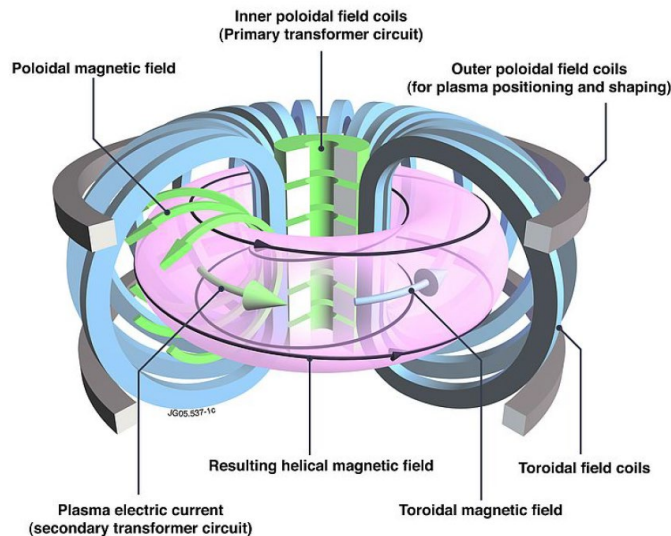


Tomography of the Fast-Ion Velocity Distribution in Tokamak Plasmas with a Physics-Based Prior

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Joint work with



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 and Martin S. Andersen, DTU Compute

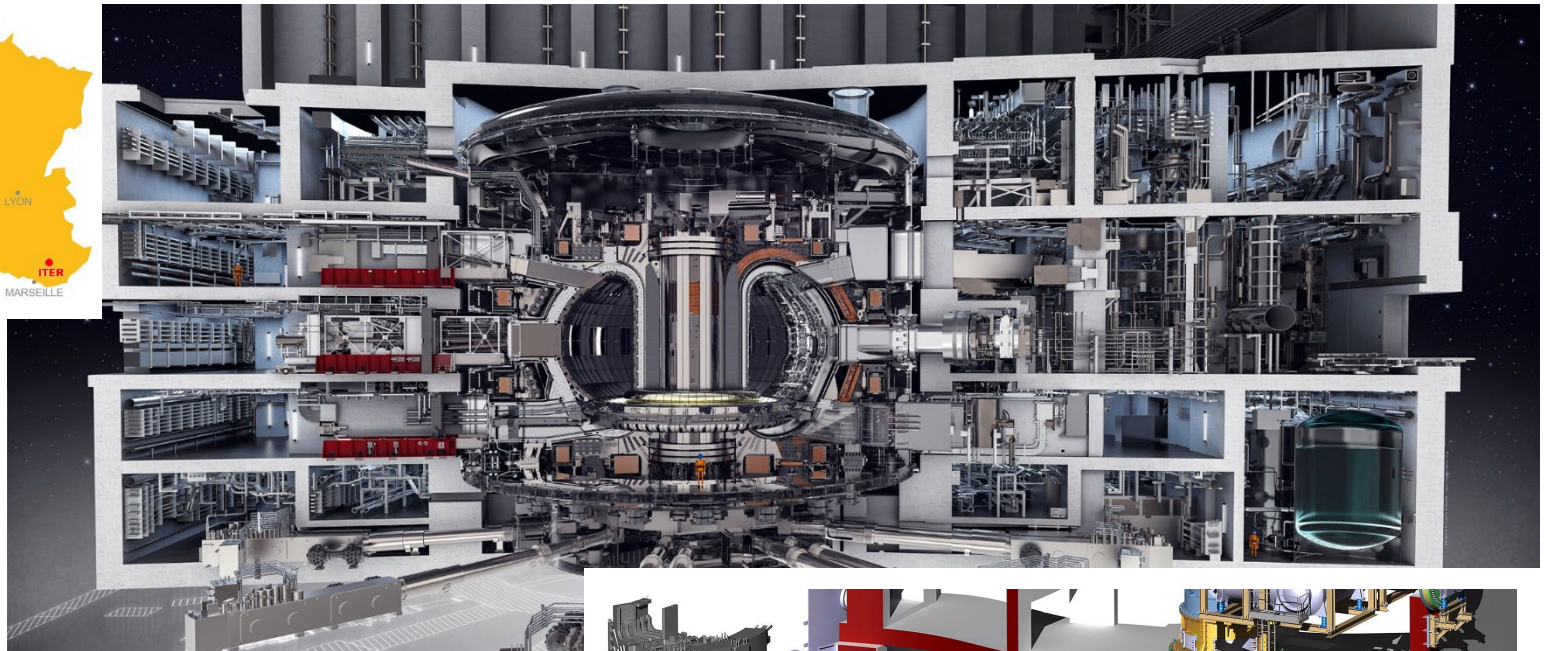
**Please don't ask me
 about the physics!**

