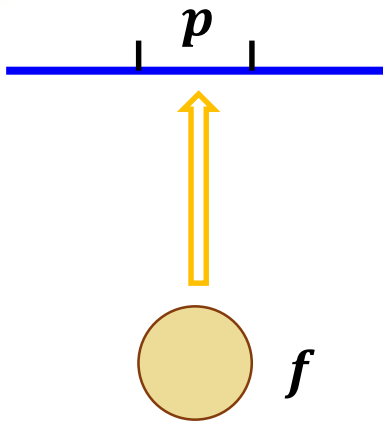
$$f = R^* H p$$



Reconstruction

- Modèle analytique — Limites

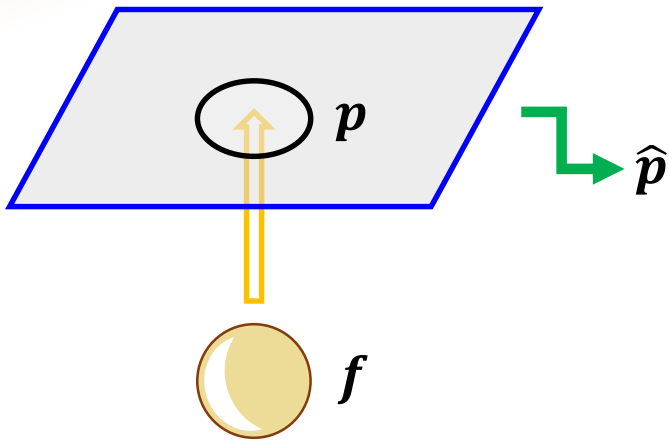
Troncature des données



Reconstruction

- Modèle analytique — Limites

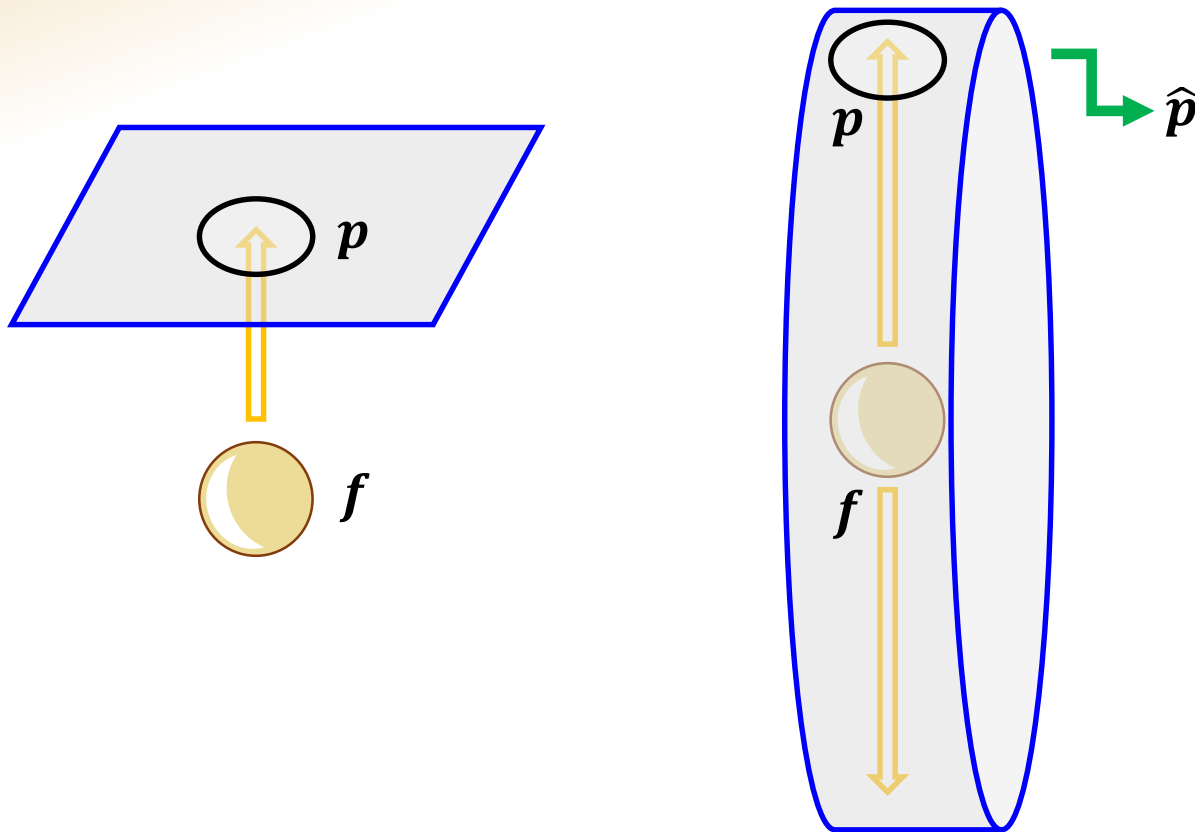
Troncature des données



Reconstruction

■ Modèle analytique — Limites

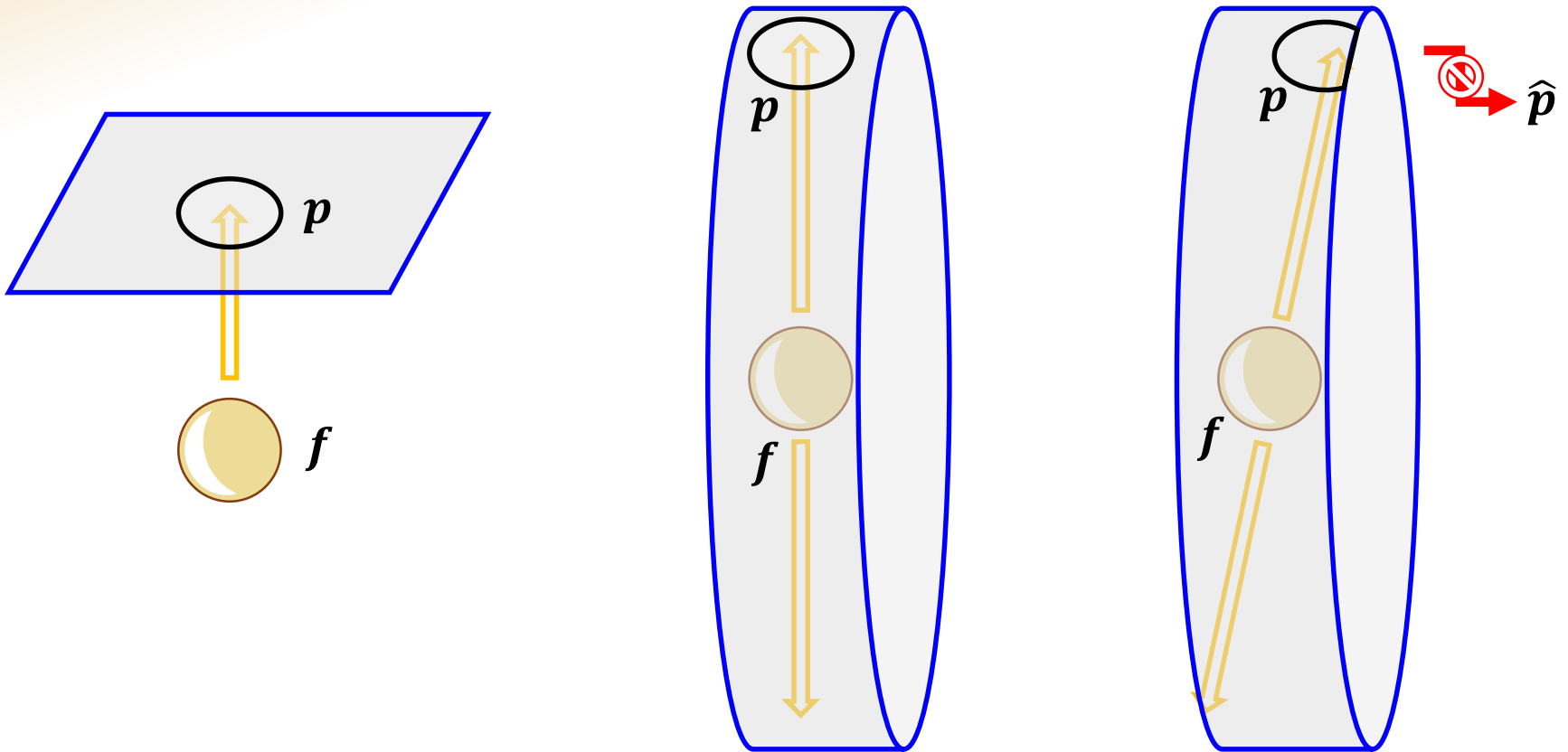
Troncature des données



Reconstruction

■ Modèle analytique — Limites

Troncature des données

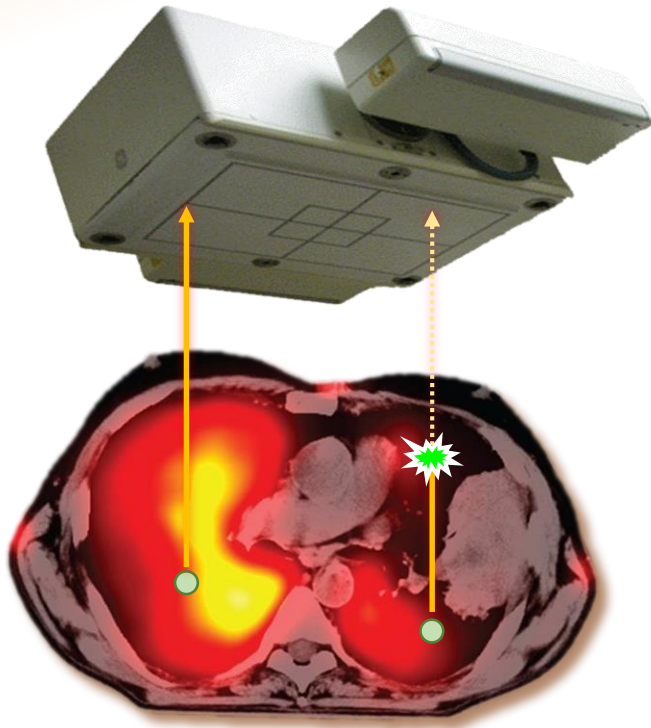


Reconstruction

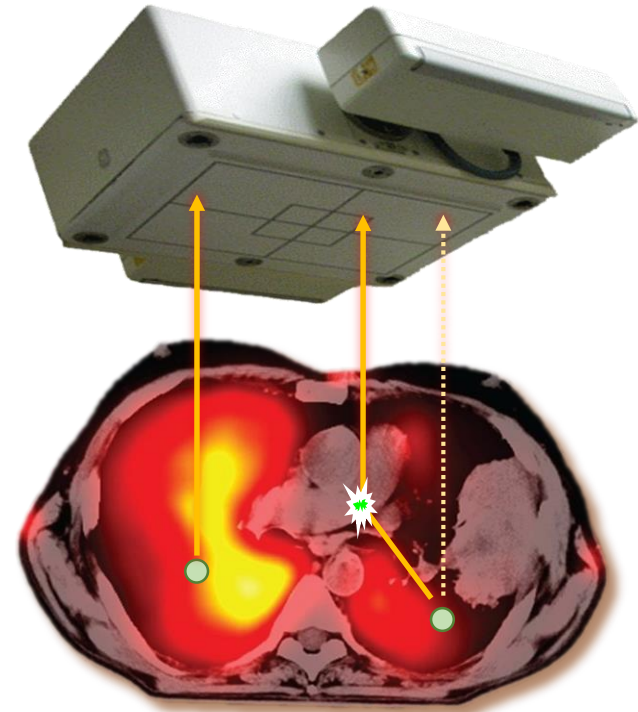
■ Modèle analytique — Limites

Interactions γ - matière

Absorption



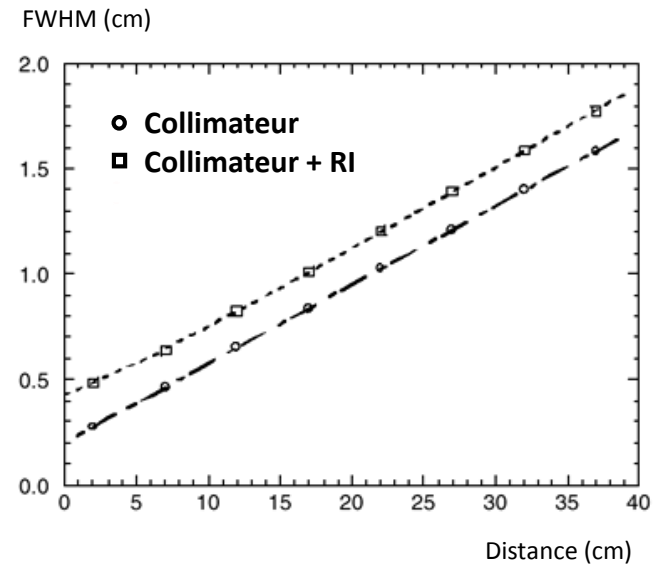
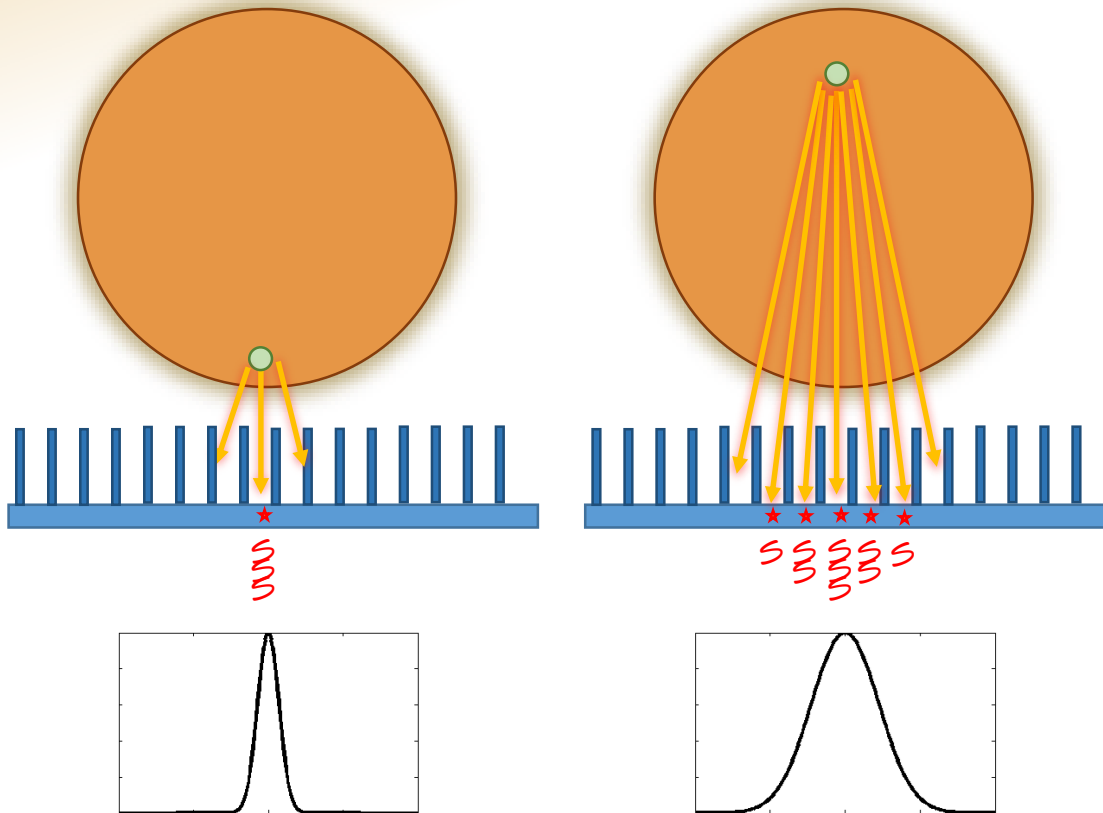
Diffusion



Reconstruction

■ Modèle analytique — Limites

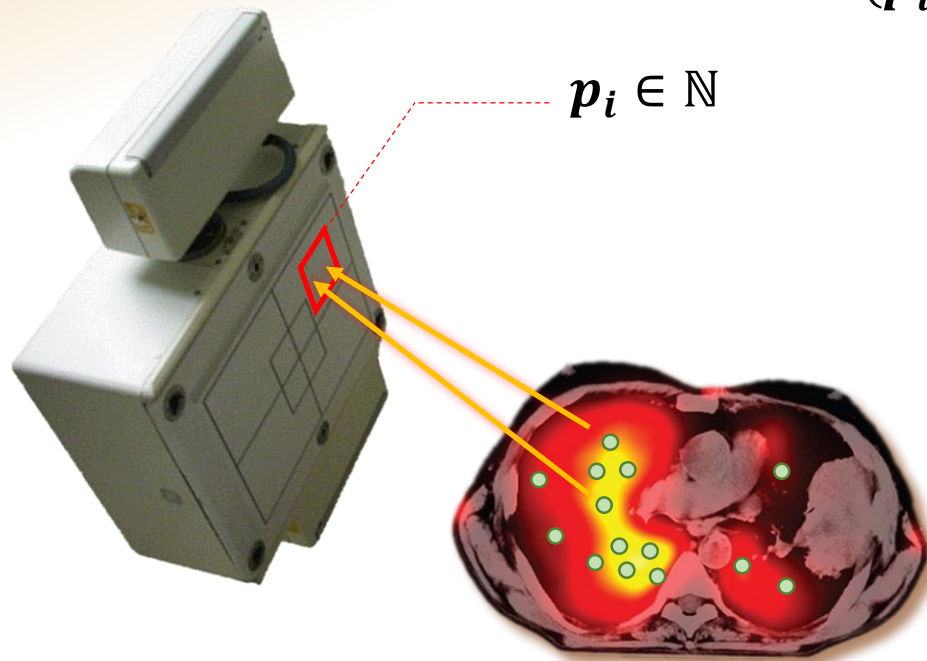
Réponse du détecteur



Reconstruction

■ Modèle analytique — Limites

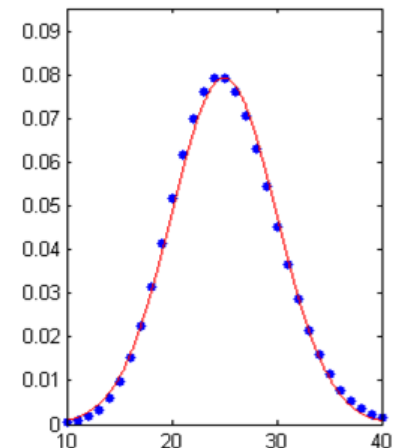
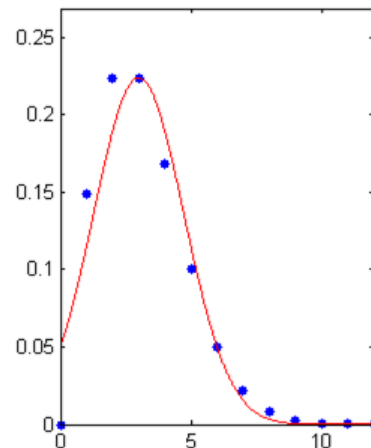
Bruit statistique



