

## Parameter identification for turbulent transport of fusion plasmas

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We consider a  $\kappa$ - $\varepsilon$  model for turbulences in fusion plasmas reduced to 1D in the radial direction with self-consistent transport [1]. The evolution of the two fields  $\kappa$ - $\varepsilon$  is governed by local dynamics and transverse / parallel transport. The considered model reads:

$$\begin{aligned}\partial_t Z &= \frac{D_{gBZ}}{\rho} \nabla_\rho \left( \rho \frac{Z^2}{Y} \nabla_\rho Z \right) + \gamma_Z Z - K Z^2 - Y, \\ \partial_t Y &= \frac{D_{gBY}}{\rho} \nabla_\rho \left( \rho \frac{Z^2}{Y} \nabla_\rho Y \right) + \gamma_Y Y - \gamma_Z \frac{Y^2}{Z^{3/2}},\end{aligned}$$

where  $Y$  and  $Z$  are the normalised  $\varepsilon$  and  $\kappa$ , and  $\rho$  the normalised radial position.  $D_{gB}$ , the diffusion coefficients,  $\gamma$ , the normalised effective growth rates and  $K$  the Kubo coefficient are the reduced parameters determining the behaviour of the system.

For a given set  $p$  of such parameters, knowing the target sets of vectors  $(Z_i^{obj})$  and  $(Y_i^{obj})$  corresponding with observation times  $t_i$ , we define the following cost function:

$$j(p) = J(Z(p), Y(p)) = \sum_i \frac{1}{2} \left( \|(Z(p))(t_i^{obj}) - Z_i^{obj}\|_2^2 + \|(Y(p))(t_i^{obj}) - Y_i^{obj}\|_2^2 \right) + \Pi(p),$$

where  $\Pi$  is a regularising penalization term, depending directly on the parameters.

Minimising the cost function  $j$  allows us to identify the best parameters  $p$ , in the sense that the corresponding trajectory of the normalised  $\kappa$  -  $\varepsilon$  model is the closest as possible to the target values. The minimisation is performed using a quasi-Newton algorithm [2] and the gradient of the cost function is obtained by automatic differentiation [3].

The identification procedure has been tested on two different cases : spreading with radial variation of the growth rates ratio, and spreading with equal variation of both growth rates and different diffusion coefficients for  $Z$  and  $Y$ . Parameters are globally very well estimated in both cases.

## REFERENCES

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