DTU Compute

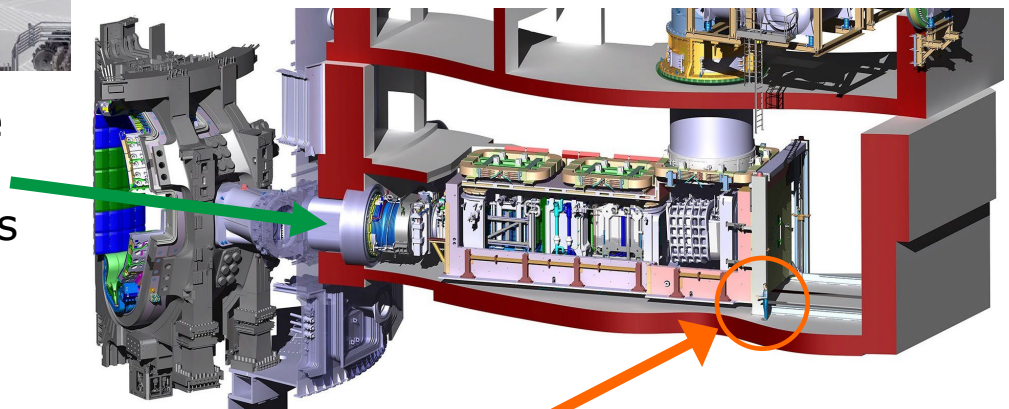
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The Tokamak Reactor

A **tokamak** uses a magnetic field to confine *plasma* in the shape of a torus. Energy produced through *fusion* is absorbed as heat → *electricity* by turbines.



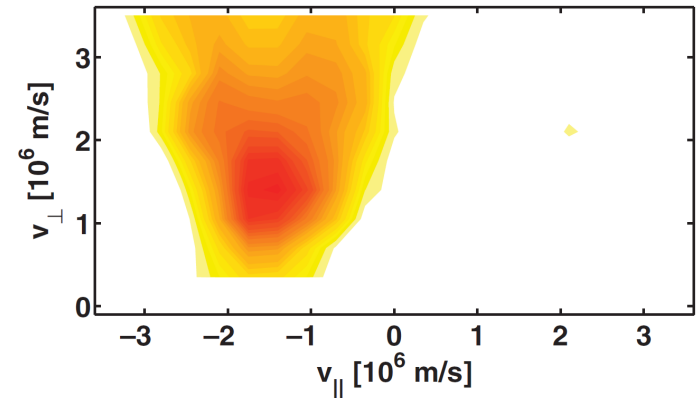
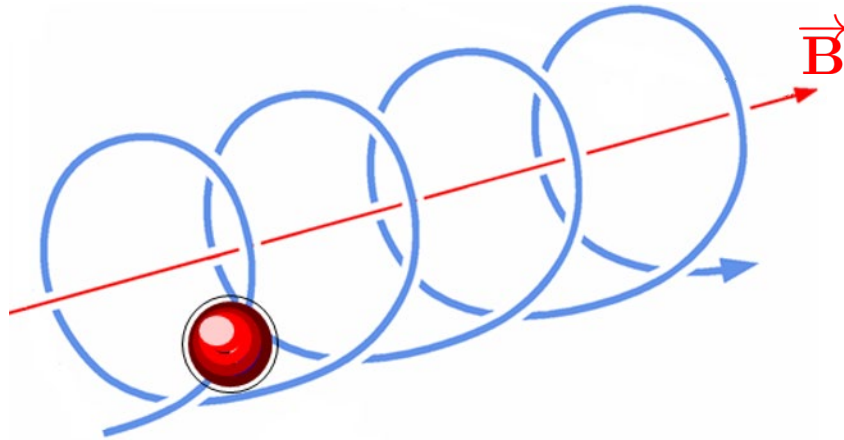
To reach fusion conditions the plasma is heated by *injecting neutral particles* (i.e., particles with no electric charge) into the plasma.