$$E(p_i) = \kappa \int_{\Gamma_i} f$$

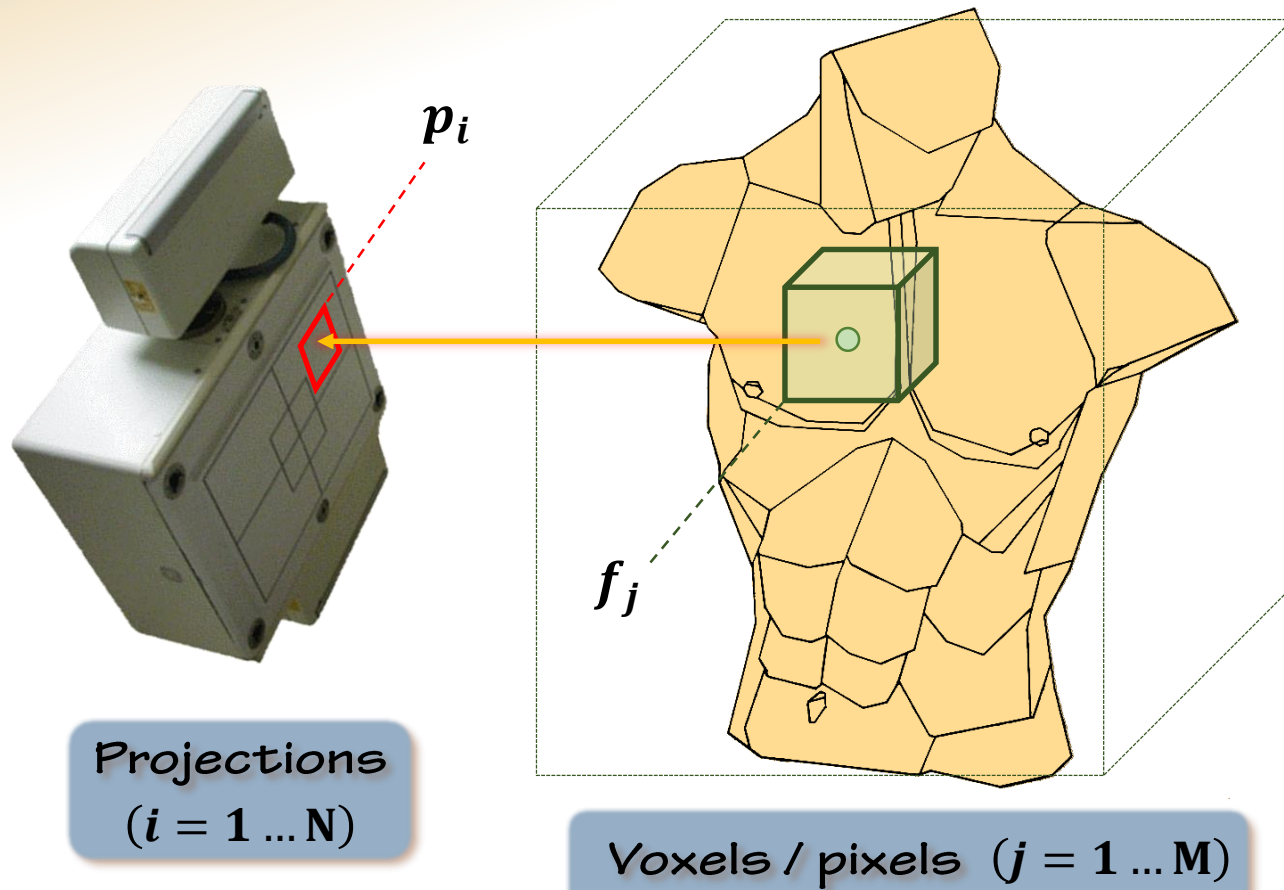
$$\text{Var}(p_i) = E(p_i)$$

$$\text{SNR} = \frac{E(p_i)}{\sigma(p_i)} \approx \sqrt{p_i}$$



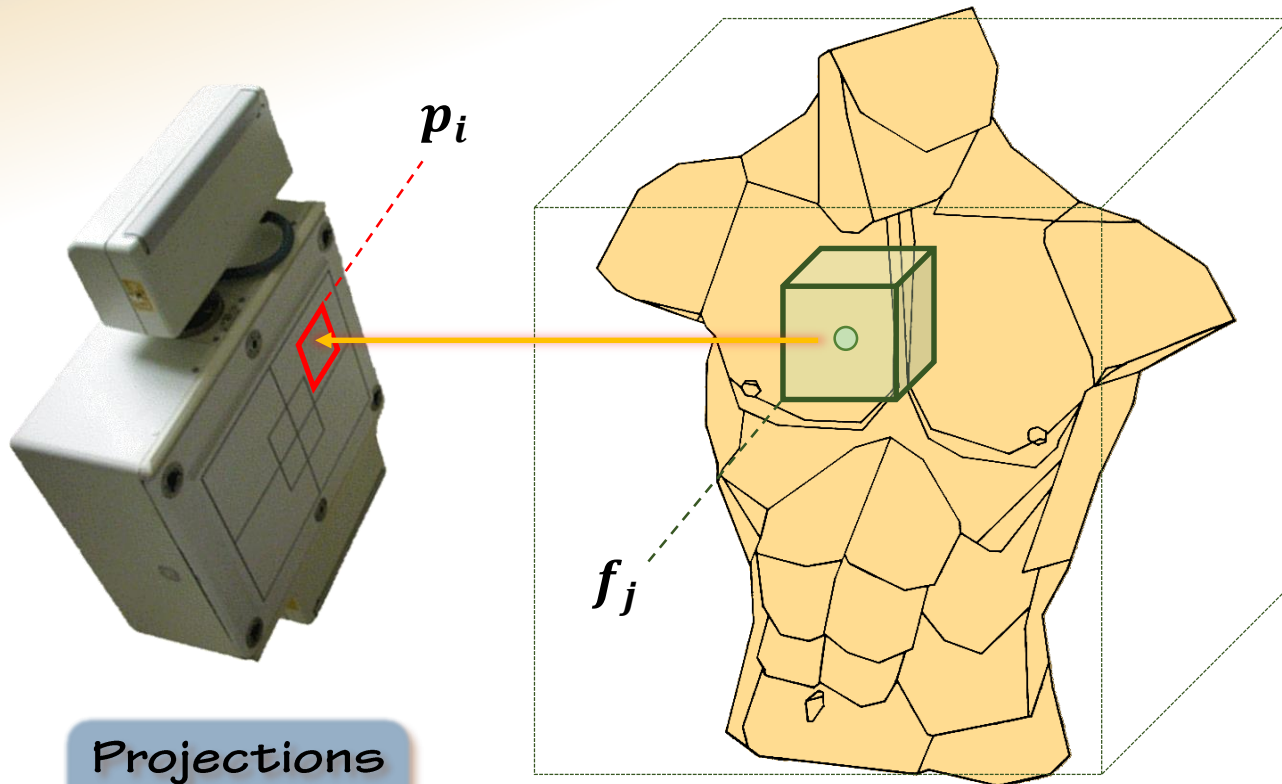
Reconstruction

■ Modèle algébrique



Reconstruction

■ Modèle algébrique



Projections
($i = 1 \dots N$)

Voxels / pixels ($j = 1 \dots M$)

Matrice système

$$\mathbf{R} \in \mathbb{R}^{N \times M}$$

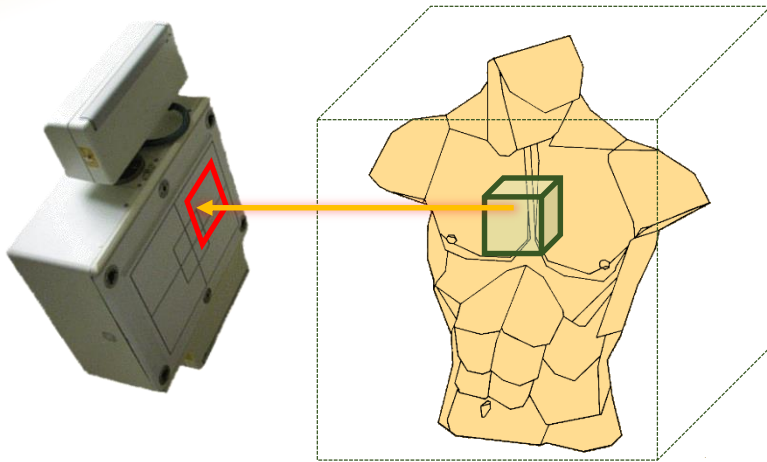
$$R_{ij} = \wp(j \rightarrow i)$$

- > Géométrie
- > Atténuation
- > Réponse du détecteur

Reconstruction

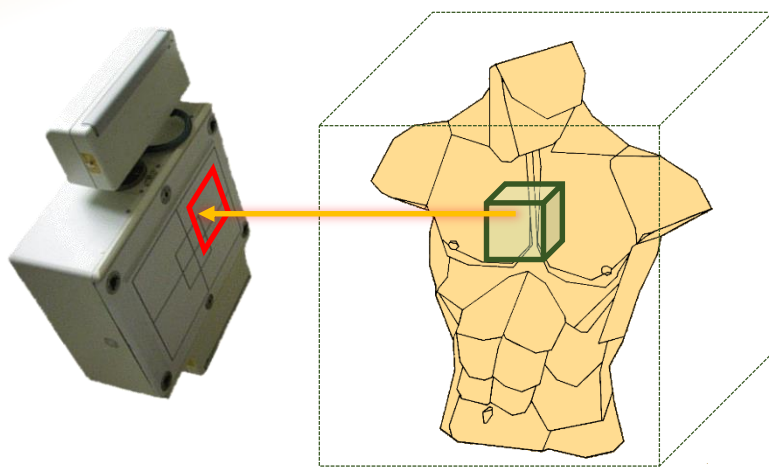
■ Modèle algébrique

$$p = Rf + s + n$$

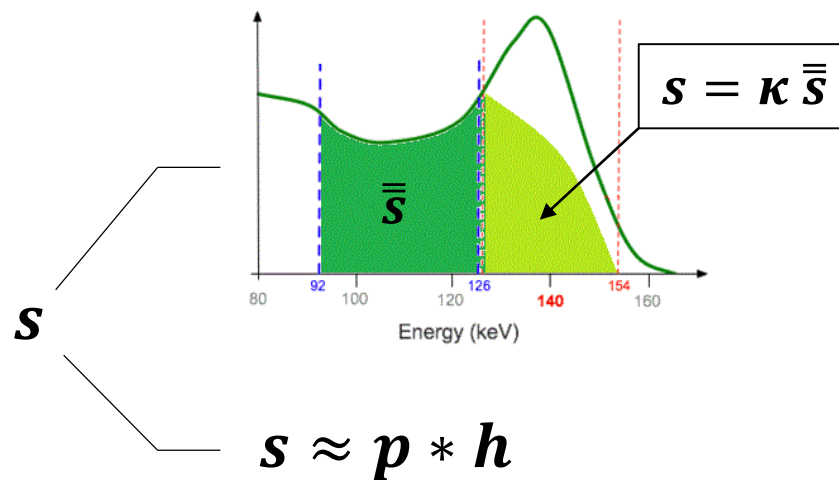


Reconstruction

■ Modèle algébrique

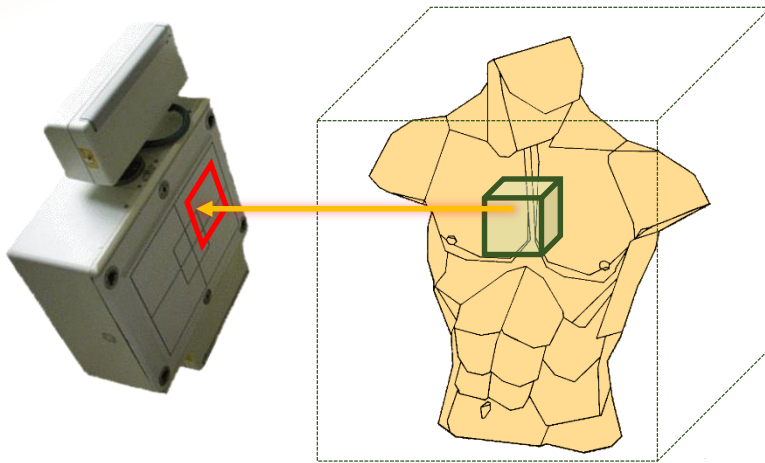


$$p = Rf + s + n$$

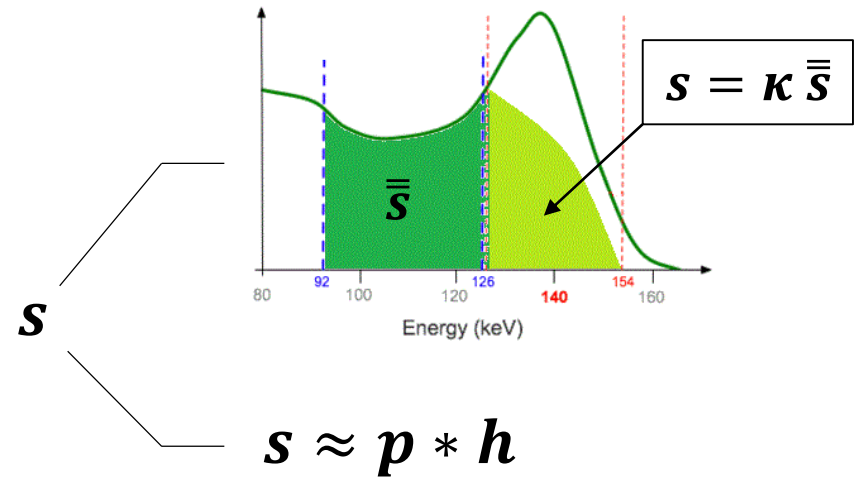


Reconstruction

■ Modèle algébrique



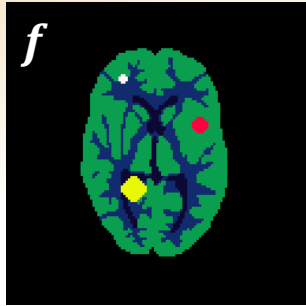
$$p = Rf + s + n$$



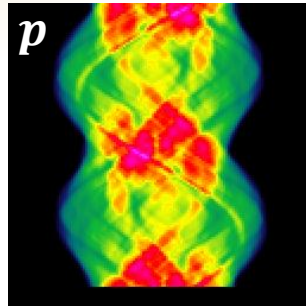
$$E(n) = 0 ; V(n) \approx p$$

Reconstruction

■ Modèle algébrique



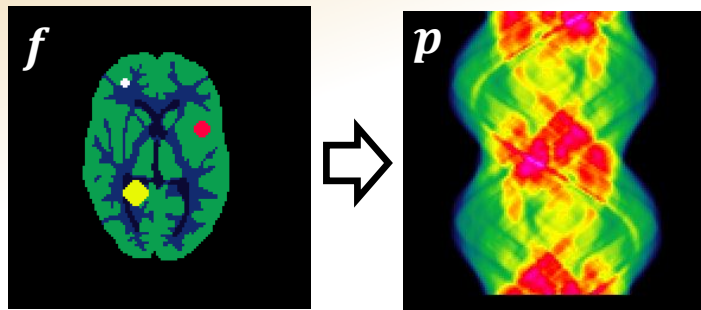
objet



$$p = Rf + n$$

Reconstruction

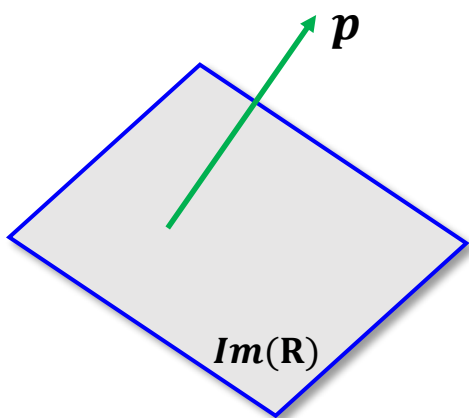
■ Modèle algébrique



objet

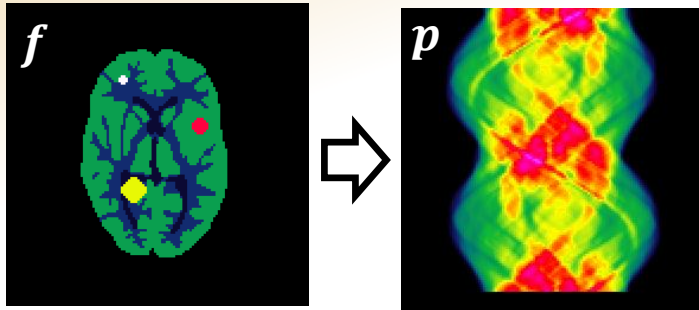
$$p = Rf + n$$

$$p \notin \text{Im}(R) \quad \nexists \bar{f} : R\bar{f} = p$$



Reconstruction

■ Modèle algébrique

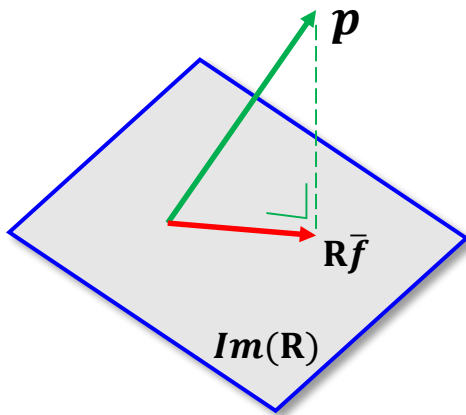


objet

$$p = \mathbf{R}f + n$$

$$p \notin \text{Im}(\mathbf{R}) \quad \nexists \bar{f} : \mathbf{R}\bar{f} = p$$

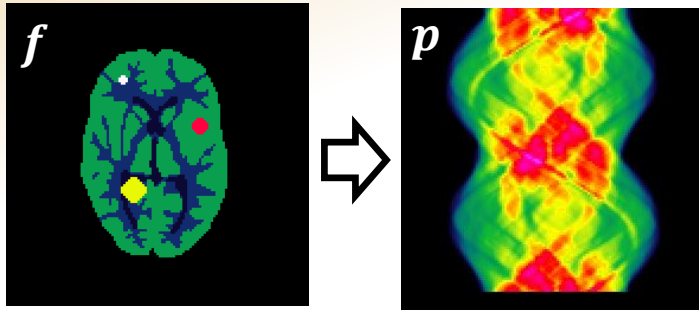
$$\bar{f} : \mathbf{R}\bar{f} = \mathbf{P}_{\text{Im}(\mathbf{R})}(p)$$



$$\bar{f} = \underset{f \in \Omega}{\text{argmin}} \{ \|\mathbf{R}f - p\| \}$$

Reconstruction

■ Modèle algébrique



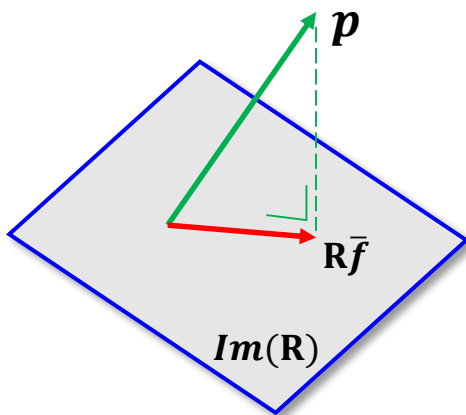
objet

$$p = Rf + n$$

$$p \notin \text{Im}(R) \quad \nexists \bar{f} : R\bar{f} = p$$

$$\bar{f} : R\bar{f} = P_{\text{Im}(R)}(p)$$

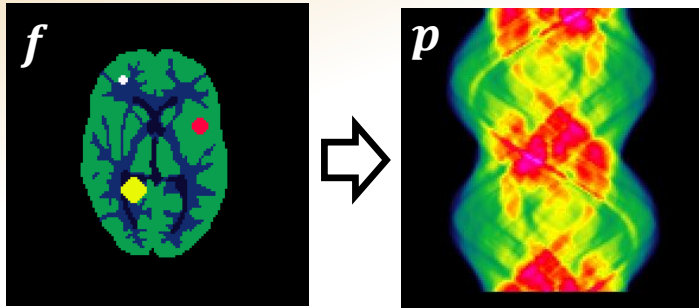
$$\forall w \in \text{Im}(R) : w^T R\bar{f} = w^T p$$



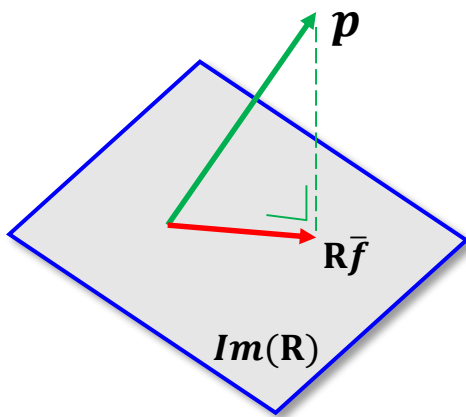
$$\bar{f} = \underset{f \in \Omega}{\text{argmin}} \{ \|Rf - p\| \}$$

Reconstruction

■ Modèle algébrique



objet



$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{ \| \mathbf{R}f - p \| \}$$

$$p = \mathbf{R}f + n$$

$$p \notin \operatorname{Im}(\mathbf{R}) \quad \nexists \bar{f} : \mathbf{R}\bar{f} = p$$

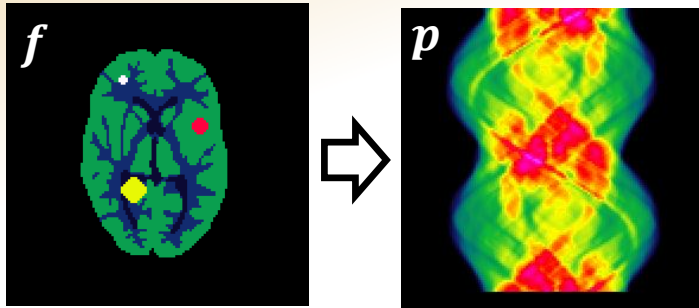
$$\bar{f} : \mathbf{R}\bar{f} = \mathbf{P}_{\operatorname{Im}(\mathbf{R})}(p)$$

$$\forall w \in \operatorname{Im}(\mathbf{R}) : w^T \mathbf{R}\bar{f} = w^T p$$

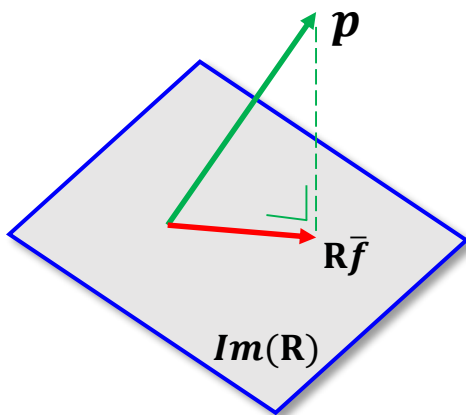
$$\forall g \in \Omega : (\mathbf{R}g)^T \mathbf{R}\bar{f} = (\mathbf{R}g)^T p$$

Reconstruction

■ Modèle algébrique



objet



$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{ \| \mathbf{R}f - p \| \}$$

$$p = \mathbf{R}f + n$$

$$p \notin \operatorname{Im}(\mathbf{R}) \quad \nexists \bar{f} : \mathbf{R}\bar{f} = p$$

$$\bar{f} : \mathbf{R}\bar{f} = \mathbf{P}_{\operatorname{Im}(\mathbf{R})}(p)$$

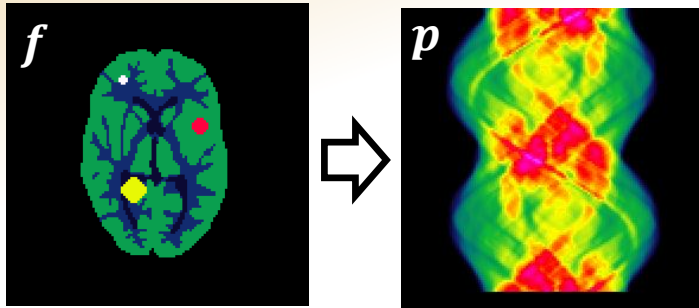
$$\forall w \in \operatorname{Im}(\mathbf{R}) : w^T \mathbf{R}\bar{f} = w^T p$$

$$\forall g \in \Omega : (\mathbf{R}g)^T \mathbf{R}\bar{f} = (\mathbf{R}g)^T p$$

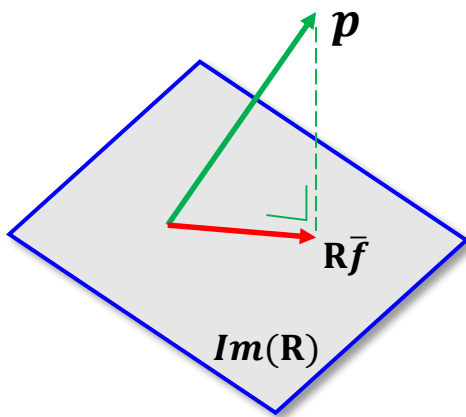
$$\forall g \in \Omega : g^T \mathbf{R}^* \mathbf{R}\bar{f} = g^T \mathbf{R}^* p$$