Scientist

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Distribution of Plasma Ion Velocities



Locally, the magnetic field is uniform and each *deuterium ion* \bullet in the plasma moves in a **helix** along the **magnetic field \vec{B}** :

- v_{\parallel} denotes the velocity along the field.
- v_{\perp} denotes the tangential speed (the radial speed is zero).

We want to know the probability distribution $f(v_{\parallel}, v_{\perp})$.

Towards the Inverse Problem

When **fast plasma ions** pass through the **neutral beam**, they undergo a reaction that creates a **newly-born fast neutral** and a **photon**.

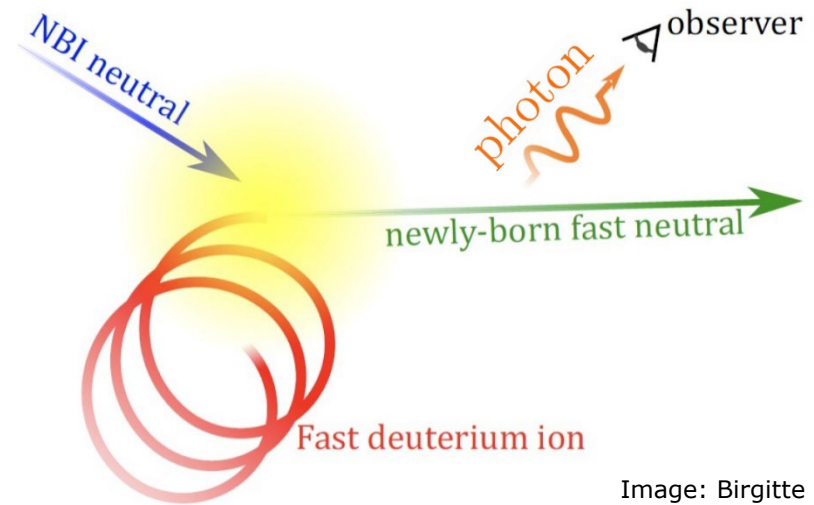


Image: Birgitte Madsen

The **photons** are *Doppler-shifted* depending on the velocity of the **fast ion**.

The inverse problem:

By measuring the Doppler-shift, we can infer about the *velocity of the fast ion*.

The Inverse Problem, Part I

On a spectral detector, we measure the *intensity per wavelength* $g(\lambda, \phi)$ of the photons, as a function of their wavelength λ .

This measurements also depend on the angle ϕ between the magnetic field \mathbf{B} and the line-of-sight \mathbf{u} to the photon detector.

The measurements are related to the velocity distribution $f(v_{\parallel}, v_{\perp})$ via the integral

$$g(\lambda, \phi) = \int_0^{\infty} \int_{-\infty}^{\infty} k(\lambda, \phi; v_{\parallel}, v_{\perp}) f(v_{\parallel}, v_{\perp}) dv_{\parallel} dv_{\perp} ,$$

in which the kernel is given by

$$k(\lambda, \phi; v_{\parallel}, v_{\perp}) = R \cdot \pi(\lambda | \phi, v_{\parallel}, v_{\perp})$$

$$\pi(\lambda | \phi, v_{\parallel}, v_{\perp}) = \text{probability density function for } \lambda$$

and R is a constant (for simplicity).

The Inverse Problem, Part II

Recall there is a Doppler shift, which we can write

$$\lambda - \lambda_D = \frac{u \lambda_D}{c} \quad \Leftrightarrow \quad u = c \frac{\lambda - \lambda_D}{\lambda_D}$$

where $\lambda_D = 656.1$ nm, $c =$ speed of light, and $u =$ the ion's velocity component along the line-of-sight \mathbf{u} to the detector. We now switch to the formulation

$$g(u, \phi) = \int_0^\infty \int_{-\infty}^\infty k(u, \phi; v_{\parallel}, v_{\perp}) f(v_{\parallel}, v_{\perp}) dv_{\parallel} dv_{\perp} ,$$

with

$$k(u, \phi; v_{\parallel}, v_{\perp}) = \frac{R}{\pi v_{\perp} \sin \phi} \left(1 - \left(\frac{u - v_{\parallel} \cos \phi}{v_{\perp} \sin \phi} \right)^2 \right)^{-1/2} .$$

When the argument to $()^{-1/2}$ is negative, we set $k = 0$.

We use quadrature discretization to obtain a linear algebraic system $Ax = b$.

Other Variables: E and p

Instead of working with the velocity distribution in the form $f(v_{\parallel}, v_{\perp})$, the physicists often prefer a *transformation of variables* to **energy E** and **pitch p** :

$$E = 1/2 m_{\text{D}} v^2 , \quad p = v_{\parallel} / v$$

$$v = \sqrt{v_{\parallel}^2 + v_{\perp}^2} , \quad m_{\text{D}} = \text{mass of deuterium ion}$$

and instead work with the distribution function $F(E, p)$:

$$I(\mathbf{u}, \phi) = \int_{-1}^1 \int_0^{\infty} K(\mathbf{u}, \phi; E, p) F(E, p) dE dp .$$

From (E, p) to $(v_{\parallel}, v_{\perp})$:

$$v_{\perp} = p \sqrt{\frac{2E}{m_{\text{D}}}} , \quad v_{\parallel} = \sqrt{\frac{2E}{m_{\text{D}}}} (1 - p^2) .$$

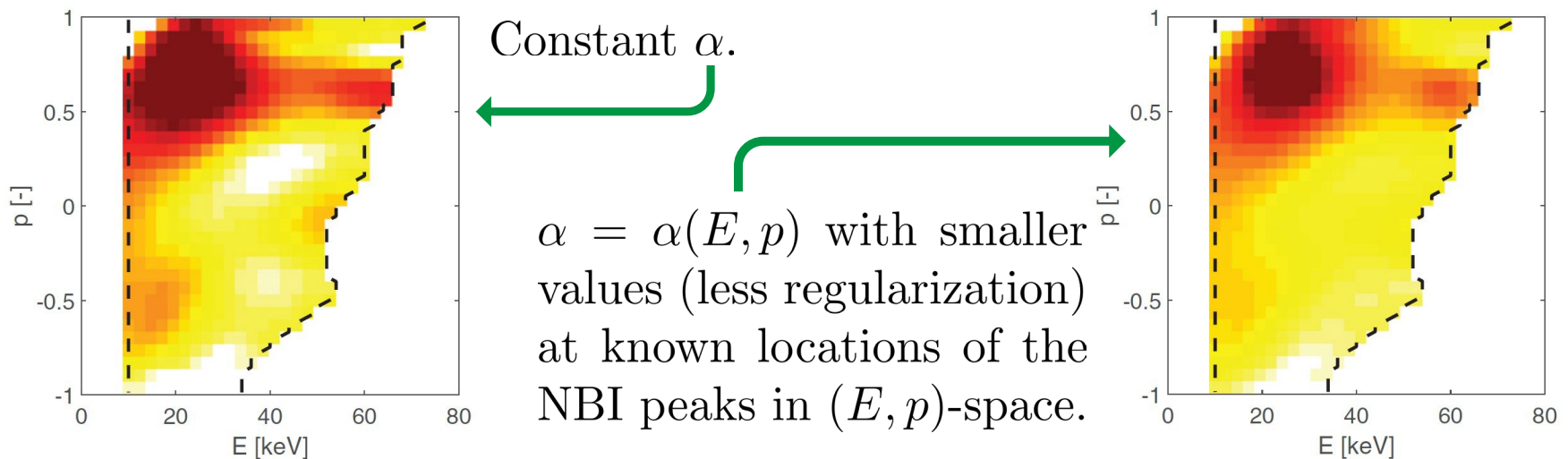
Utilizing Knowledge about the Physics

Tikhonov regularization in general form, with constraint:

$$\min_x \|Ax - b\|_2 + \alpha^2 \|Lx\|_2^2 \quad \text{s.t.} \quad x \geq 0,$$

where $\alpha = \alpha(E, p)$ may depend on energy and pitch; Salewski et al. (2016).