Reconstruction

■ Modèle algébrique



objet



$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{ \| \mathbf{R}f - p \| \}$$

$$p = \mathbf{R}f + n$$

$$p \notin \operatorname{Im}(\mathbf{R}) \quad \nexists \bar{f} : \mathbf{R}\bar{f} = p$$

$$\bar{f} : \mathbf{R}\bar{f} = \mathbf{P}_{\operatorname{Im}(\mathbf{R})}(p)$$

$$\forall w \in \operatorname{Im}(\mathbf{R}) : w^T \mathbf{R}\bar{f} = w^T p$$

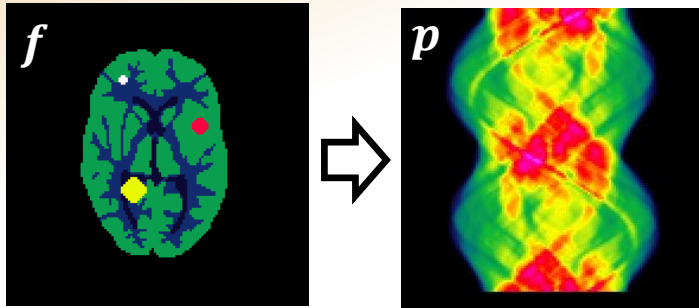
$$\forall g \in \Omega : (\mathbf{R}g)^T \mathbf{R}\bar{f} = (\mathbf{R}g)^T p$$

$$\forall g \in \Omega : g^T \mathbf{R}^* \mathbf{R}\bar{f} = g^T \mathbf{R}^* p$$

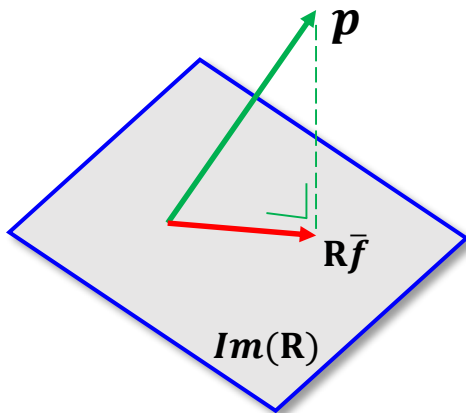
$$\mathbf{R}^* \mathbf{R}\bar{f} = \mathbf{R}^* p$$

Reconstruction

■ Modèle algébrique



objet



$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{ \| \mathbf{R}f - p \| \}$$

$$p = \mathbf{R}f + n$$

$$p \notin \operatorname{Im}(\mathbf{R}) \quad \nexists \bar{f} : \mathbf{R}\bar{f} = p$$

$$\bar{f} : \mathbf{R}\bar{f} = \mathbf{P}_{\operatorname{Im}(\mathbf{R})}(p)$$

$$\forall w \in \operatorname{Im}(\mathbf{R}) : w^T \mathbf{R}\bar{f} = w^T p$$

$$\forall g \in \Omega : (\mathbf{R}g)^T \mathbf{R}\bar{f} = (\mathbf{R}g)^T p$$

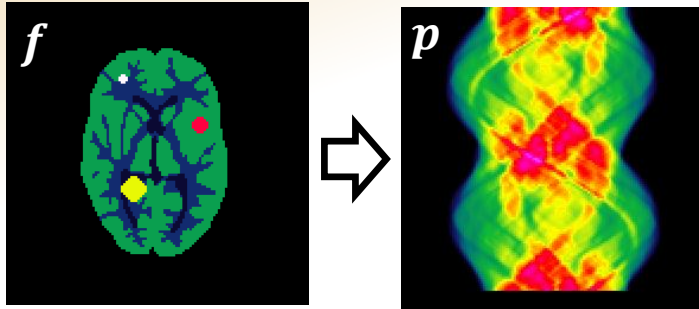
$$\forall g \in \Omega : g^T \mathbf{R}^* \mathbf{R}\bar{f} = g^T \mathbf{R}^* p$$

$$\mathbf{R}^* \mathbf{R}\bar{f} = \mathbf{R}^* p$$

$$\bar{f} = (\mathbf{R}^* \mathbf{R})^{-1} \mathbf{R}^* p$$

Reconstruction

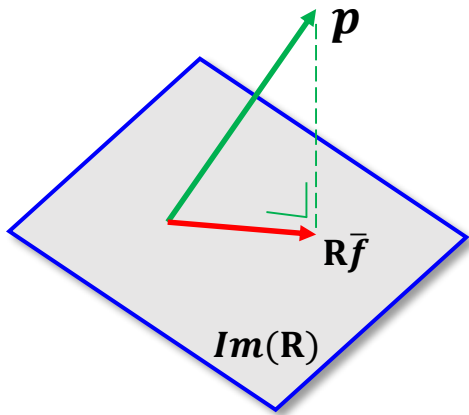
■ Modèle algébrique



objet

$$p = \mathbf{R}f + s + n$$

$$p \notin \text{Im}(\mathbf{R}) \quad \nexists \bar{f} : \mathbf{R}\bar{f} = p$$

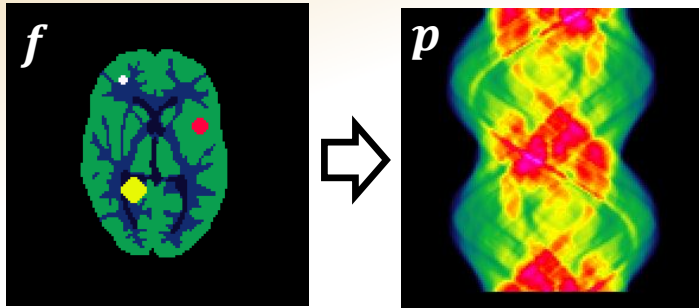


$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{ \|\mathbf{R}f - p\| \}$$

$$\bar{f} = (\mathbf{R}^* \mathbf{R})^{-1} \mathbf{R}^* p$$

Reconstruction

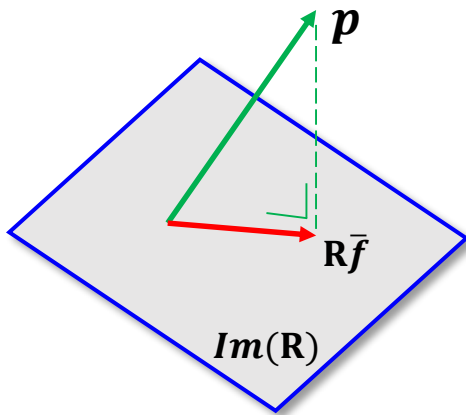
■ Modèle algébrique



objet

$$p = \mathbf{R}f + s + n$$

$$p \notin \text{Im}(\mathbf{R}) \quad \nexists \bar{f} : \mathbf{R}\bar{f} = p$$



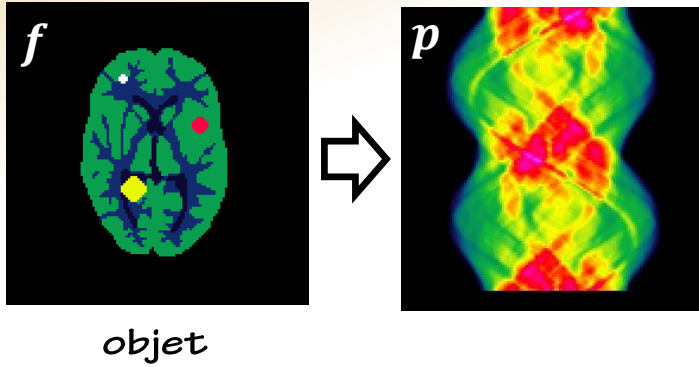
$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{ \|\mathbf{R}f - p\| \}$$

$$\bar{f} = (\mathbf{R}^* \mathbf{R})^{-1} \mathbf{R}^* p$$

$$\begin{aligned} \dim(\mathbf{R}^* \mathbf{R}) &= \sigma(10^5) \\ \kappa(\mathbf{R}^* \mathbf{R}) &\gg \end{aligned}$$

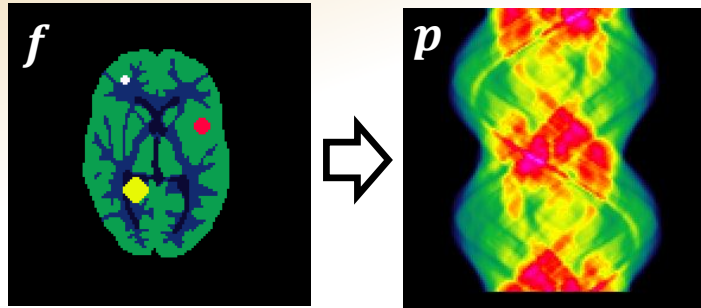
Reconstruction

■ Méthodes itératives

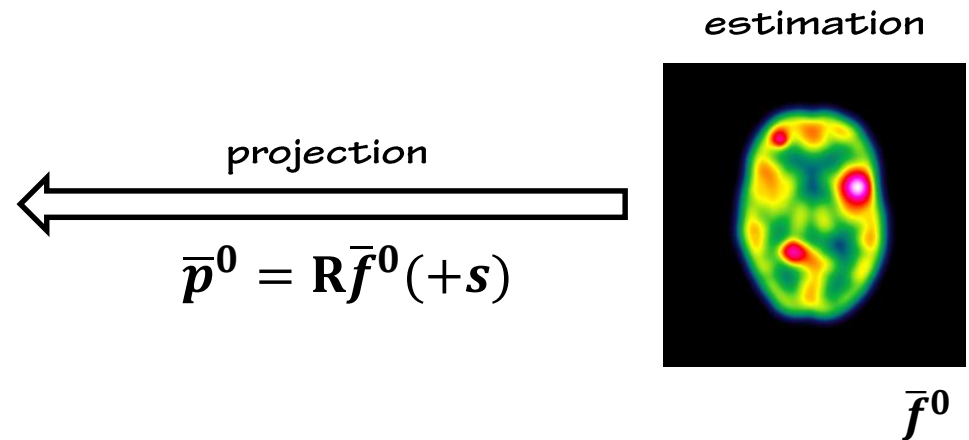


Reconstruction

■ Méthodes itératives

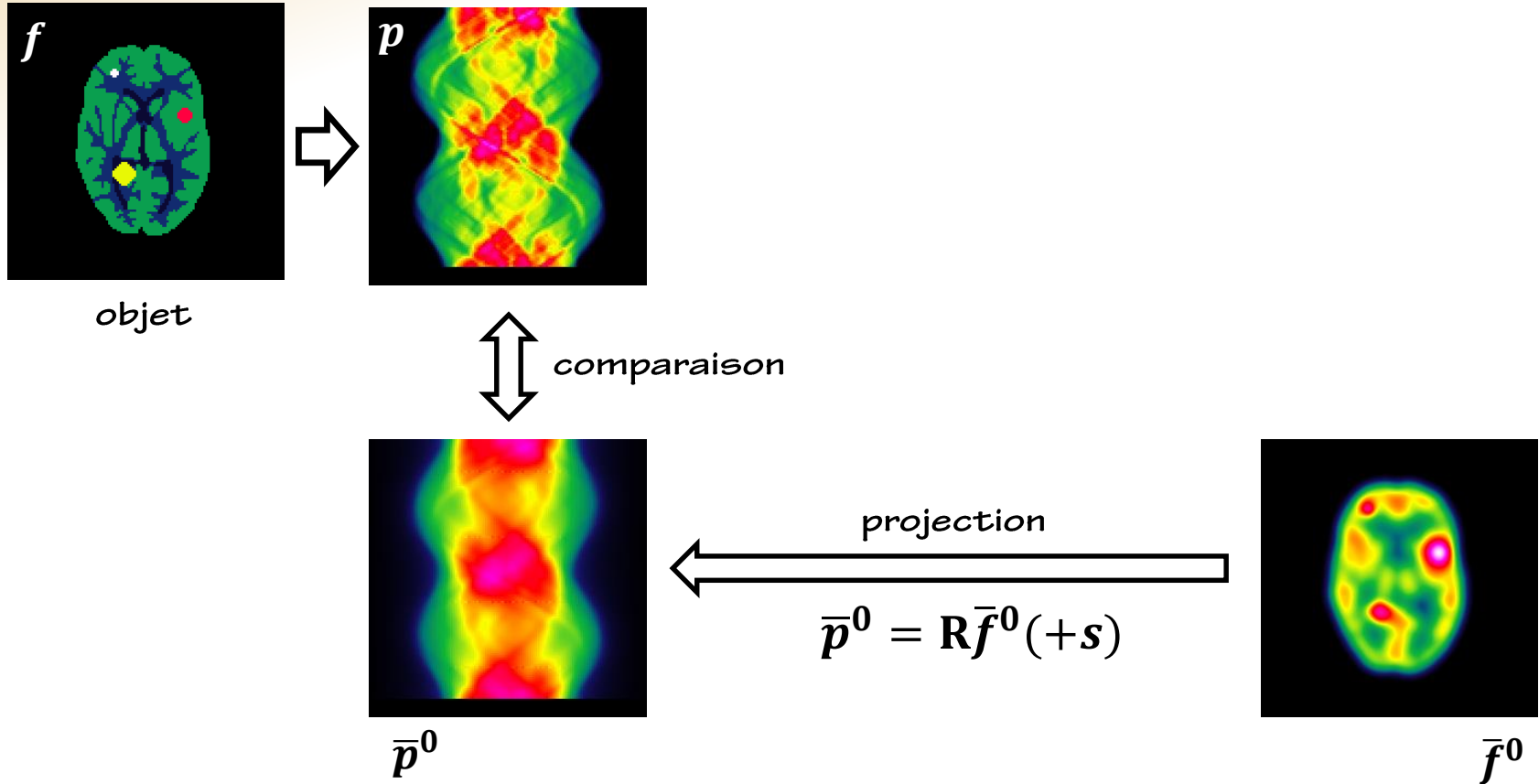


objet



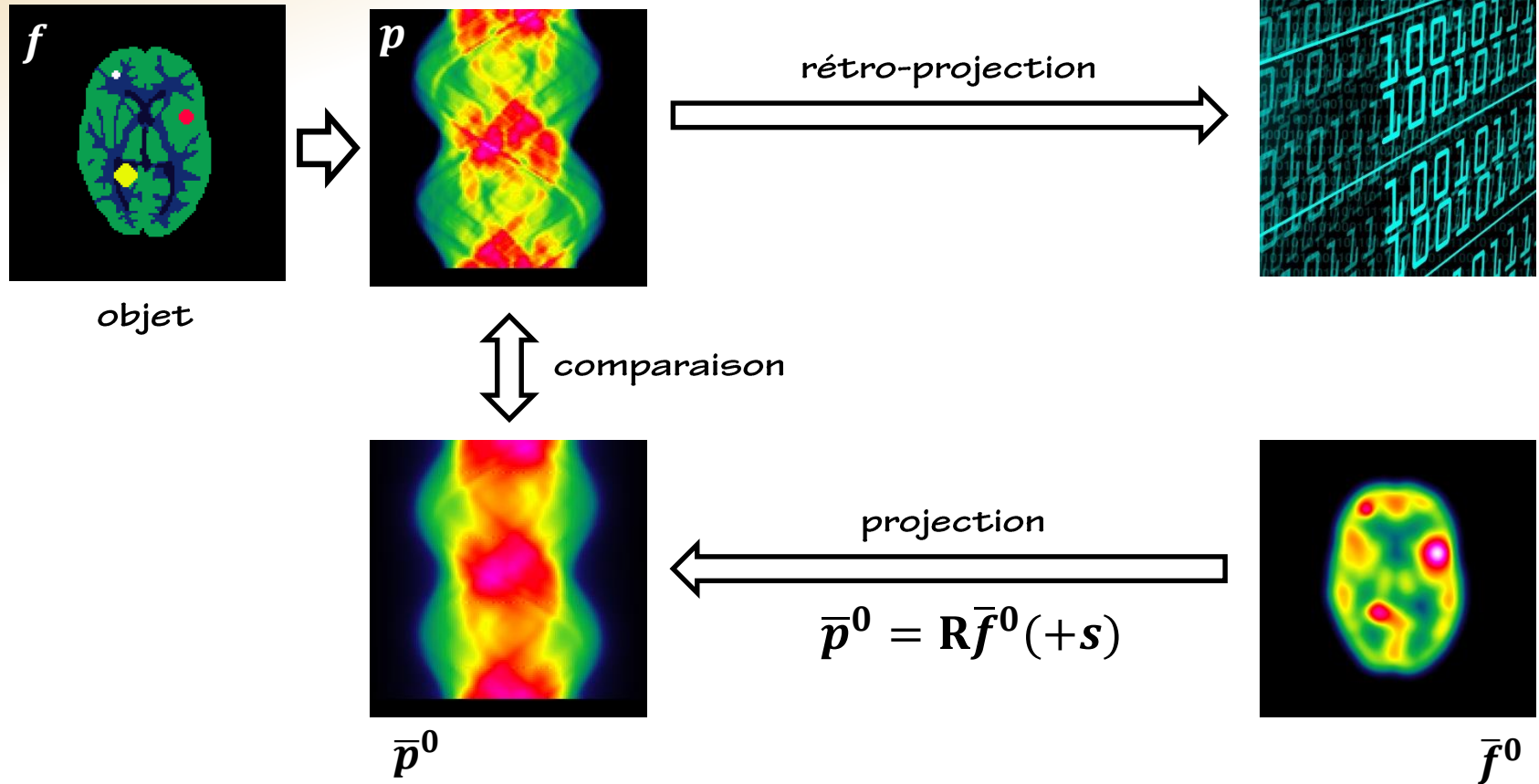
Reconstruction

■ Méthodes itératives



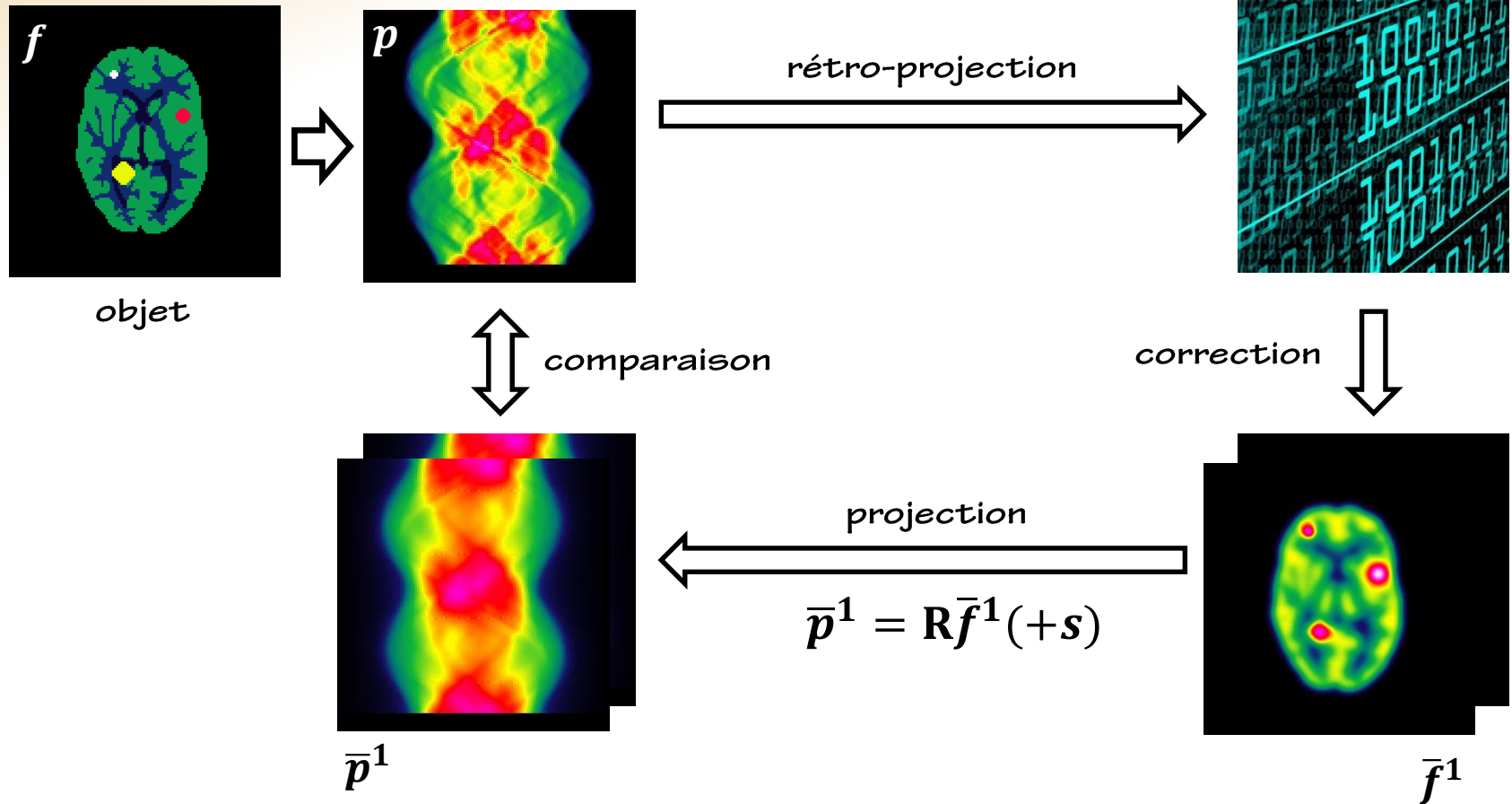
Reconstruction

■ Méthodes itératives



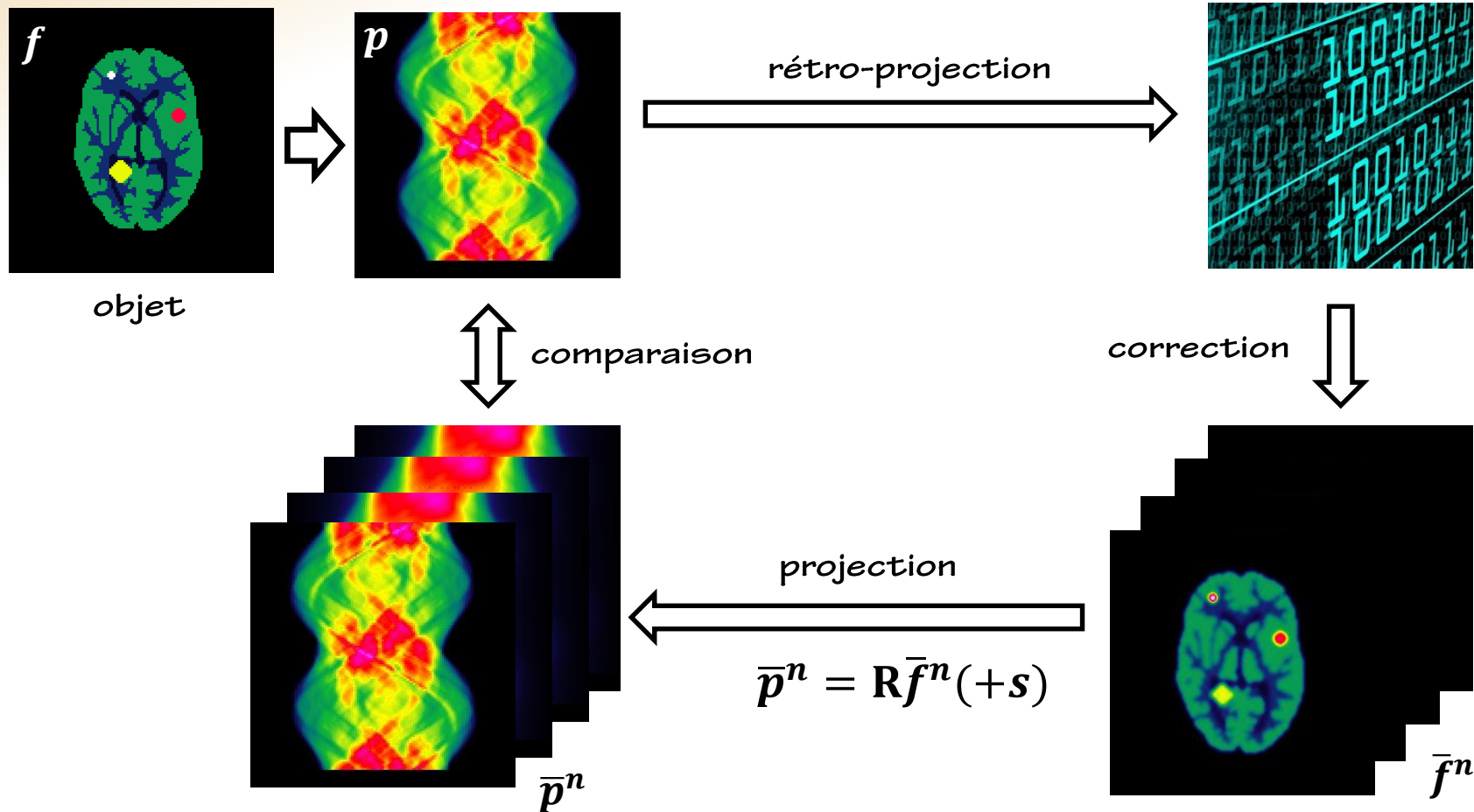
Reconstruction

■ Méthodes itératives



Reconstruction

■ Méthodes itératives



Reconstruction

■ Méthodes itératives



$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = \|\mathbf{R}f - \mathbf{p}\|^2$$