Prior	Benefit	Risk
$x \geq 0$	Improves solution	$x < 0$ could diagnose data error
$L \sim 1.$ deriv	Gives smooth solutions	Misses spikes and ridges
$\alpha(E, p)$	Accounts for NBI peaks	Might introduce spurious peaks.



Physics Prior via Slowing-Down Functions

Instead of working with a standard “pixel basis” for $F(E, p)$, we can use suited expansion functions $\psi_1, \psi_2, \dots, \psi_{n_{sd}}$ and write the reconstruction

$$F(E, p) = \sum_{j=1}^{n_{sd}} c_j \psi_j(E, p) .$$

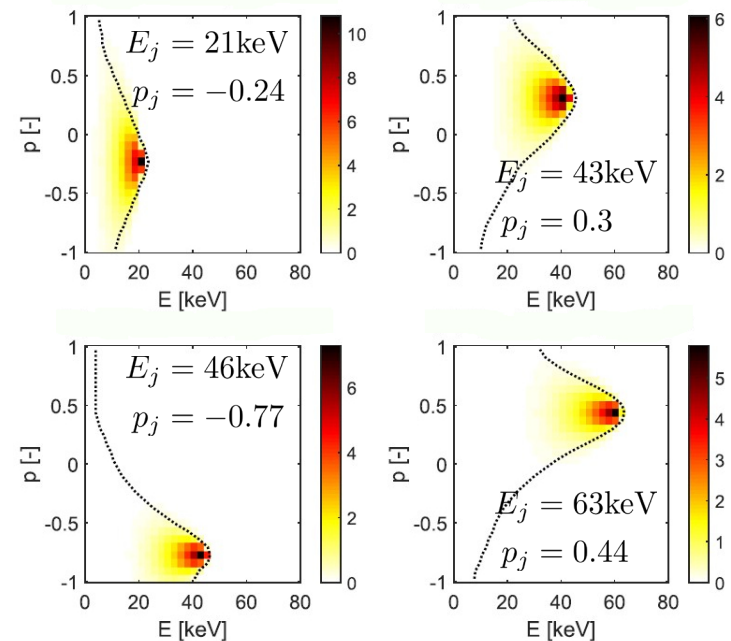
The discretized and regularized problem for $\{c_j\}$ then takes the basic form

$$\min_c \|A \Psi c - b\|_2^2 + \alpha^2 \|c\|_2^2 , \quad \Psi \in \mathbb{R}^{n \times n_{sd}} , \quad x = \Psi c ,$$

where the columns of Ψ are samples of the functions $\psi_j =$ **slowing down functions**.

Each ψ_j is a distribution F excited by a δ -function $\delta(u_j, \phi_j)$ corresponding to (E_j, p_j) , revealing how the plasma ions “**slow down**” due to collisions.

Hence, the basis functions ψ_j represent the physics of the plasma.



Interpretation of Basis with Physics Prior

The case $n_{sd} = n \rightarrow$ as many slowing down functions as pixels (a basis)

Ψ is square and we assume it has full rank. Recall $x = \Psi c \Leftrightarrow c = \Psi^{-1}x$:

$$\min_c \|A \Psi c - b\|_2^2 + \alpha^2 \|c\|_2^2 \quad \Leftrightarrow \quad \min_x \|A x - b\|_2^2 + \alpha^2 \|\Psi^{-1}x\|_2^2 .$$