Méthode **LS**

Reconstruction

■ Méthodes itératives



Méthode **LS**

$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = \|\mathbf{R}f - \mathbf{p}\|^2$$

$$\bar{f} = (\mathbf{R}^* \mathbf{R})^{-1} \mathbf{R}^* \mathbf{p}$$

$$\dim(\mathbf{R}^* \mathbf{R}) = \sigma(10^5)$$

$$\kappa(\mathbf{R}^* \mathbf{R}) \gg$$

Reconstruction

■ Méthodes itératives



Méthode **LS**

$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = \|\mathbf{R}f - p\|^2$$

$$\bar{f} = (\mathbf{R}^* \mathbf{R})^{-1} \mathbf{R}^* p$$

$$\bar{f}^{n+1} = \bar{f}^n + \eta \mathbf{R}^* (p - \mathbf{R} \bar{f}^n)$$

Reconstruction

■ Méthodes itératives



Méthode **LS**

$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = \|\mathbf{R}f - \mathbf{p}\|^2$$

$$\bar{f} = (\mathbf{R}^* \mathbf{R})^{-1} \mathbf{R}^* \mathbf{p}$$

$$\bar{f}^{n+1} = \bar{f}^n + \eta \mathbf{R}^* (\mathbf{p} - \mathbf{R} \bar{f}^n)$$

correction additive

Reconstruction

■ Méthodes itératives



Méthode LS

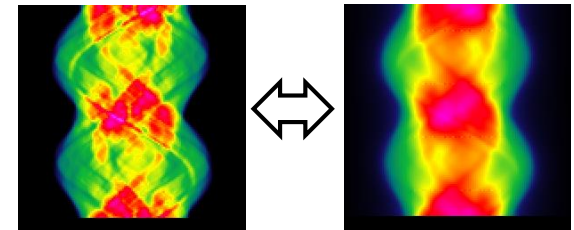
$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = \|\mathbf{R}f - \mathbf{p}\|^2$$

$$\bar{f} = (\mathbf{R}^* \mathbf{R})^{-1} \mathbf{R}^* \mathbf{p}$$

$$\bar{f}^{n+1} = \bar{f}^n + \eta \mathbf{R}^* (\mathbf{p} - \mathbf{R} \bar{f}^n)$$

comparaison



Reconstruction

■ Méthodes itératives



Méthode **LS**

$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{J(f)\}$$

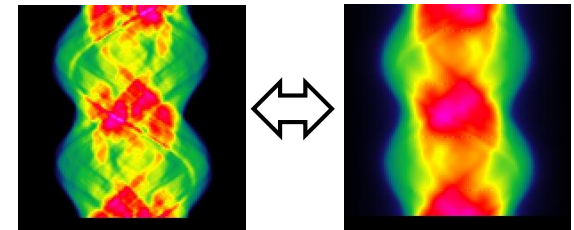
$$J(f) = \|\mathbf{R}f - \mathbf{p}\|^2$$

$$\bar{f} = (\mathbf{R}^* \mathbf{R})^{-1} \mathbf{R}^* \mathbf{p}$$

$$\bar{f}^{n+1} = \bar{f}^n + \eta \mathbf{R}^* (\mathbf{p} - \mathbf{R} \bar{f}^n)$$

rétro-projection

comparaison



Reconstruction

■ Méthodes itératives



Méthode ML

$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{ J(f) \}$$

$$J(f) = -\log \{ \wp(p|f) \}$$

Reconstruction

■ Méthodes itératives



Méthode ML

$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{ J(f) \}$$

$$J(f) = -\log \{ \wp(\mathbf{p}|f) \}$$

$$\wp(\mathbf{p}|f) = \prod_i \frac{e^{-\check{p}_i} \check{p}_i^{p_i}}{p_i!}$$

$$\check{p}_i = (\mathbf{R}f)_i$$

Reconstruction

■ Méthodes itératives



Méthode ML

$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = -\log\{\varphi(\mathbf{p}|f)\}$$

$$J(f) = \sum_i \{\mathbf{R}f_i - p_i \log(\mathbf{R}f_i)\}$$

$$\varphi(\mathbf{p}|f) = \prod_i \frac{e^{-\check{p}_i} \check{p}_i^{p_i}}{p_i!}$$
$$\check{p}_i = (\mathbf{R}f)_i$$

Reconstruction

■ Méthodes itératives



Méthode ML

$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = -\log\{\varphi(p|f)\}$$

$$J(f) = \sum_i \{Rf_i - p_i \log(Rf_i)\}$$

$$\bar{f}^{n+1} = \bar{f}^n \times \left(\mathbf{R}^* \frac{p}{\mathbf{R} \bar{f}^n} \right)$$

Reconstruction

■ Méthodes itératives



Méthode ML

$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = -\log\{\varphi(p|f)\}$$

$$J(f) = \sum_i \{Rf_i - p_i \log(Rf_i)\}$$

$$\bar{f}^{n+1} = \bar{f}^n \times \left(\mathbf{R}^* \frac{p}{\mathbf{R} \bar{f}^n} \right)$$

correction multiplicative

Reconstruction

■ Méthodes itératives



Méthode ML

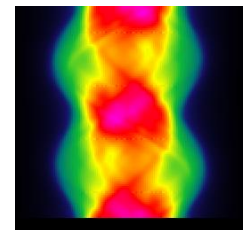
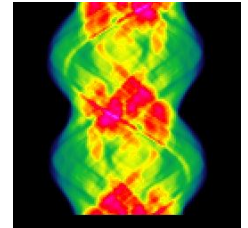
$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = -\log\{\varphi(p|f)\}$$

$$J(f) = \sum_i \{Rf_i - p_i \log(Rf_i)\}$$

$$\bar{f}^{n+1} = \bar{f}^n \times \left(R^* \frac{p}{R \bar{f}^n} \right)$$

comparaison



Reconstruction

■ Méthodes itératives



Méthode ML

$$\bar{f} = \underset{f \in \Omega}{\operatorname{argmin}} \{ J(f) \}$$

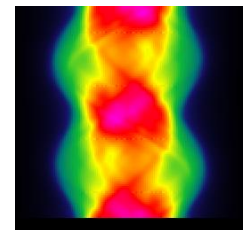
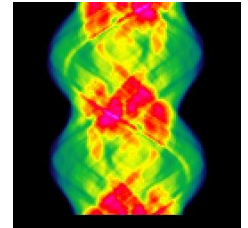
$$J(f) = -\log \{ \wp(p|f) \}$$

$$J(f) = \sum_i \{ Rf_i - p_i \log(Rf_i) \}$$

$$\bar{f}^{n+1} = \bar{f}^n \times \left(\boxed{\mathbf{R}^*} \frac{\mathbf{p}}{\mathbf{R} \bar{f}^n} \right)$$

rétro-projection

comparaison



Régularisation

$$p = Rf$$

R est mal conditionné : $\kappa(\mathbf{R}) \gg$

Régularisation

$$\mathbf{p} = \mathbf{R}f$$

\mathbf{R} est mal conditionné : $\kappa(\mathbf{R}) \gg$

$$\frac{\|\Delta f\|}{\|f\|} \leq \kappa(\mathbf{R}) \frac{\|\Delta p\|}{\|p\|} \approx \sigma\left(\frac{1}{\sqrt{p}}\right)$$

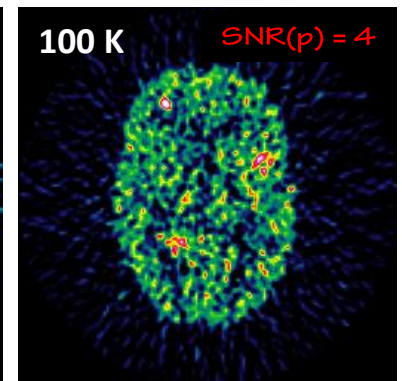
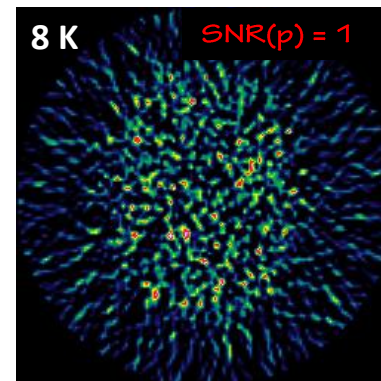
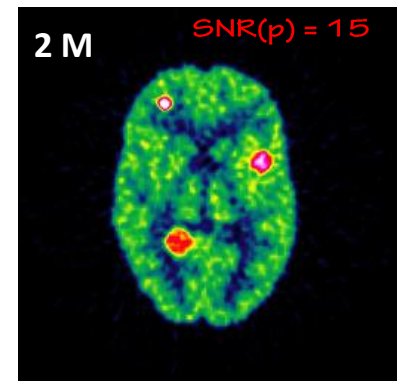
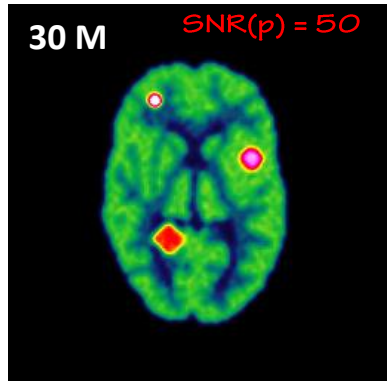
Régularisation

$$p = Rf$$

R est mal conditionné : $\kappa(\mathbf{R}) \gg$

$$\frac{\|\Delta f\|}{\|f\|} \leq \kappa(\mathbf{R}) \frac{\|\Delta p\|}{\|p\|} \approx \sigma\left(\frac{1}{\sqrt{p}}\right)$$

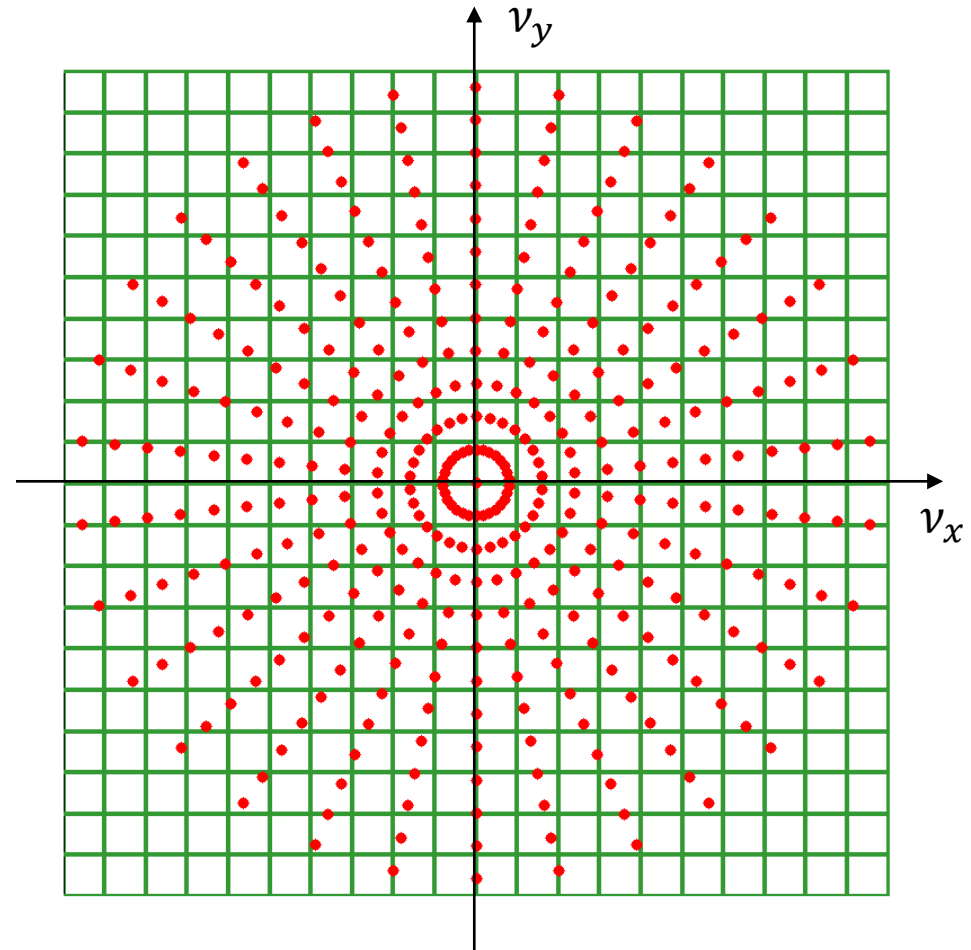
⇒ Sensibilité au bruit



Régularisation

■ Reconstruction analytique

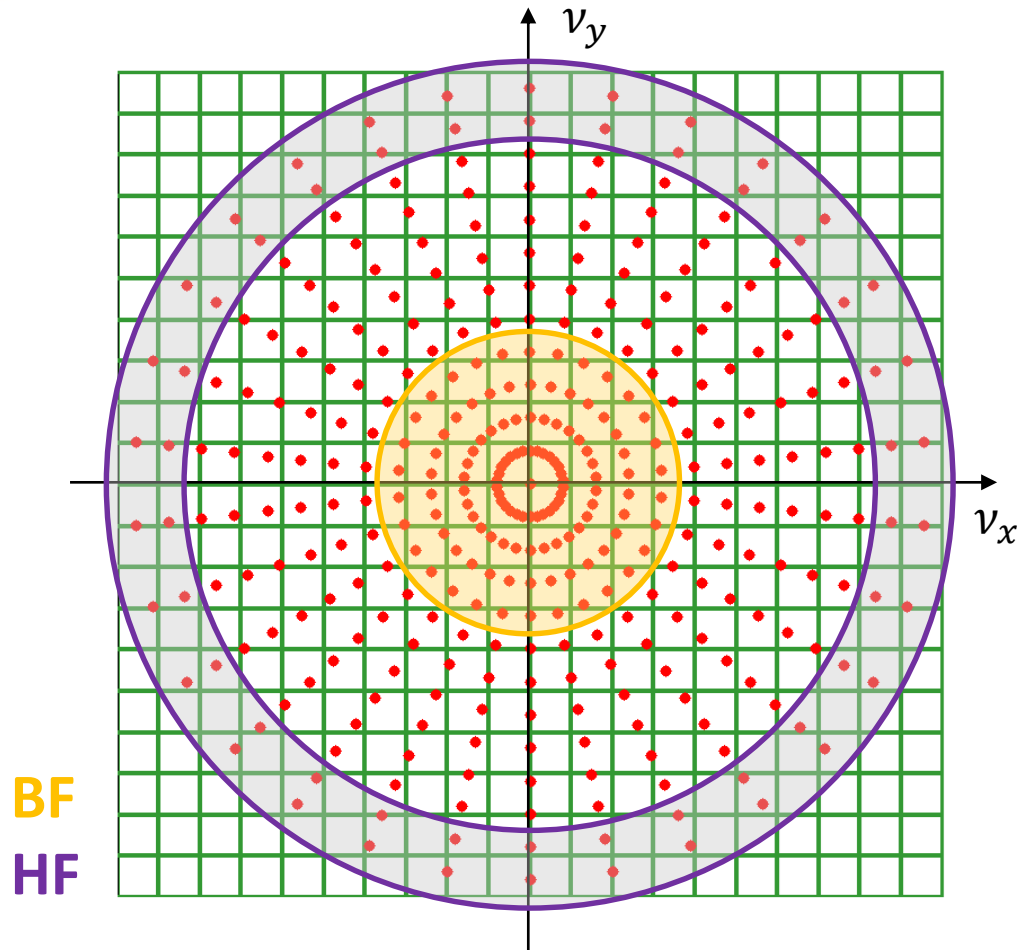
$$\hat{f}(v \cos \theta, v \sin \theta) = \hat{p}_\theta(v)$$



Régularisation

■ Reconstruction analytique

$$\hat{f}(v \cos \theta, v \sin \theta) = \hat{p}_\theta(v)$$

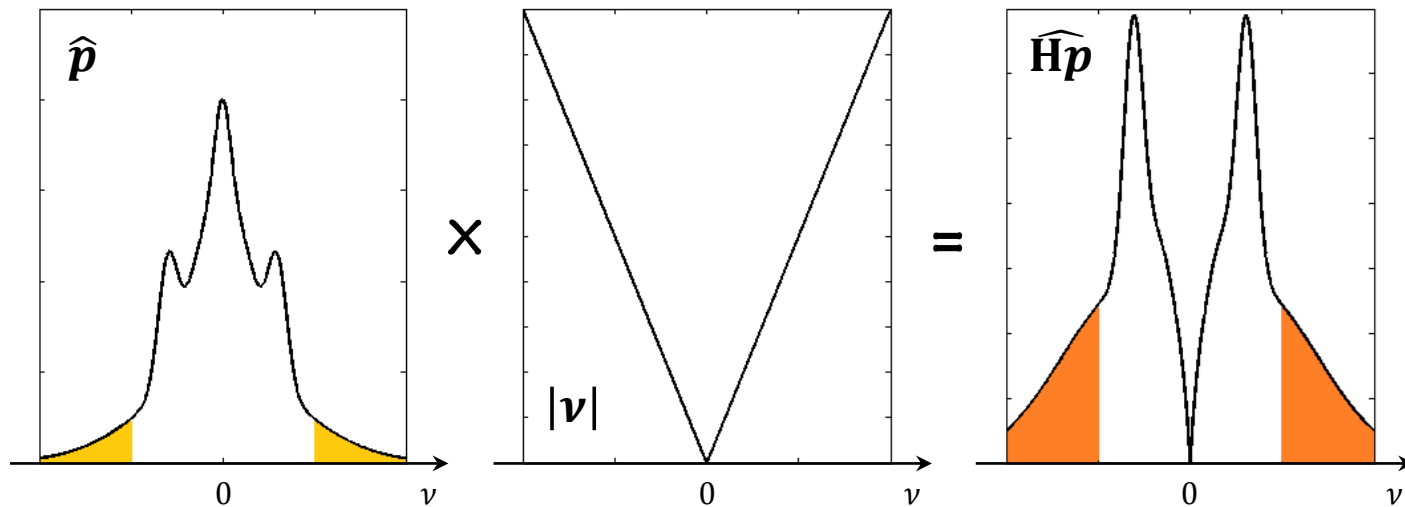
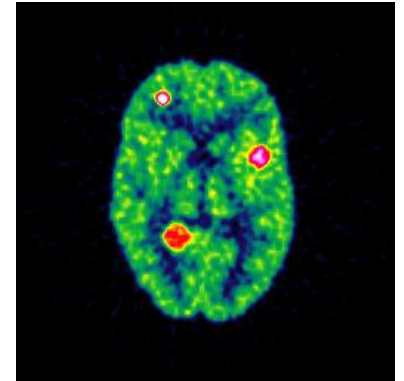