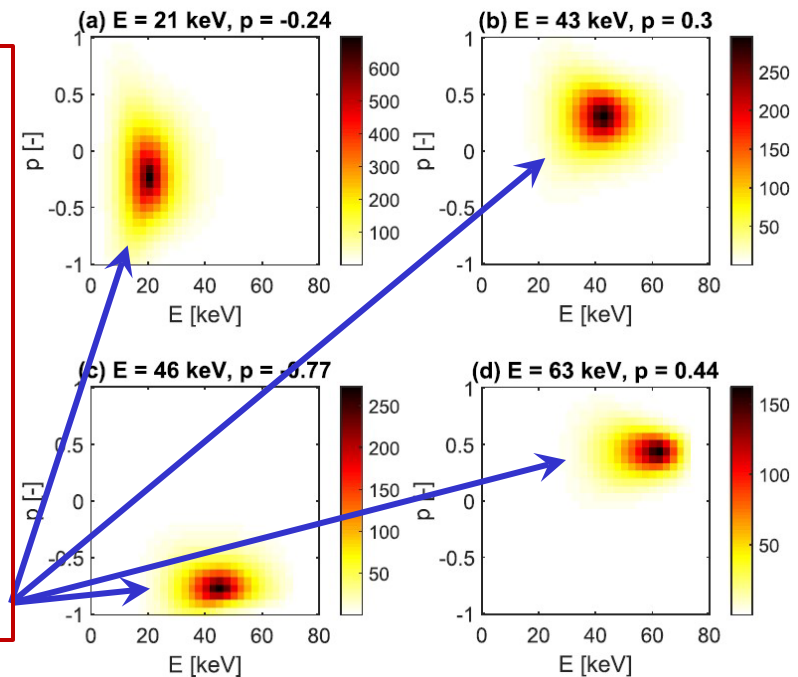
We can interpret the use of Ψ as a regularizer Ψ^{-1} for x in the reg. term.

In the Bayesian framework, the regularization term $\alpha^2 \|\Psi^{-1}x\|_2^2$ represents a Gaussian prior for x with covariance matrix

$$C_x = (\alpha^2 (\Psi^{-1})^T \Psi^{-1})^{-1} = \alpha^{-2} \Psi \Psi^T .$$

This C_x ensures smoothness by correlating a given pixel to the pixels in its vicinity.

Rows of C_x reshaped to (E, p) domain



Interpretation, Part II

And now: the **standard-form transformation**

$$\min_x \|Ax - b\|_2^2 + \alpha^2 \|Lx\|_2^2 \quad \Leftrightarrow \quad \min_{\xi} \|(AL^\#)\xi - b\|_2^2 + \alpha^2 \|\xi\|_2^2$$

where $\xi = Lx \Leftrightarrow x = L^\#\xi$, and $L^\# =$ some kind of inverse of L .

For the case on the previous slide ($n = n_{sd}$) we immediately identify $L = \Psi^{-1}$.

The case $n_{sd} > n \rightarrow$ more exp. functions than pixels (overcomplete system)

The matrix $\Psi = \begin{bmatrix} \circ & \circ \end{bmatrix}$ is “obese” and we assume it has full rank. In this case

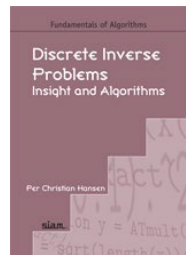
$$L^\# = L^\dagger = \text{pseudoinverse of } L, \quad \text{cf. (H, 2010, §8.4 case 2).}$$

Hence we identify

$$L^\dagger = \Psi \quad \text{and thus} \quad L = (L^\dagger)^\dagger = \Psi^\dagger$$

and

$$C_x = (\alpha^2 (\Psi^\dagger)^T \Psi^\dagger)^{-1} = \alpha^{-2} \Psi \Psi^T, \quad \text{similar to before.}$$



Interpretation, Part III

The case $n_{sd} < n \rightarrow$ fewer expansion functions than pixels

$\Psi = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}$ is “tall & skinny” and we assume full row rank.

Write

$$\Psi = (Q, Q_0) \begin{pmatrix} R \\ 0 \end{pmatrix} = Q R, \quad \text{range}(Q_0) = \text{range}(\Psi)^\perp = \text{null}(\Psi^T).$$

For a general x we must have:

$$x = \Psi c + Q_0 w \Leftrightarrow Q R c = x - Q_0 w \Leftrightarrow c = R^{-1} Q^T x = \Psi^\dagger x.$$

Let $P = Q_0 Q_0^T =$ orthogonal projector on $\text{null}(\Psi^T)$.