Régularisation

■ Reconstruction analytique

$$f = R^* H p$$

$$\widehat{H} p = |v| \widehat{p}$$

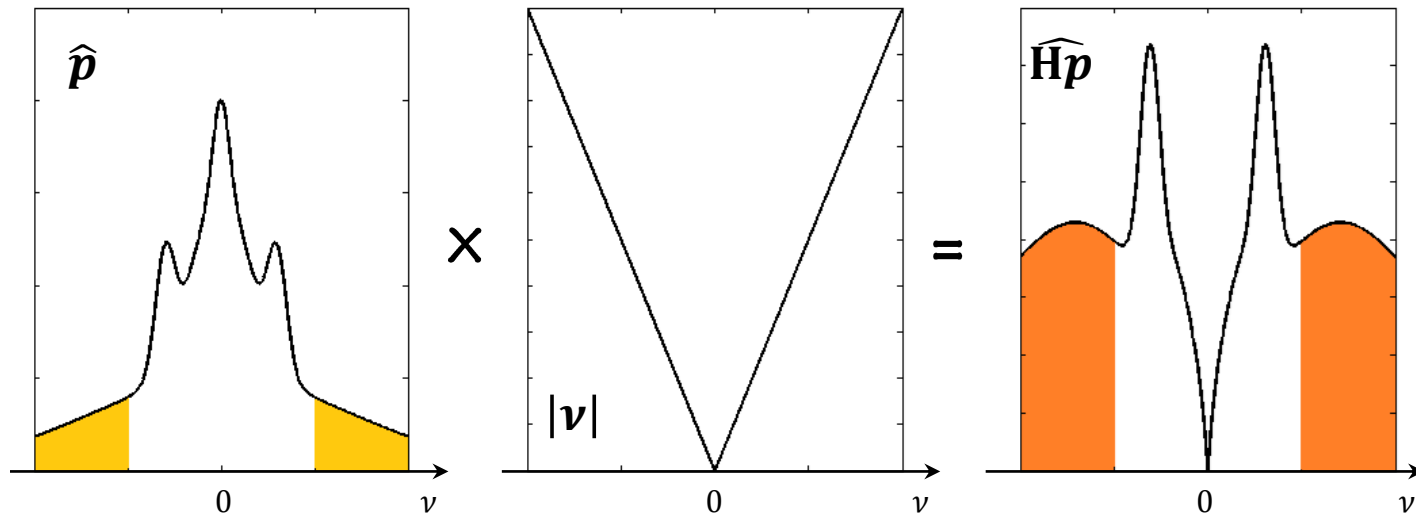
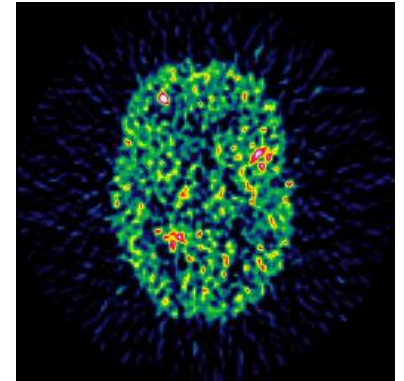


Régularisation

■ Reconstruction analytique

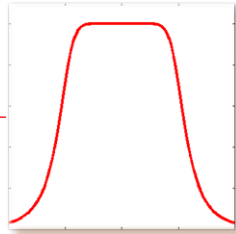
$$f = R^* H p$$

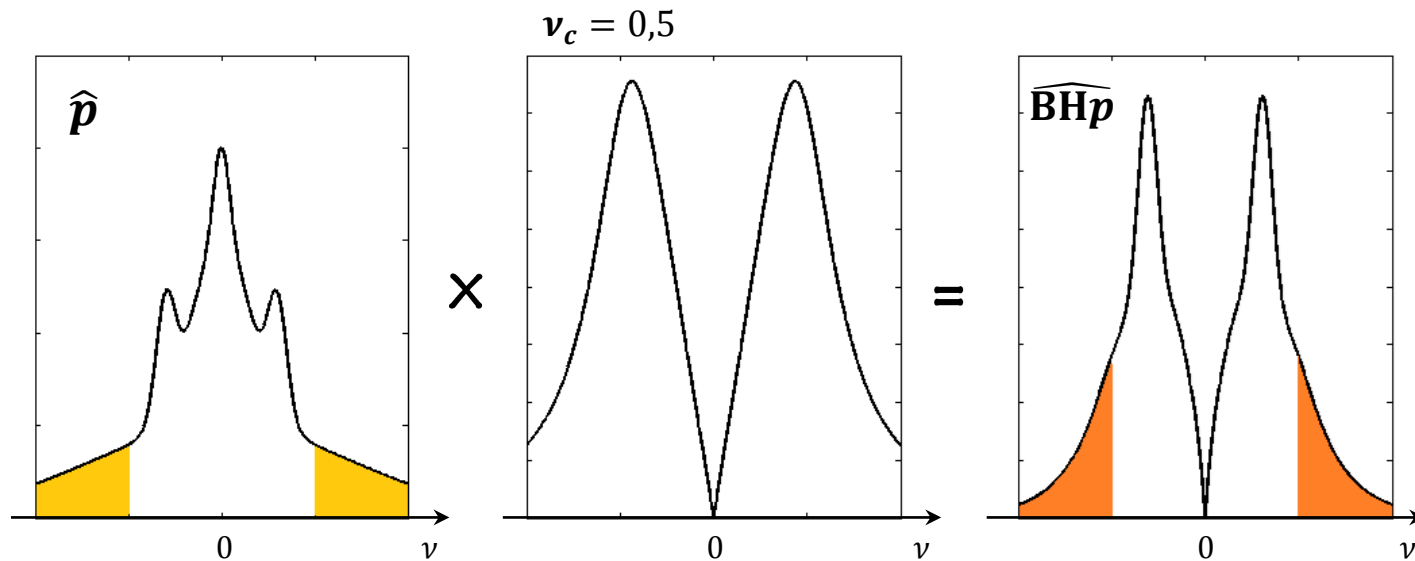
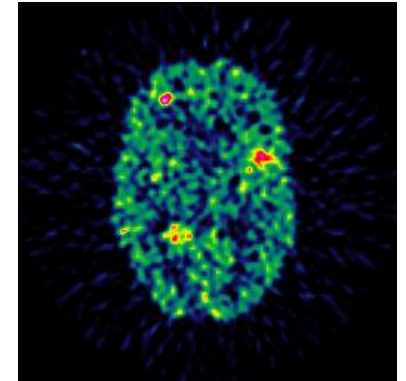
$$\widehat{H} p = |v| \widehat{p}$$



Régularisation

■ Reconstruction analytique

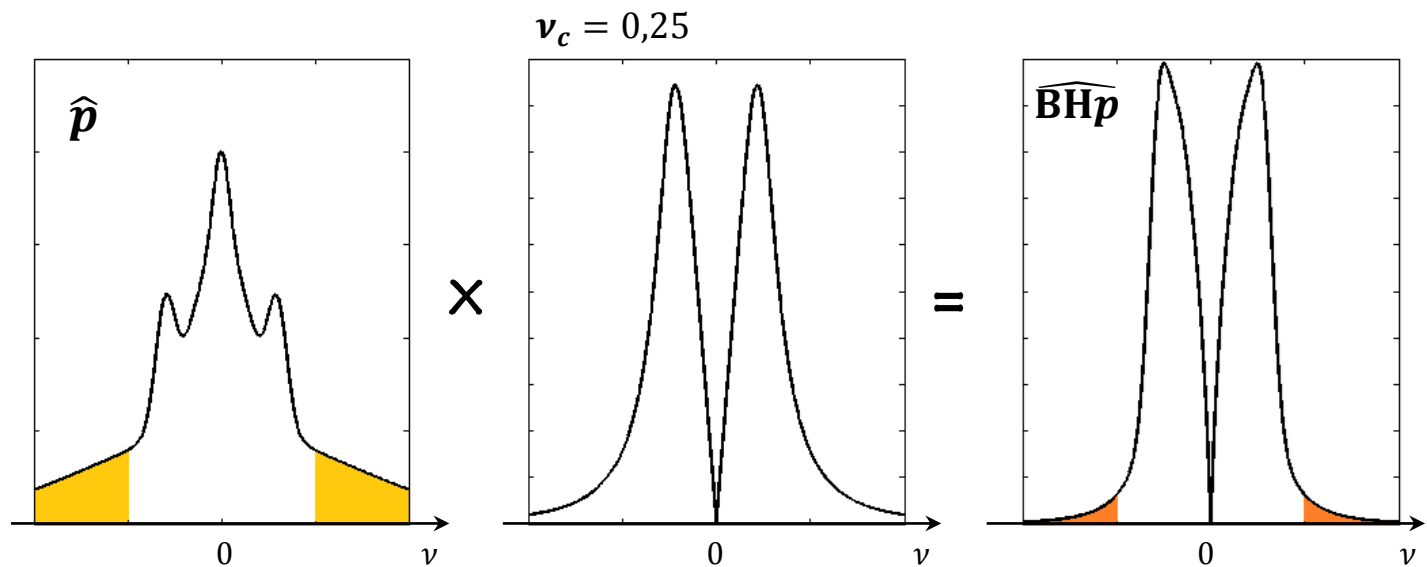
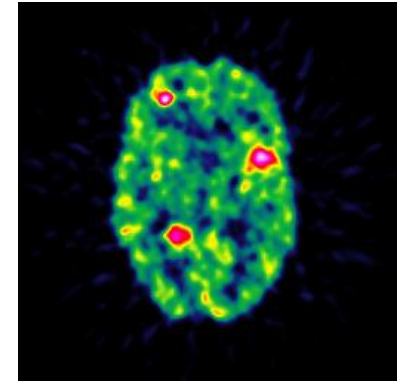
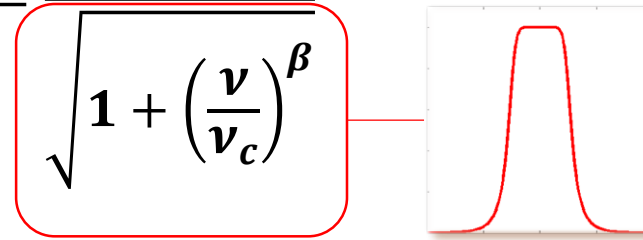
$$f = \mathbf{R}^* \mathbf{B} \mathbf{H} p \quad \widehat{\mathbf{B} \mathbf{H} p} = \frac{|\nu| \hat{p}}{\sqrt{1 + \left(\frac{\nu}{\nu_c}\right)^\beta}}$$




Régularisation

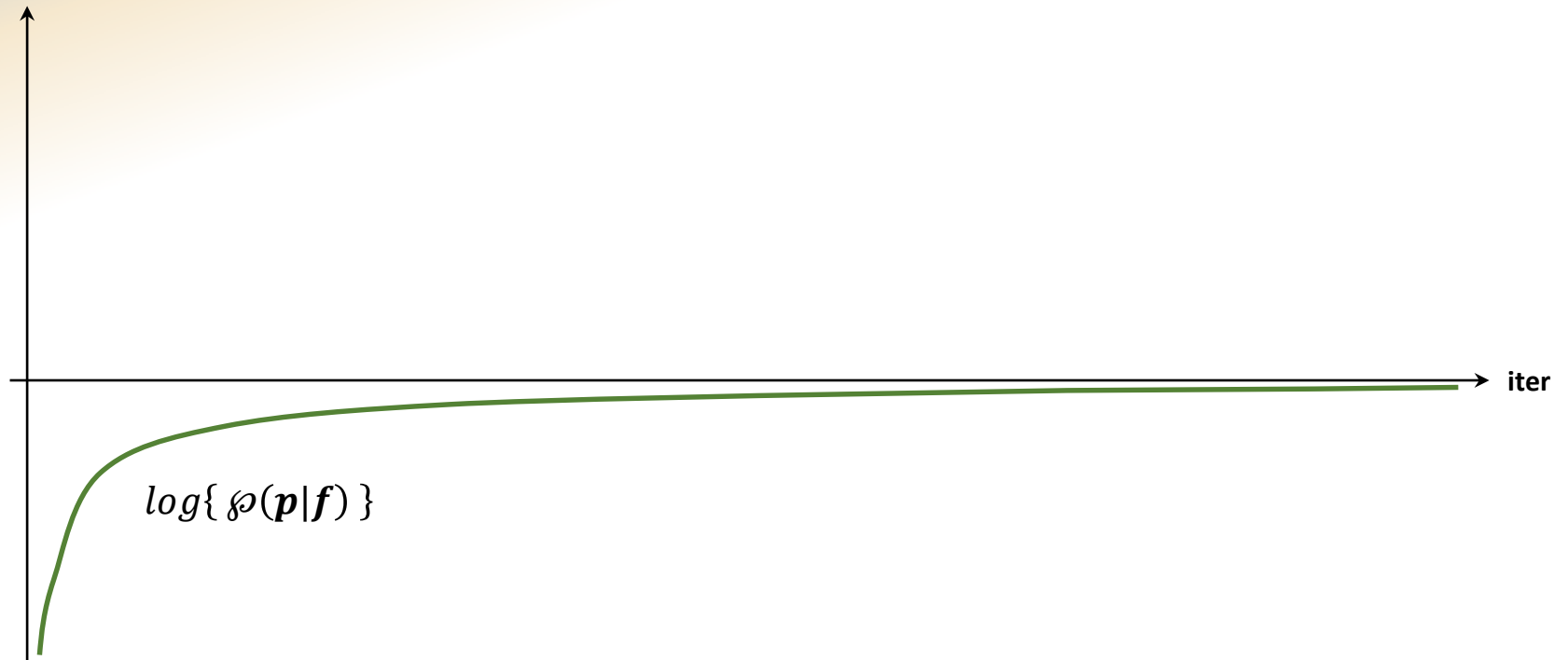
■ Reconstruction analytique

$$f = R^* \text{BH}p \quad \widehat{\text{BH}p} = \frac{|\nu| \hat{p}}{\sqrt{1 + \left(\frac{\nu}{\nu_c}\right)^\beta}}$$



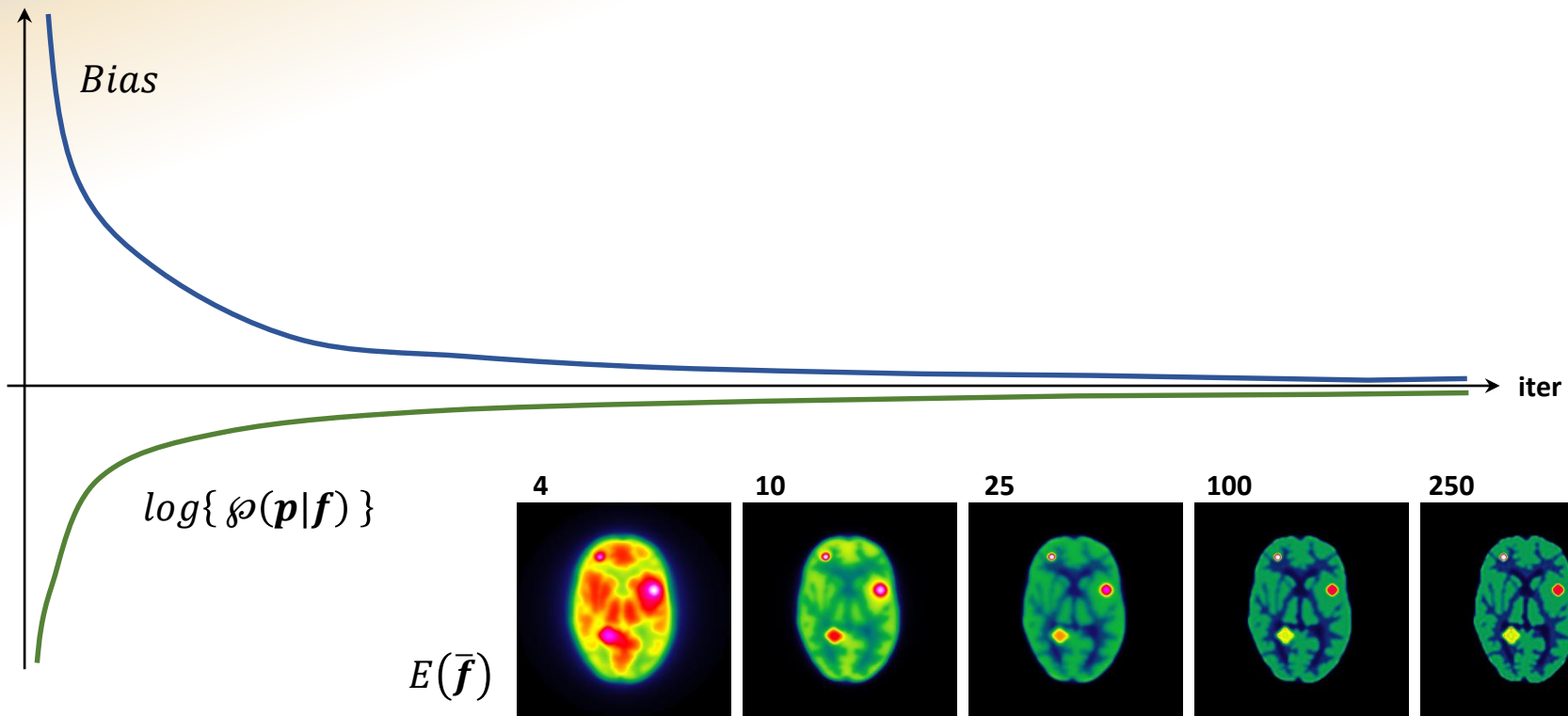
Régularisation

■ Reconstruction itérative



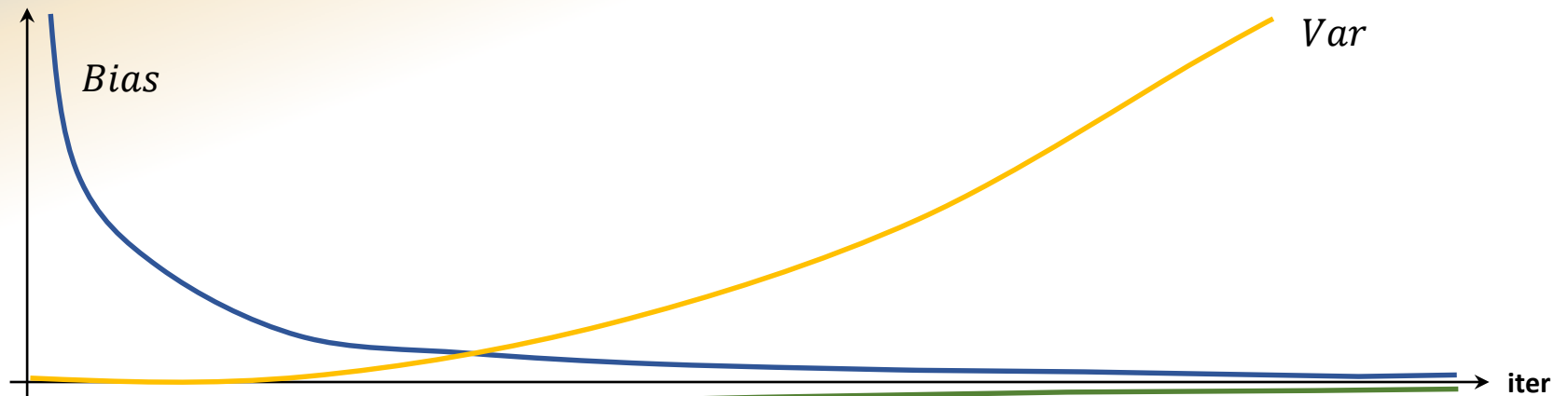
Régularisation

■ Reconstruction itérative



Régularisation

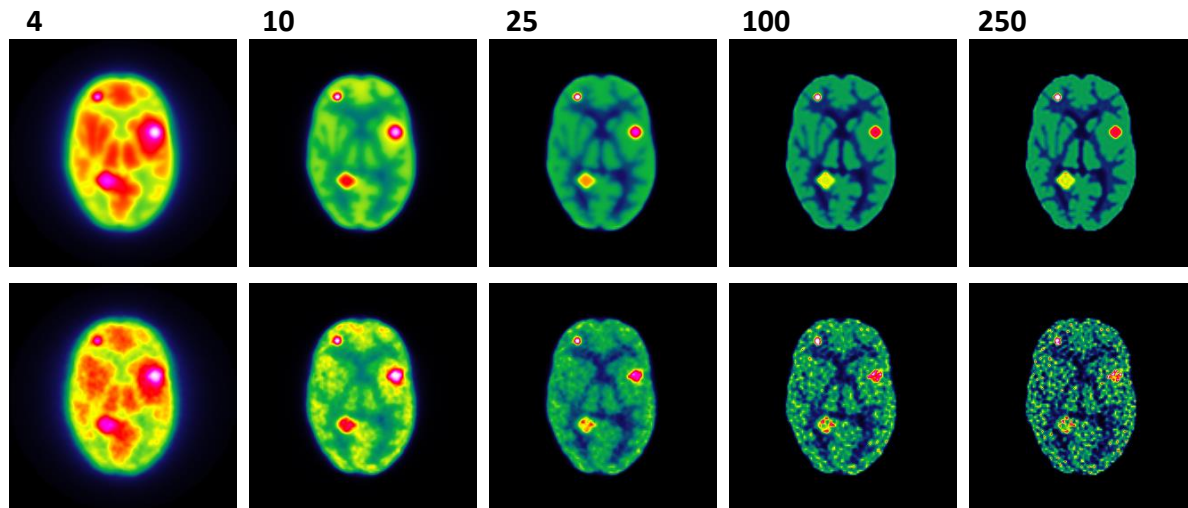
■ Reconstruction itérative



$\log\{\rho(p|f)\}$

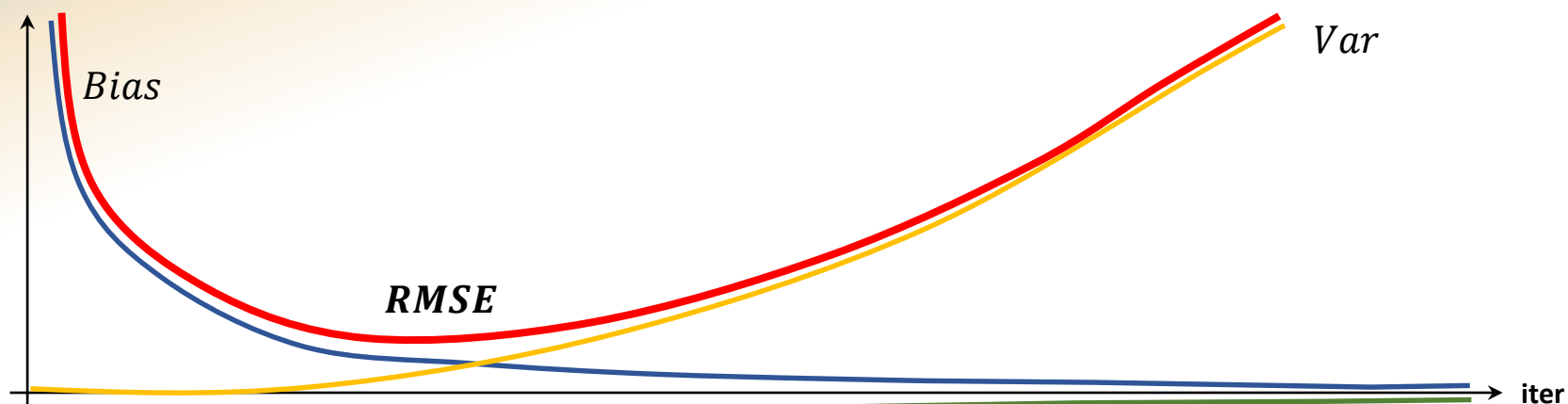
$E(\bar{f})$

\bar{f}



Régularisation

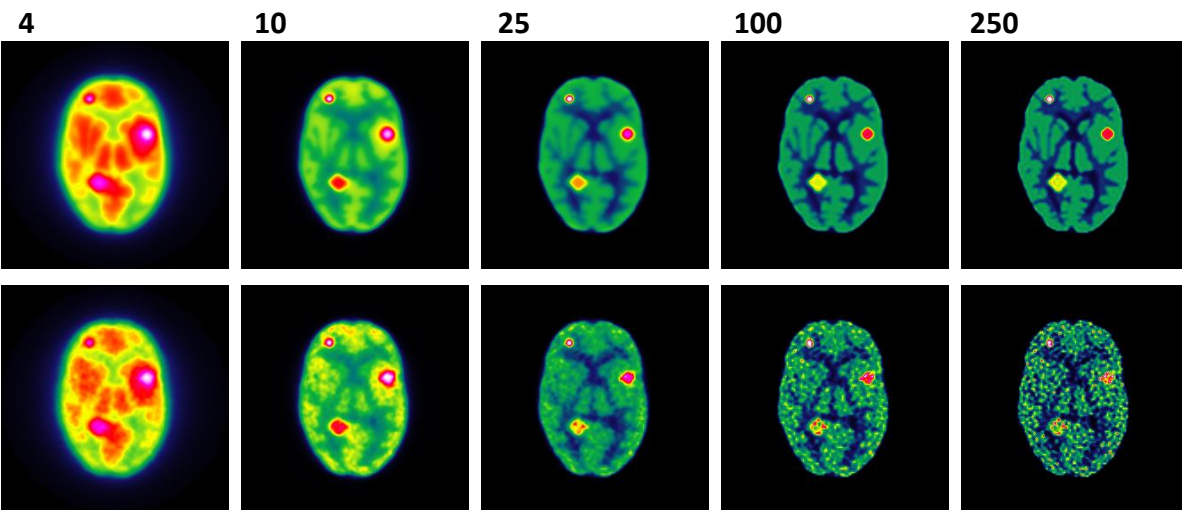
■ Reconstruction itérative



$\log\{\varphi(p|f)\}$

$E(\bar{f})$

\bar{f}

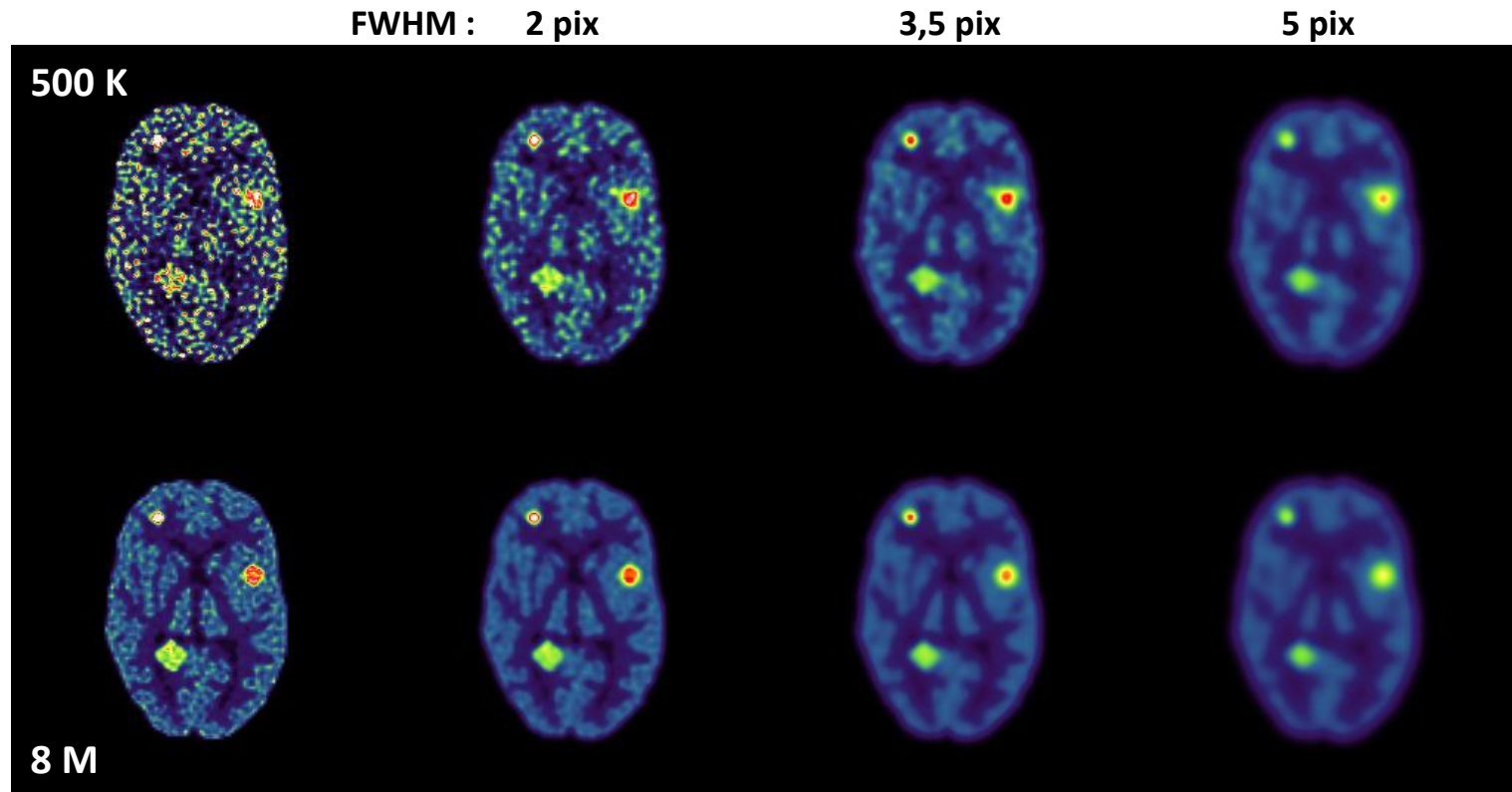


Régularisation

■ Reconstruction itérative

Post-filtrage

$$h * \bar{f}$$



Régularisation

■ Reconstruction itérative

Tikhonov

$$\bar{\mathbf{f}} = \underset{\mathbf{f} \in \mathcal{C}}{\operatorname{argmin}} \{J(\mathbf{f})\}$$

$$J(\mathbf{f}) = \|\mathbf{R}\mathbf{f} - \mathbf{p}\|^2 + \rho(\mathbf{f})$$

Régularisation

■ Reconstruction itérative

Tikhonov

$$\bar{f} = \underset{f \in C}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = \|Rf - p\|^2 + \rho(f)$$

Adéquation
Surjectivité
↓ biais

Régularisation

■ Reconstruction itérative

Tikhonov

$$\bar{f} = \underset{f \in \mathcal{C}}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = \|Rf - p\|^2 + \rho(f)$$

Adéquation
Surjectivité
↓ biais

Régularisation
Injectivité
↓ variance

Régularisation

■ Reconstruction itérative

Tikhonov

$$\bar{f} = \underset{f \in C}{\operatorname{argmin}} \{J(f)\}$$

$$J(f) = \|Rf - p\|^2 + \beta \|f\|^2$$

Régularisation

■ Reconstruction itérative

Tikhonov

$$\bar{\mathbf{f}} = \underset{\mathbf{f} \in \mathcal{C}}{\operatorname{argmin}} \{J(\mathbf{f})\}$$

$$J(\mathbf{f}) = \|\mathbf{R}\mathbf{f} - \mathbf{p}\|^2 + \beta\|\mathbf{f}\|^2$$

$$\bar{\mathbf{f}} = (\mathbf{R}^*\mathbf{R} + \beta\mathbf{I})^{-1} \mathbf{R}^* \mathbf{p}$$

$$\bar{\mathbf{f}}^{n+1} = (1 - \beta)\bar{\mathbf{f}}^n + \eta \mathbf{R}^* (\mathbf{p} - \mathbf{R} \bar{\mathbf{f}}^n)$$

Régularisation

■ Reconstruction itérative

MAP

$$\bar{f} = \underset{f \in \mathcal{C}}{\operatorname{argmin}} \{ J(f) \}$$

Posterior

$$J(f) = -\log \{ \wp(f|p) \}$$

Régularisation

■ Reconstruction itérative

MAP

$$\bar{f} = \underset{f \in \mathcal{C}}{\operatorname{argmin}} \{ J(f) \}$$

Posterior

$$J(f) = -\log \{ \wp(f|\mathbf{p}) \}$$

$$\wp(f|\mathbf{p}) = \frac{\wp(\mathbf{p}|f) \wp(f)}{\wp(\mathbf{p})}$$

Régularisation

■ Reconstruction itérative

MAP

$$\bar{f} = \underset{f \in \mathcal{C}}{\operatorname{argmin}} \{ J(f) \}$$

Posterior

$$J(f) = -\log \{ \wp(f|\mathbf{p}) \}$$

$$\wp(f|\mathbf{p}) = \frac{\wp(\mathbf{p}|f) \wp(f)}{\wp(\mathbf{p})}$$

$$J(f) = -\log \{ \wp(\mathbf{p}|f) \} - \log \{ \wp(f) \}$$

Régularisation

■ Reconstruction itérative

MAP

$$\bar{f} = \underset{f \in \mathcal{C}}{\operatorname{argmin}} \{ J(f) \}$$

Posterior

$$J(f) = -\log \{ \wp(f|p) \}$$

Likelihood

$$J(f) = -\log \{ \wp(p|f) \} - \log \{ \wp(f) \}$$

Adéquation
Surjectivité
↓ biais

Régularisation

■ Reconstruction itérative

MAP

$$\bar{f} = \underset{f \in \mathcal{C}}{\operatorname{argmin}} \{ J(f) \}$$

Posterior

$$J(f) = -\log\{ \wp(f|p) \}$$

Likelihood

Prior

$$J(f) = -\log\{ \wp(p|f) \} - \log\{ \wp(f) \}$$

Adéquation
Surjectivité
↓ biais

Régularisation
Injectivité
↓ variance

Régularisation

■ Reconstruction itérative

MAP

$$\bar{f} = \underset{f \in \mathcal{C}}{\operatorname{argmin}} \{ J(f) \}$$

Posterior

$$J(f) = -\log\{ \wp(f|p) \}$$

Likelihood

Prior

$$J(f) = -\log\{ \wp(p|f) \} - \log\{ \wp(f) \}$$

$$\wp(f) = \kappa e^{-\beta U}$$

Régularisation

■ Reconstruction itérative

MAP

$$\bar{f} = \underset{f \in \mathcal{C}}{\operatorname{argmin}} \{ J(f) \}$$

Posterior

$$J(f) = -\log \{ \wp(f|p) \}$$

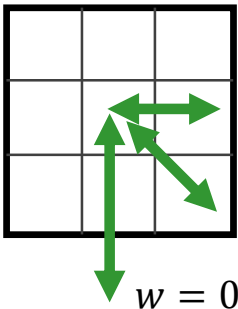
Likelihood

Prior

$$J(f) = -\log \{ \wp(p|f) \} - \log \{ \wp(f) \}$$

$$U = \sum_{i,j} w_{ij} \|f_i - f_j\|^2$$

$$\wp(f) = \kappa e^{-\beta U}$$



$$w_{1st} = 1$$

$$w_{2nd} = 1/\sqrt{2}$$

$$w = 0$$

Régularisation

■ Reconstruction itérative

MAP

$$\bar{f} = \underset{f \in \mathcal{C}}{\operatorname{argmin}} \{ J(f) \}$$

Posterior

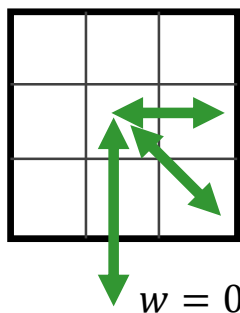
$$J(f) = -\log \{ \wp(f|p) \}$$

Likelihood

Prior

$$J(f) = -\log \{ \wp(p|f) \} - \log \{ \wp(f) \}$$

$$U = \sum_{i,j} w_{ij} \|f_i - f_j\|^2$$



$$w_{1st} = 1$$

$$w_{2nd} = 1/\sqrt{2}$$

$$w = 0$$

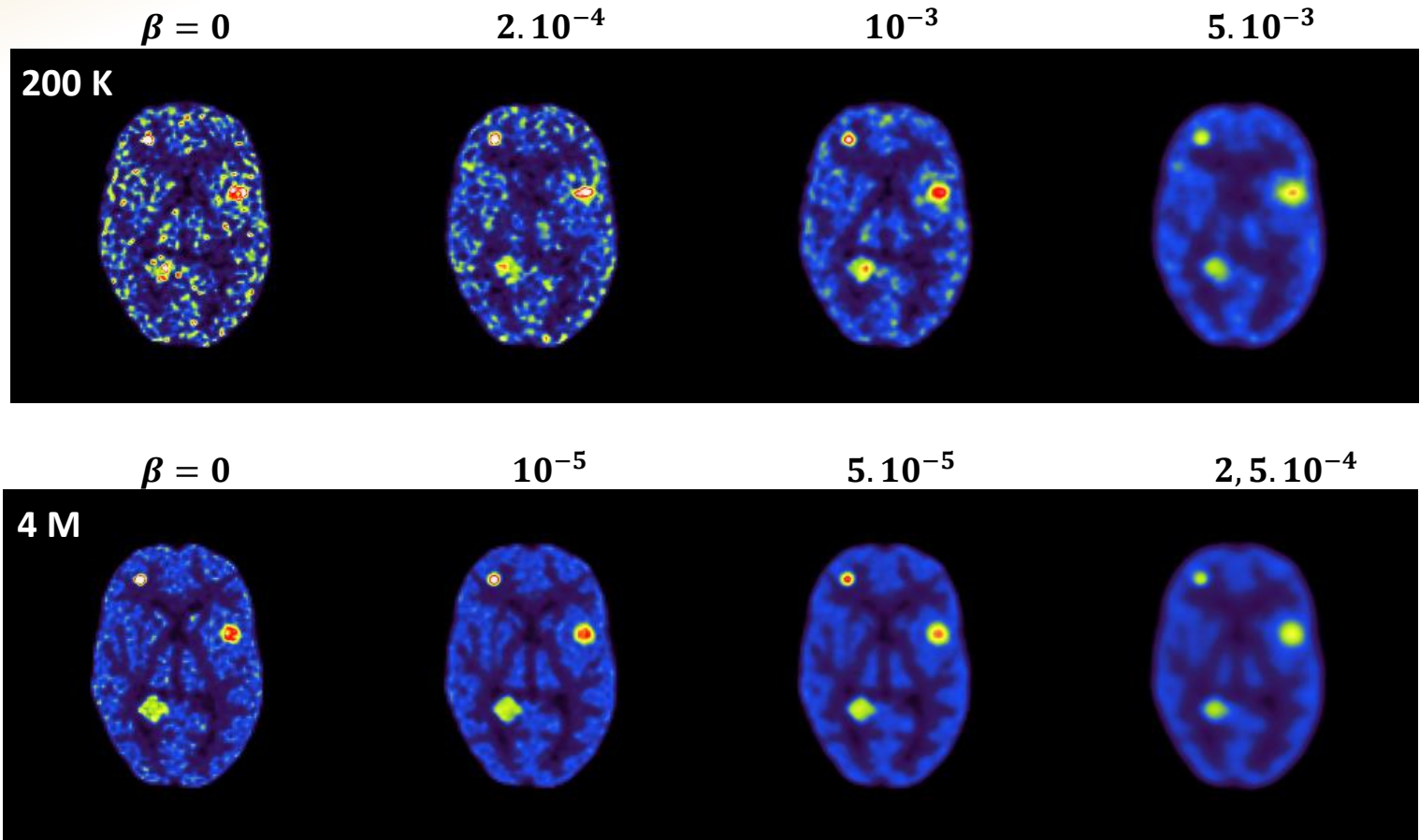
$$\wp(f) = \kappa e^{-\beta U}$$

$$\bar{f}^{n+1} = \frac{\bar{f}^n}{1 + \beta \nabla U} \times \left(\mathbf{R}^* \frac{p}{\mathbf{R} \bar{f}^n} \right)$$

Régularisation

■ Reconstruction itérative

MAP



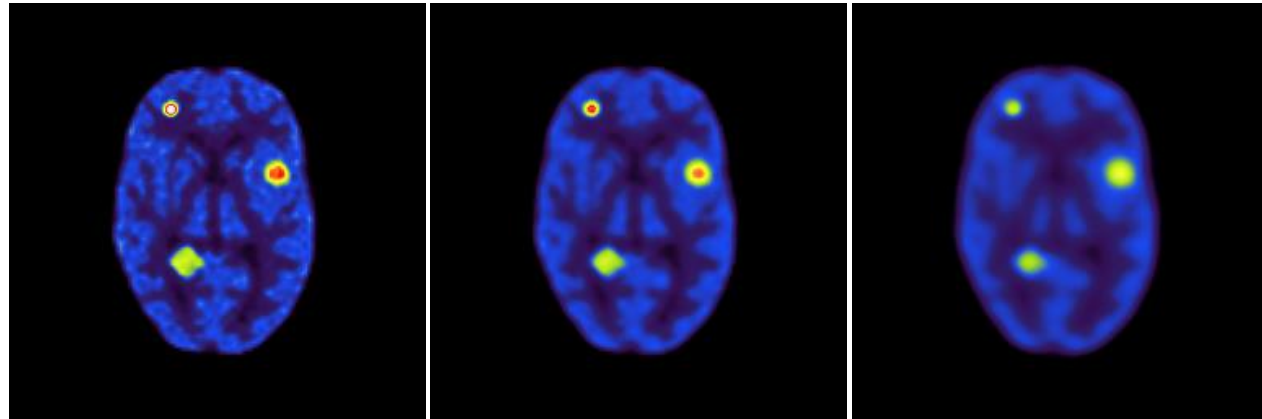
Régularisation

■ Reconstruction itérative

MAP

Quad. prior

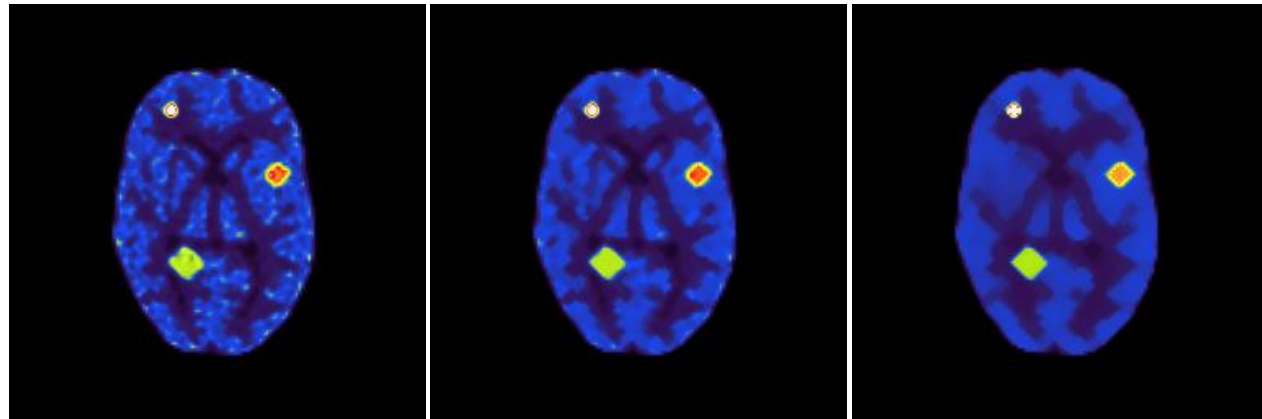
$$U = \sum_{i,j} w_{ij} \|f_i - f_j\|^2$$



Median prior

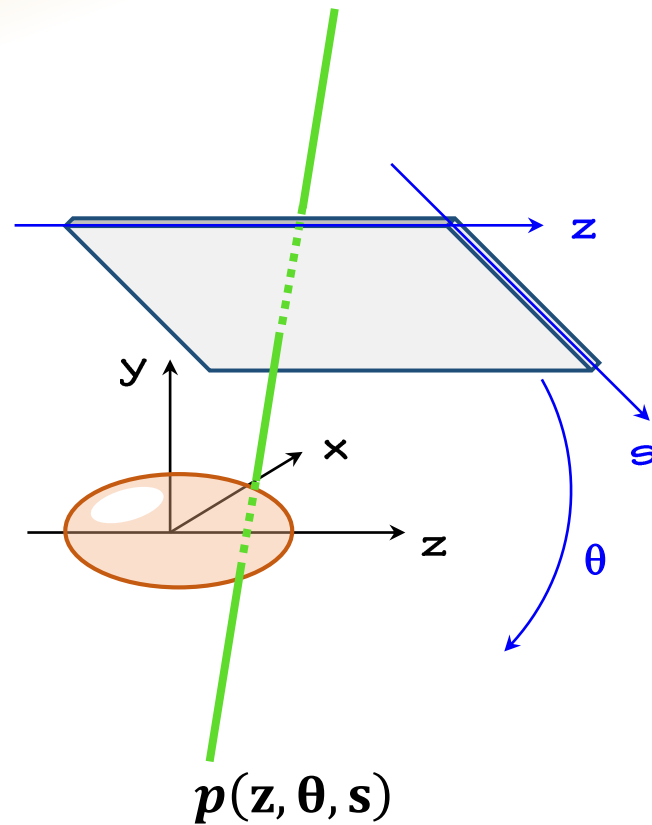
$$U = \sum_{i,j} w_{ij} |f_i - f_j|$$

« edge-preserving »