We want $x \in \text{range}(\Psi)$ and hence $Q_0 w = P x = 0$, and we arrive at

$$\min_x \|A x - b\|_2^2 + \alpha^2 \|\Psi^\dagger x\|_2^2 \quad \text{s.t.} \quad P x = 0.$$

The constraint is always satisfied because $c = \Psi^\dagger x$. Thus, we get

$$C_x = (\alpha^2 (\Psi^\dagger)^T \Psi^\dagger)^{-1} = \alpha^{-2} \Psi \Psi^T, \quad \text{similar to before.}$$

Ex: Covariance Matrix for a “Skinny” Basis

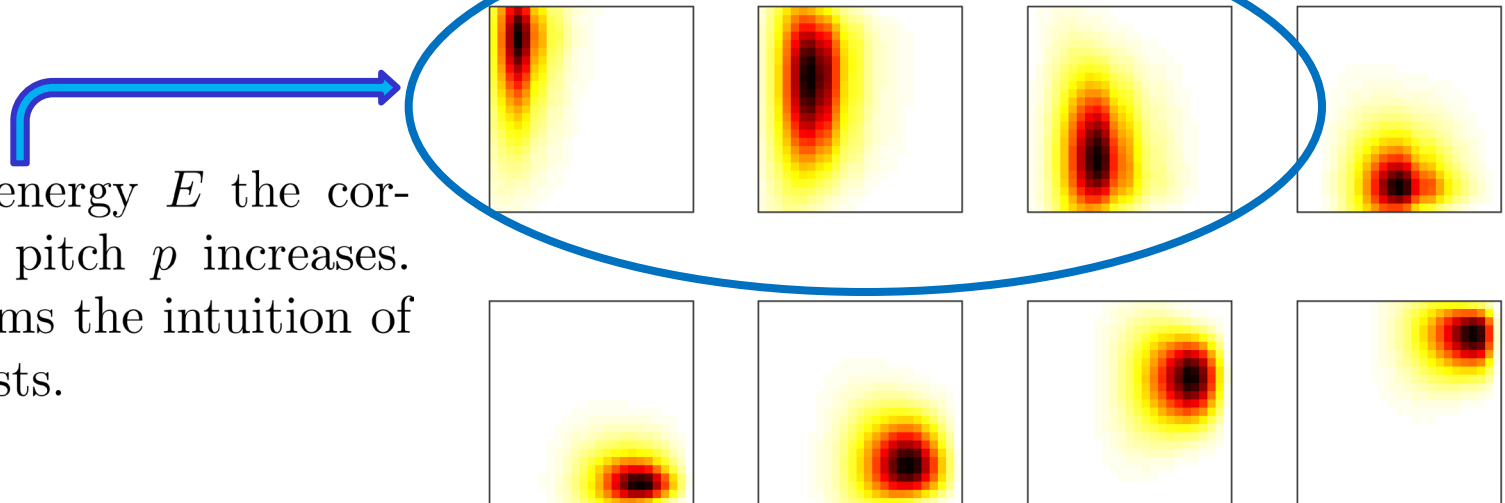
Recap: we want to *interpret the role of the slowing-down basis* Ψ via

- the regularization term $\alpha^2 \|\Psi^\dagger x\|_2^2$ \leftarrow Tikhonov approach
- the covariance matrix $C_x = \alpha^{-2} \Psi \Psi^T$ \leftarrow Bayesian approach

We continue with the case where $\Psi \in \mathbb{R}^{625 \times 170}$ is “tall & skinny,” i.e., we have fewer basis functions ψ_j than the number of unknowns (pixels in the image).

We show **rows** of C_x reshaped to (E, p) domain, as before, they represent local averaging.

For lower energy E the correlation in pitch p increases. This confirms the intuition of the physicists.



Conclusion

- The slowing-down functions provide a set of basis vectors that represent the behavior of the ions in the plasma.
- We can interpret this as a regularization term in the Tikhonov formulation.
- Specifically, the use of these basis functions imposes local smoothing.
- The smoothing that we observe confirms the intuition of the physicists.
- Next step: deal with linearly dependent expansion functions.

References

- B. Madsen, J. Huang, M. Salewski, H. Järleblad, P. C. Hansen + 19, *Fast-ion velocity-space tomography using slowing-down regularization in EAST plasmas with co- and counter-current neutral beam injection*, Plasma Physics and Controlled Fusion, 62 (2020), 115019, doi 10.1088/1361-6587/abb79b.
- B. S. Schmidt, M. Salewski, D. Moseev, M. Baquero-Ruiz, P. C. Hansen + 16, *4D and 5D phase-space tomography using slowing-down physics regularization*, Nuclear Fusion, 63 (2023), 076016, doi 10.1088/1741-4326/acd6a6.