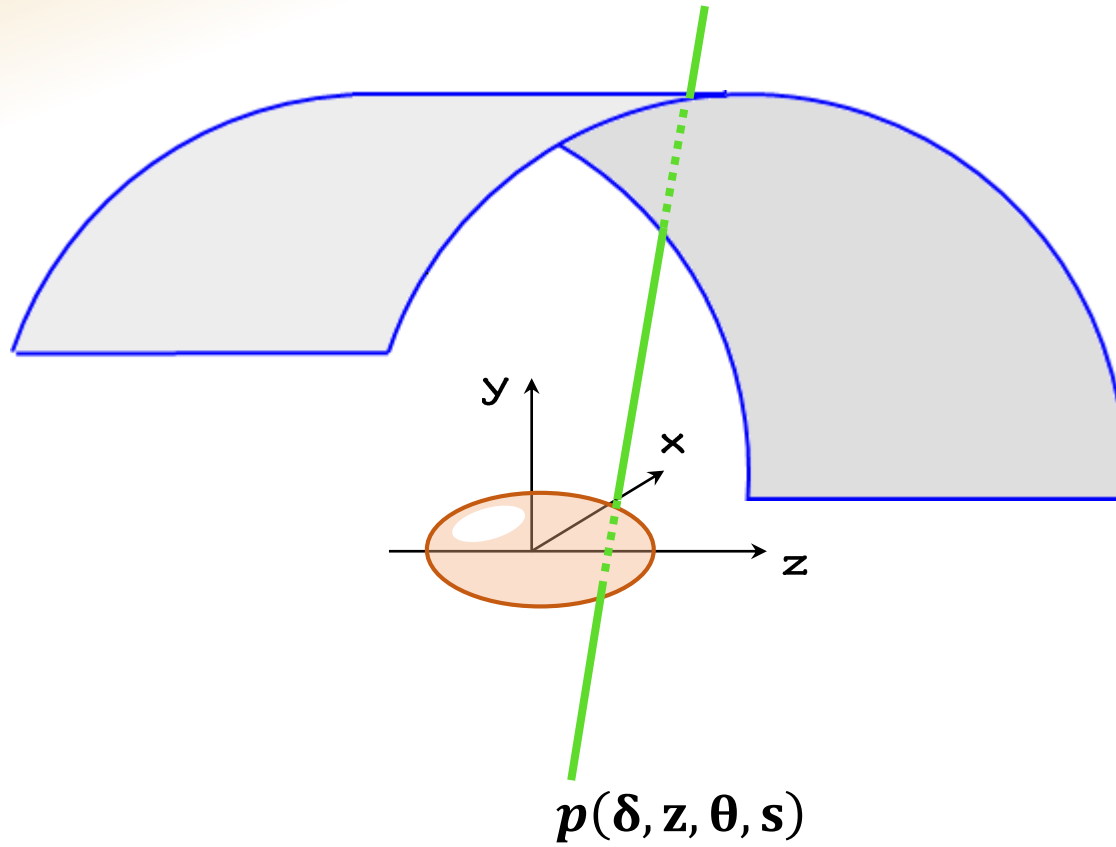
TEP 3D

■ Encodage TEMP



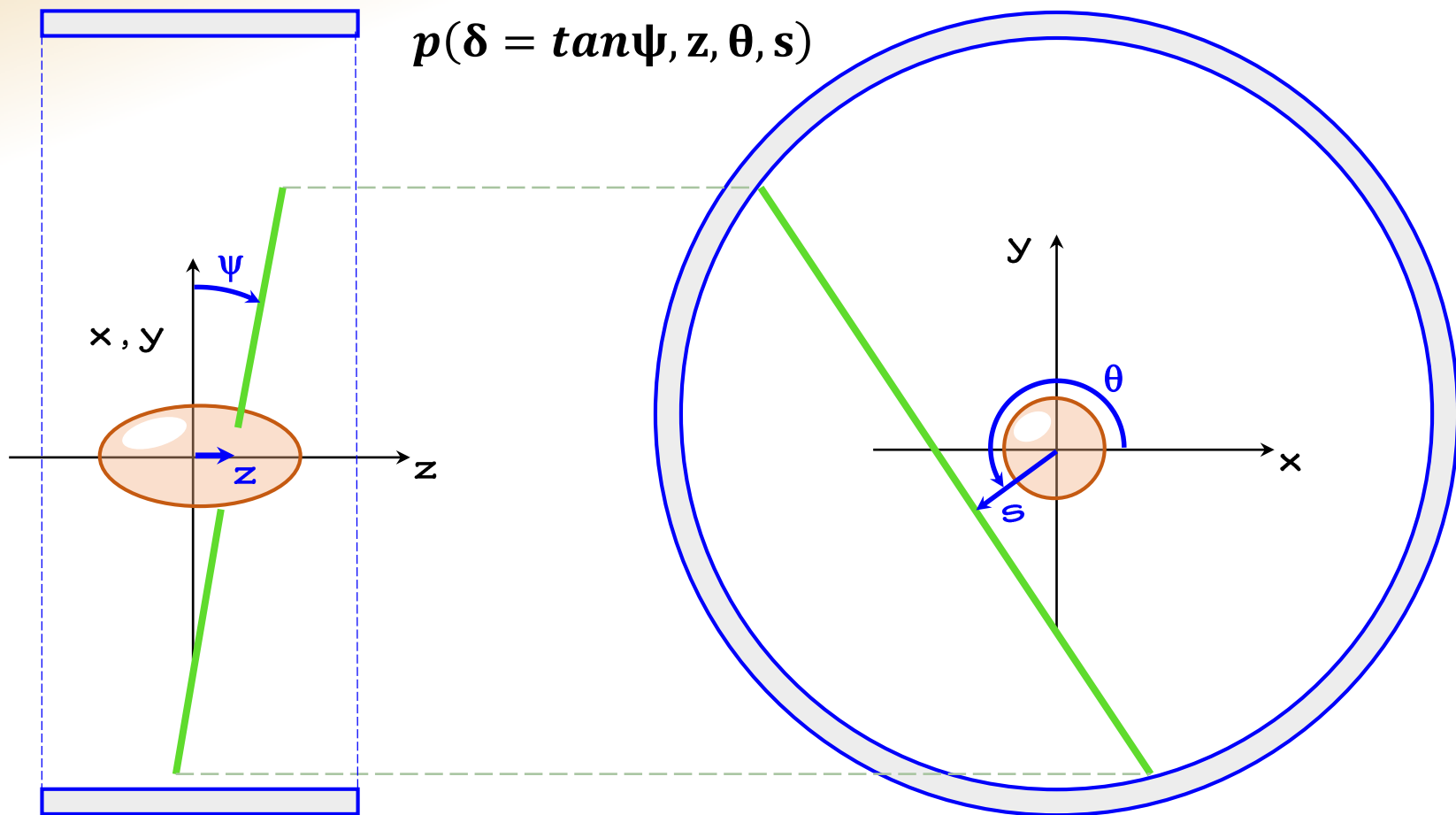
TEP 3D

■ Encodage TEP



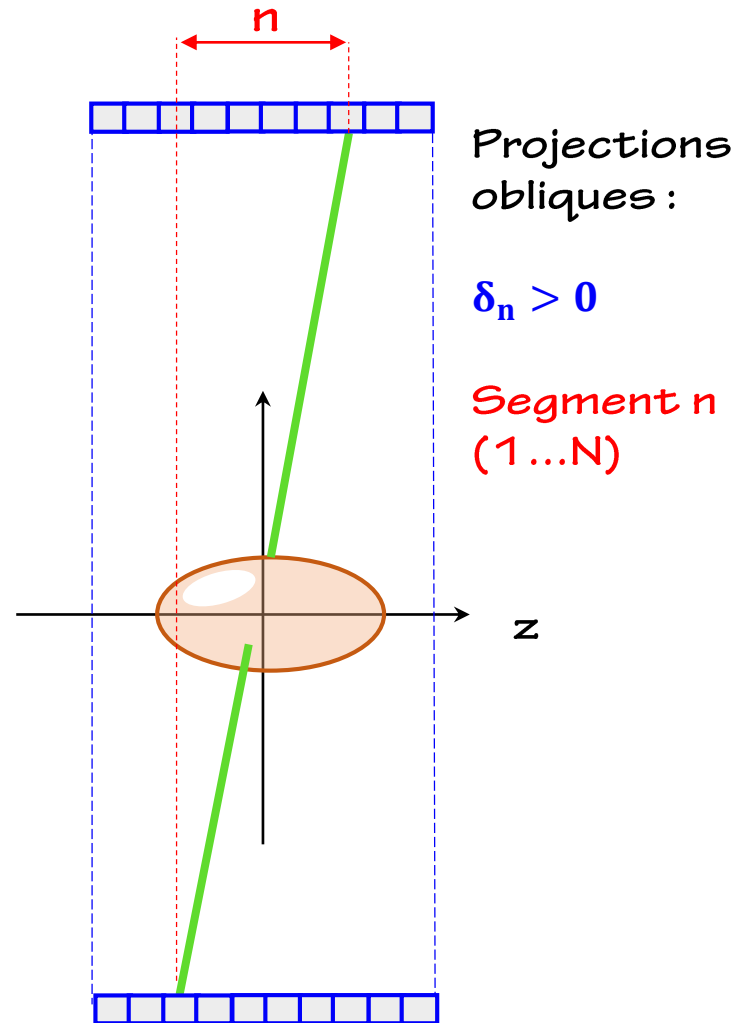
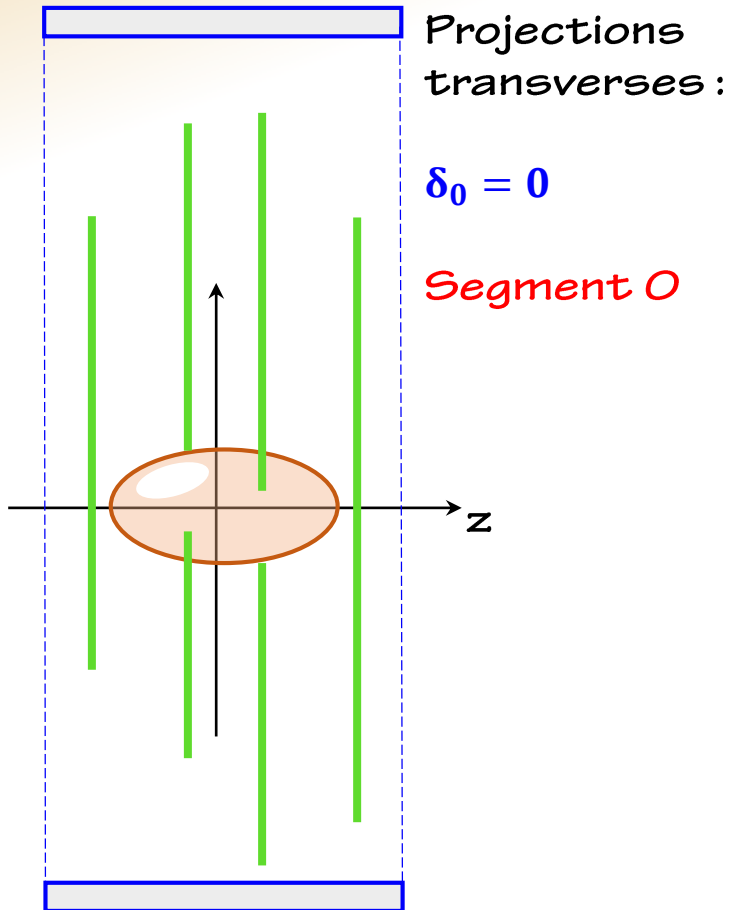
TEP 3D

■ Encodage TEP



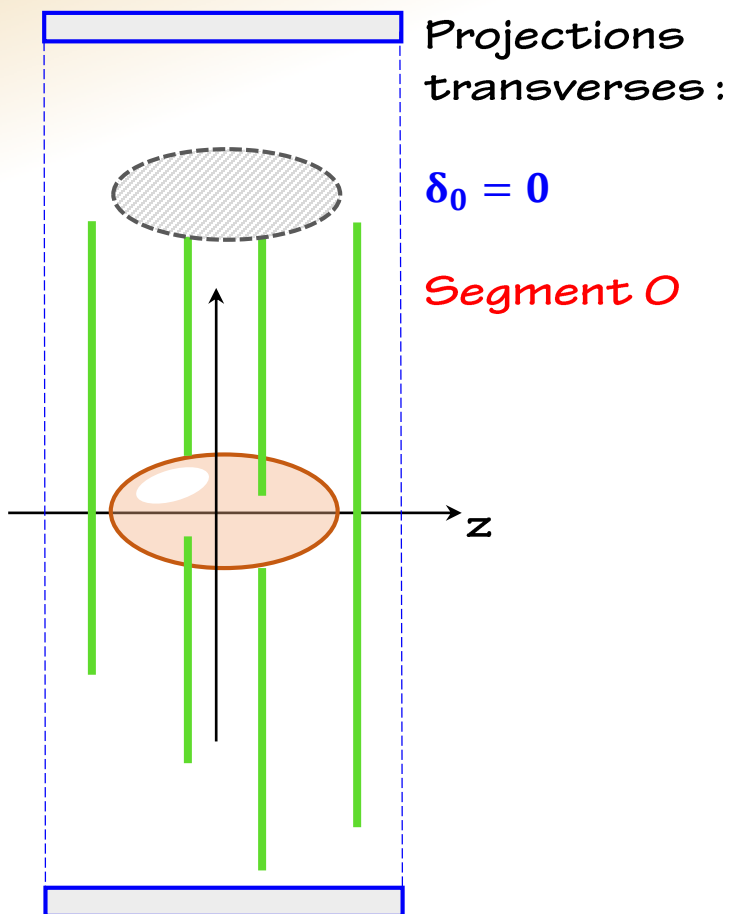
TEP 3D

■ Encodage TEP



TEP 3D

■ Encodage TEP



COMPLETES
&
SUFFISANTES

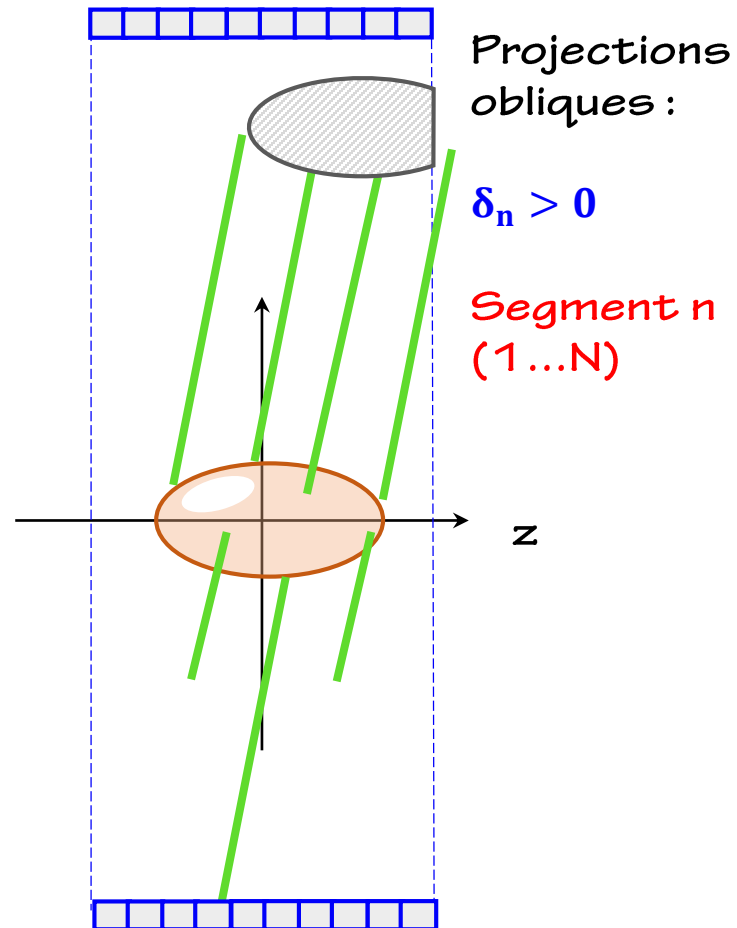
$$p(\mathbf{0}, z, \theta, s) \rightarrow f(x, y, z)$$

TEP 3D

■ Encodage TEP

TRONQUEES

$$p(\delta_{n>0}, z, \theta, s) \rightarrow f(x, y, z)$$



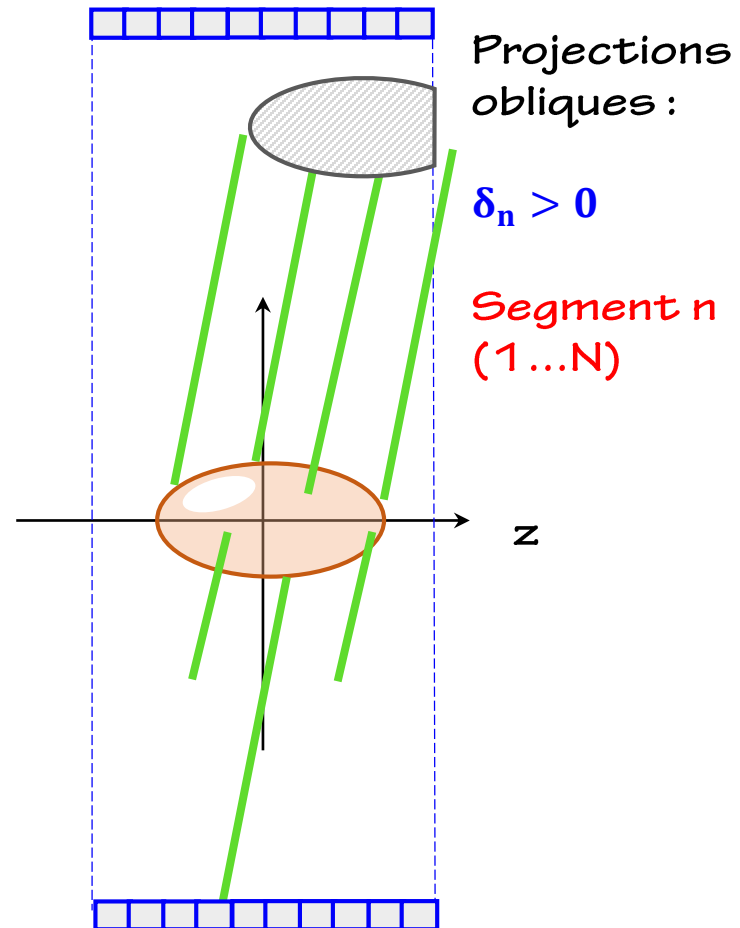
TEP 3D

■ Encodage TEP

TRONQUEES
&
REDONDANTES

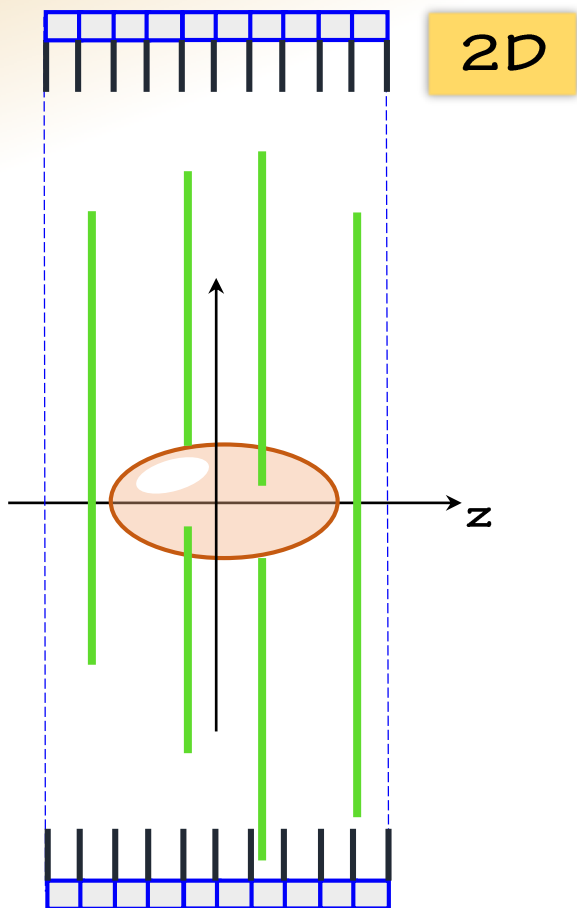
$$p(\delta_{n>0}, z, \theta, s) \rightarrow f(x, y, z)$$

$$\left. \begin{array}{l} p(0, z, \theta, s) \\ p(\delta_1, z, \theta, s) \\ p(\delta_2, z, \theta, s) \\ p(\delta_3, z, \theta, s) \\ \vdots \end{array} \right\} \rightarrow f(x, y, z)$$



TEP 3D

■ Reconstruction



$$p(0, z, \theta, s) \rightarrow f(x, y, z)$$

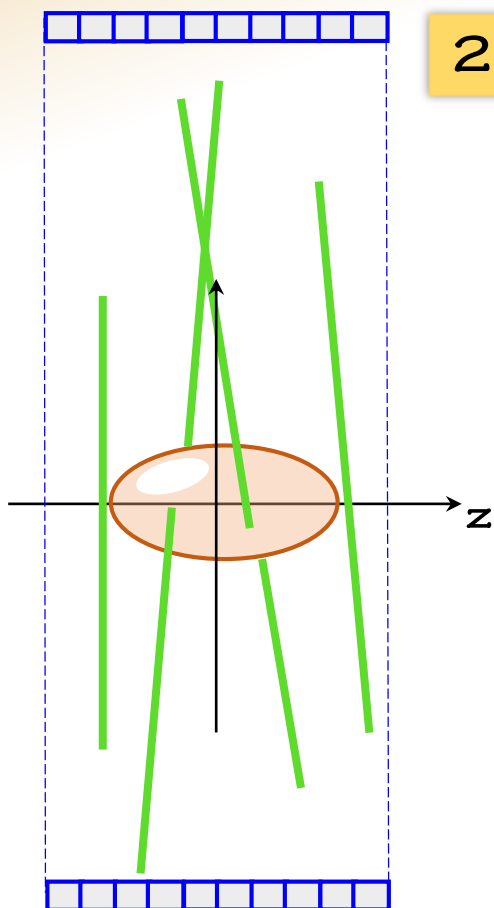
Collimation

Très rapide

SNR ↓

TEP 3D

■ Reconstruction



2D 1/2

$$\underbrace{p(0, z, \theta, s), p(\delta_1, z, \theta, s), p(\delta_2, z, \theta, s), \dots}_{\downarrow}$$
$$p(0, z, \theta, s)$$
$$\downarrow$$
$$f(x, y, z)$$

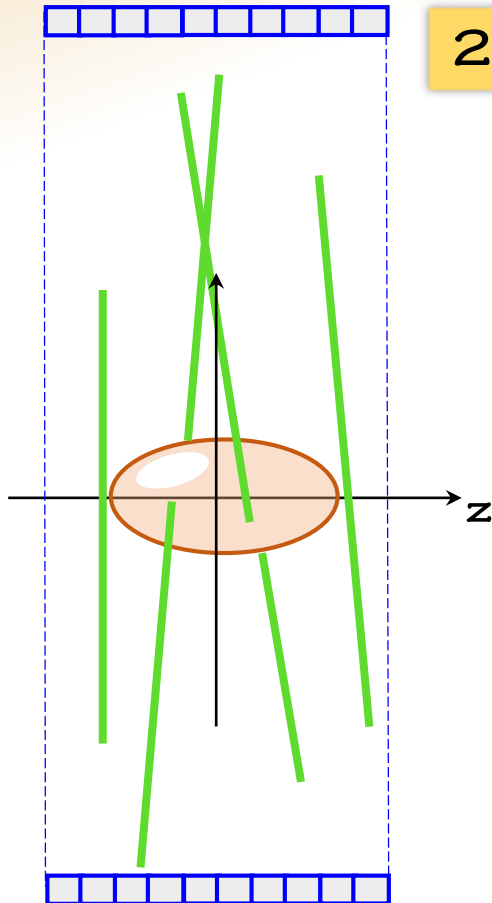
« Rebinning »

Rapide

Approximatif

TEP 3D

■ Reconstruction



2D 1/2

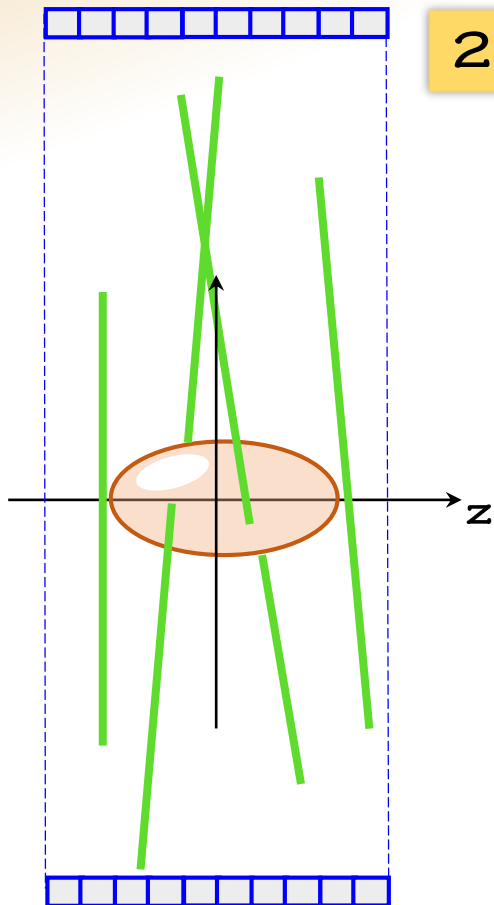
REBINNING EXACT

$$\hat{p}(0, \zeta, k, \omega) = e^{-i k a \tan\left(\frac{\delta \zeta}{\omega'}\right)} \hat{p}(\delta, \zeta, k, \omega')$$

$$\omega' = \sqrt{\omega^2 - \delta^2 \zeta^2}$$

TEP 3D

■ Reconstruction



2D 1/2

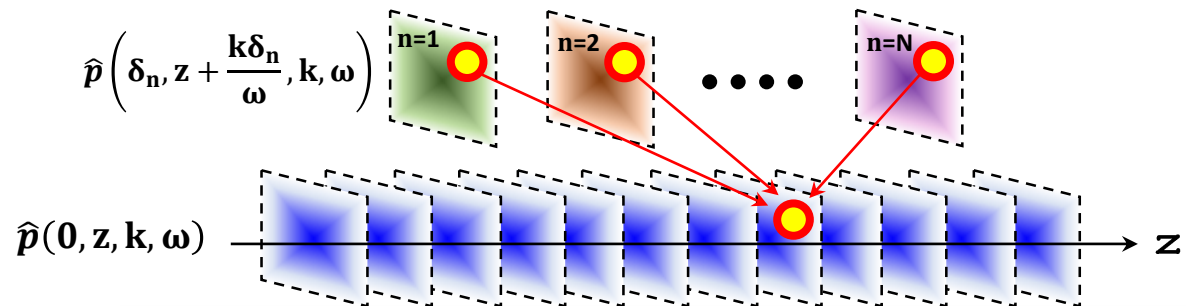
REBINNING EXACT

$$\hat{p}(0, \zeta, \mathbf{k}, \omega) = e^{-i \mathbf{k} \delta \tan(\frac{\delta \zeta}{\omega})} \hat{p}(\delta, \zeta, \mathbf{k}, \omega')$$

$$\omega' = \sqrt{\omega^2 - \delta^2 \zeta^2}$$

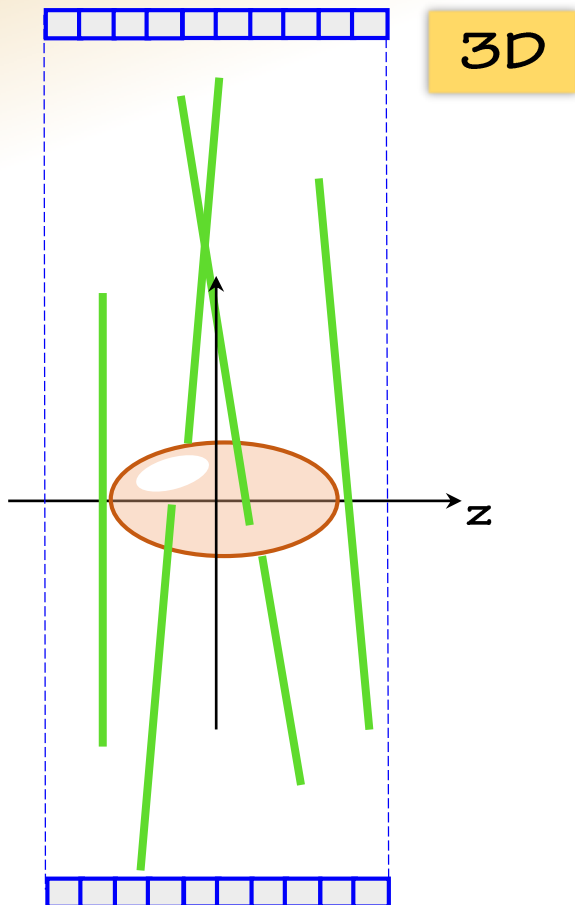
REBINNING APPROCHE

$$\hat{p}(0, z, \mathbf{k}, \omega) = \hat{p}\left(\delta, z + \frac{\mathbf{k} \delta}{\omega}, \mathbf{k}, \omega\right)$$



TEP 3D

■ Reconstruction



$$\left. \begin{array}{l} p(0, z, \theta, s) \\ p(\delta_1, z, \theta, s) \\ p(\delta_2, z, \theta, s) \\ p(\delta_3, z, \theta, s) \\ \vdots \end{array} \right\} \rightarrow f(x, y, z)$$

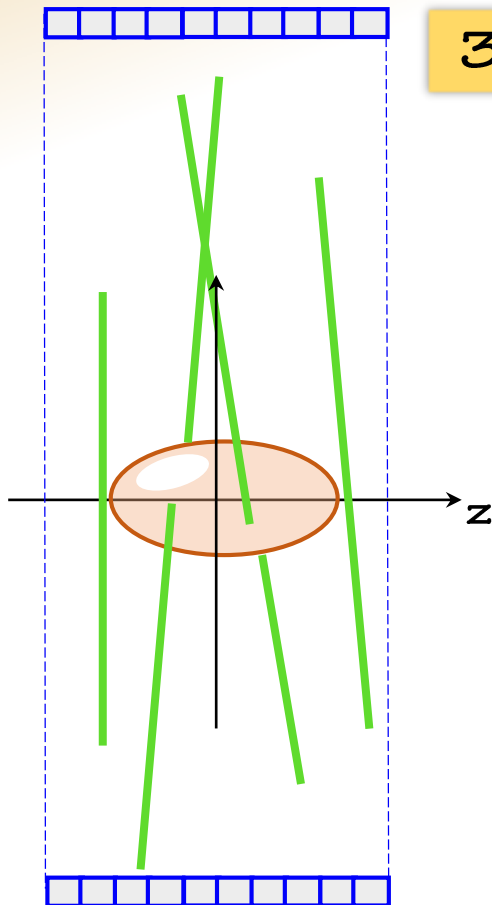
3D vrai

Reconstruction
analytique ou itérative

SNR \nearrow

TEP 3D

■ Reconstruction



3D

ALGEBRIQUE - ITERATIF

$$p(\delta, z, \theta, s) = \mathbf{R}_{3D} f(x, y, z)$$

$$\mathbf{R}_{3D} : \mathbb{R}^2 \times C_1 \times \{0 \dots N\} \rightarrow \mathbb{R}^3$$

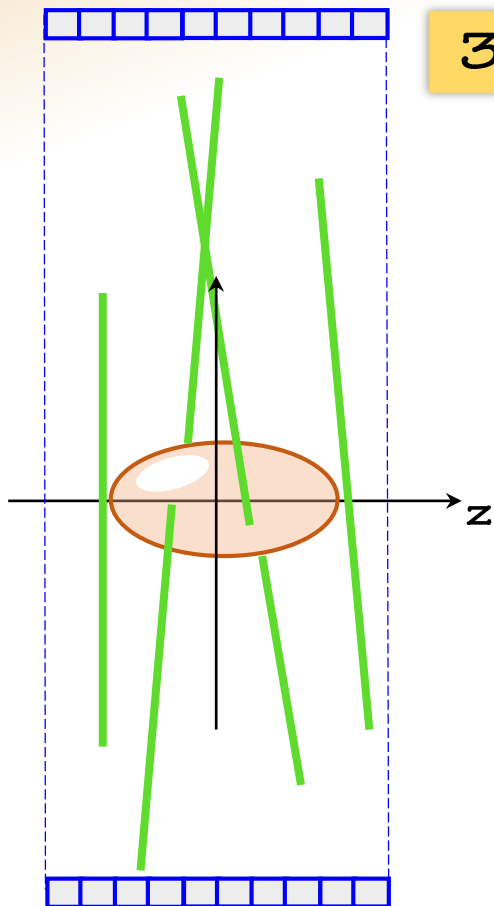
$$\text{Dim}(\mathbf{R}_{3D}) \gg$$

Stockage ?

Temps de calcul ...

TEP 3D

■ Reconstruction



3D

ANALYTIQUE : Fourier ou RPF

!! Complétion des données obliques !!

$$p(0, z, \theta, s) \rightarrow \tilde{f}(x, y, z)$$

$\downarrow R \tilde{f}$

$$\tilde{p}(\delta_{n>0}, z, \theta, s)$$

$$p(\delta_{n>0}, z, \theta, s)$$

Fusion

$$p(\delta_{n>0}, z, \theta, s)$$

$$p(0, z, \theta, s)$$

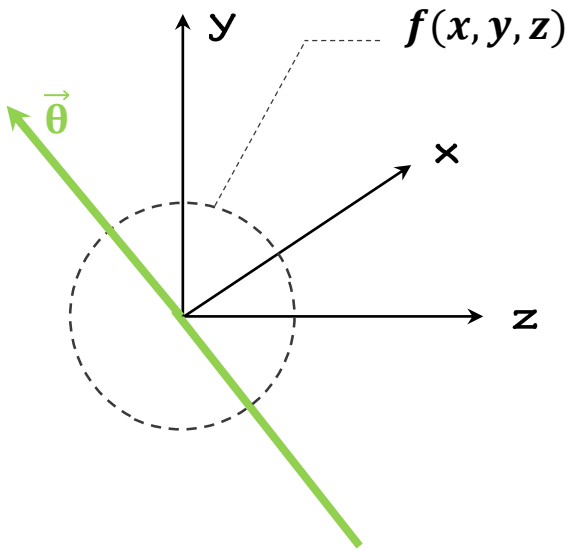
$$\rightarrow f(x, y, z)$$

TEP 3D

■ Reconstruction

3D

Synthèse de Fourier 3D

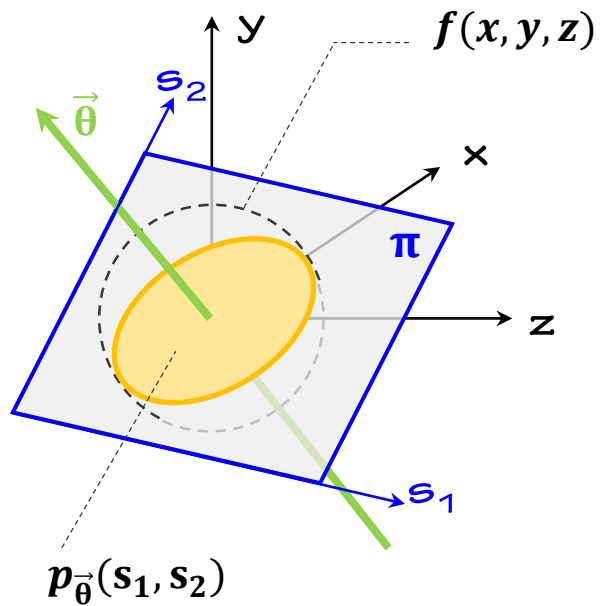


TEP 3D

■ Reconstruction

3D

Synthèse de Fourier 3D

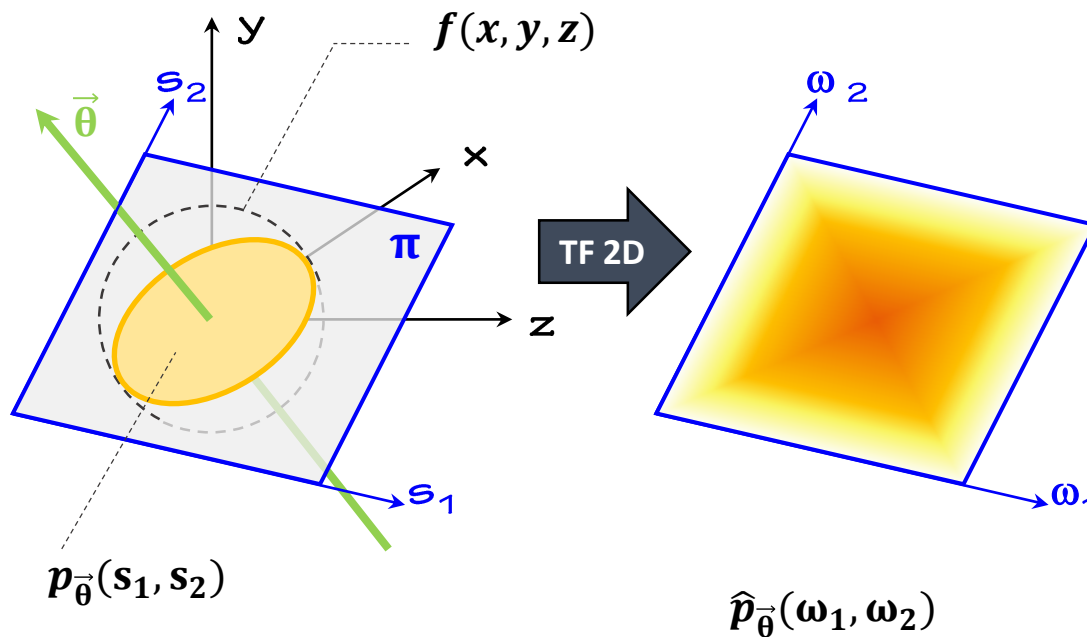


TEP 3D

■ Reconstruction

3D

Synthèse de Fourier 3D



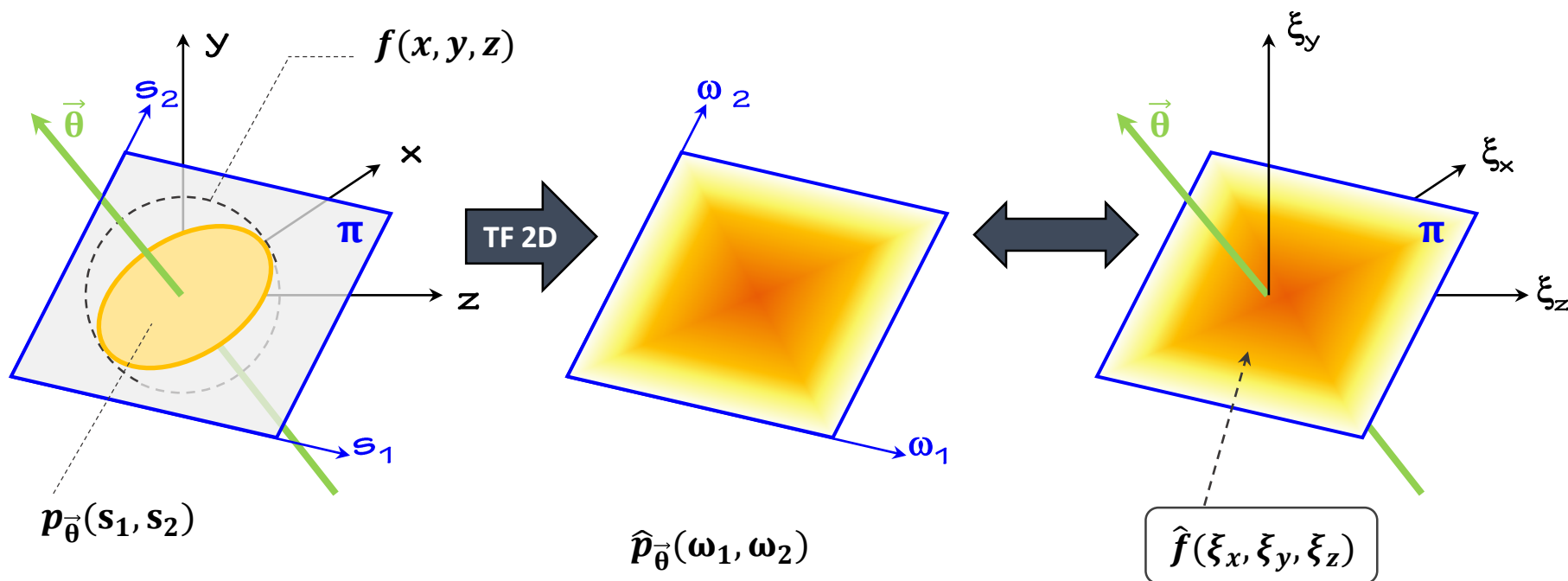
TEP 3D

■ Reconstruction

3D

Synthèse de Fourier 3D

$$\hat{f}([\xi_x, \xi_y, \xi_z] = R_{\vec{\theta}}[\omega_1, \omega_2]) = \hat{p}_{\vec{\theta}}(\omega_1, \omega_2)$$



TEP 3D

■ Reconstruction

3D

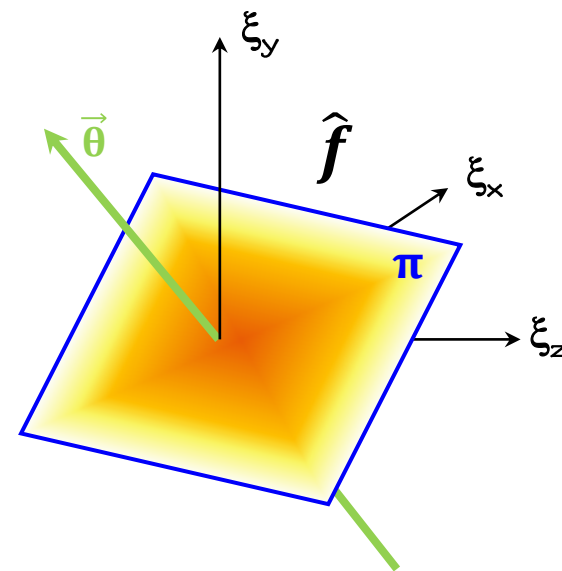
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Condition d'ORLOV

Nécessaire et suffisante pour que π décrive \mathbb{R}^3

$$\begin{aligned} \vec{\theta} &\in \Omega \subset S^2 \\ \exists C \in S^2 : C \subset \Omega \\ \forall C \in S^2 : \Omega \cap C &\neq \emptyset \end{aligned}$$



TEP 3D

■ Reconstruction

3D

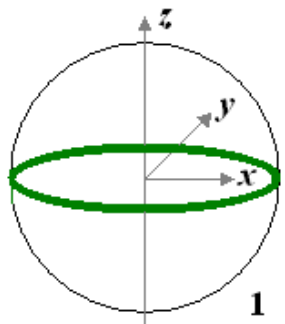
Synthèse de Fourier 3D

$$\hat{f}([\xi_x, \xi_y, \xi_z] = R_{\vec{\theta}}[\omega_1, \omega_2]) = \hat{p}_{\vec{\theta}}(\omega_1, \omega_2)$$

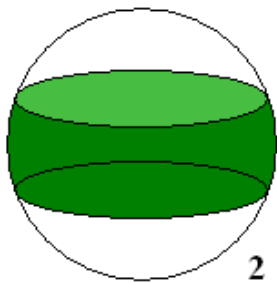
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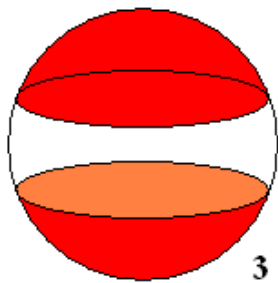
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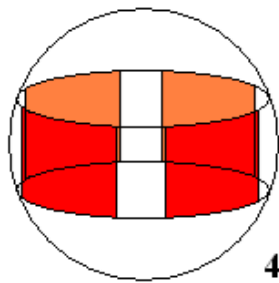
1



2



3



4

