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Optimization of turbulence reduced model free parameters based on L-mode experiments and 2D transport simulations

Serafina Baschetti^{1*} | Hugo Bufferand¹ | Guido Ciraolo¹ | Nicolas Fedorczak¹ | Philippe Ghendrih¹ | Eric Serre² | Patrick Tamain¹

¹CEA, IRFM, St-Paul-Lez-Durance, France
²M2P2, Aix-Marseille Université, Marseille, France

*Correspondence

Serafina Baschetti, CEA, IRFM, F-13108
St-Paul-Lez-Durance, France.
Email: sera_na.baschetti@cea.fr

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In this paper, a κ – ε transport model is presented as a turbulence reduction tool for a typical ohmic L-mode discharge plasma in a divertor-configured tokamak. Taking a Tokamak à configuration variable (TCV) study case, a feedback loop procedure is performed using the SolEdge2D code to acquire plasma diffusivity at the outer mid-plane. The κ – ε model is calibrated through its free parameters with the aim of recovering the diffusivity calculated in the feedback procedure. Finally, it is shown that the model can self-consistently calculate diffusivity in the whole domain, recovering the poloidal asymmetries due to interchange instabilities.

KEYWORDS

plasma transport, reduced model, turbulence

1 | INTRODUCTION

In order to achieve successful operation of future magnetic confinement fusion devices like ITER and DEMO, it is inevitable that many technological challenges, among which the handling of particle and power exhaust on narrow areas of the machine, notably the divertor targets, are to be faced.

Depending on their composition and geometry, plasma-facing components can tolerate a prescribed amount of heat flux over which degradation and melting can occur. An efficient design of the device involves the understanding of the physics governing the width λ_{SOL} of the scrape-off layer (SOL), which is the distance from the magnetic separatrix over which plasma quantities such as density and temperature are spread on open magnetic field lines. λ_{SOL} strongly depends on the competition between parallel and perpendicular transport (of mass, momentum, and energy), and in particular on turbulence self-organization.

From the quasi-linear analysis of transport equations, assuming a scale separation between fast fluctuations and slowly evolving mean quantities, one can demonstrate^[1] that the fluxes can be linearly related to the gradient of intensive quantities. In that sense, the turbulent particle flux $\Gamma_{\perp}$ is found to be proportional to the gradient of the density, and the proportionality constant $D_{\perp}$ is referred to the turbulent diffusivity: $\Gamma_{\perp} = -D_{\perp} \nabla_{\perp} n$. In this framework,^[2] the SOL width, which is governed by the competition between parallel and perpendicular transport, can be found equalizing $\tau_{\parallel} = L_{\parallel}/c_s$ and $\tau_{\perp} = \lambda_{\text{SOL}}^2/D_{\perp}$, where $\tau_{\parallel}$ and $\tau_{\perp}$ are the parallel and perpendicular particle transport time scales, respectively, $L_{\parallel}$ is the characteristic parallel length of spatial variations along the SOL, and c_s is the sound velocity in the plasma. Thus one obtains $\lambda_{\text{SOL}} = \sqrt{L_{\parallel} D_{\perp}/c_s}$. In practice, the coefficient $D_{\perp}$ is prescribed in order to reproduce the experimental results. This semi-empirical approach lacks the essence of predictability, and so we propose in this paper a model trying to predict this coefficient on a more theoretical basis.

First-principles codes implement self-consistent equations for turbulent transport and would be adequate to provide a theoretical prediction for turbulent transport. However, their applications to realistic-scale tokamaks are still too demanding on numerical resources. To go beyond this limit, an intermediate way between the empirical approach and the first-principles one

would be to develop a reduced model that accounts for the physics driving the turbulence (e.g. interchange instability, flow shear, etc.).

In this work, a first-step turbulence reduction is presented,^[3] inspired by Reynolds averaged Navier–Stokes approach widely used by the computational fluid dynamics (CFD) community, to investigate turbulence of neutral fluids. In the first section, the turbulence intensity κ is defined and the κ – ε model is introduced, identifying interchange instability as the main source of turbulence. Then, an expression for transport diffusivities is given from the turbulence intensity. The reduced approach brings down the number of degrees of freedom. However, since it is not first principle, a few free parameters remain to be set. We propose in this paper a method to set these free parameters from the experimental data. To do so, we present in the second section a typical L-mode plasma on Tokamak à configuration variable (TCV) that we use to tune our model. In the third section, a scan on the free parameters of the κ – ε model is performed to close the problem with a criterion of minimum error between the experimentally inferred diffusivities and the ones calculated with the turbulence model. Finally, first results from the Soledge2D simulation with the κ – ε model activated for the TCV study case are shown.

2 | A κ – ε MODEL FOR TURBULENT DIFFUSIVITY

Being able to predict effective diffusivities describing edge plasma turbulence is a key issue for a proper analysis of the experimental data and even more for extrapolations to future devices with transport codes. From the perspective of numerical simulations for ITER, this means to add to the mass, momentum, and energy balance equations a model to consistently describe the turbulence intensity and its consequence on the transport coefficients. A large variety of transport-reduced models can already be found in the literature.^[4] Coupling turbulence models with transport code have also already been investigated.^[5] In this paper, we follow the approach proposed in ref. ^[3], where a reduced model inspired by the κ – ε approach has been applied to the transport code SolEdge2D-Eirene.

We recall here the definition of the turbulence intensity $\kappa = \frac{1}{2}\langle \tilde{u}^2 \rangle$, where $\tilde{u}$ is the oscillating component of the velocity field, which describes the energy associated with fluctuations. The evolution of κ is assumed to be governed by the transport equation, where the source, the damping, and the transport of turbulence are described as

$$\partial_t(n\kappa) + \nabla(n\kappa v_\perp^d + n\kappa u_\parallel \mathbf{b}) = \nabla(nD_\kappa \nabla_\perp \kappa) + \gamma \kappa n + S_n^{iz} \kappa - n\varepsilon, \quad (1)$$

where $v_\perp^d$ is the perpendicular component of velocity due to drifts, $\mathbf{b} = \mathbf{B}/B$, γ is the interchange instability growth rate, S_n^{iz} is a source term due to ionization of neutrals, and $n\varepsilon$ is a sink term characterizing turbulence damping. In this work, we consider that ε is the self-saturation of turbulence: $\varepsilon = \Delta\omega\kappa^2$, where the self-saturation coefficient is defined as $\Delta\omega = \Delta\omega_0 \frac{c_s}{R}$, R being the device's major radius.

Concerning the turbulence drive, the growth rate of the interchange instability is given by

$$\gamma = \frac{c_s}{R} \cdot \left(\gamma_0 \sqrt{R^2 \frac{\nabla p_i \cdot \nabla B_\phi}{p_i B_\phi}} - \Theta + \beta T_e^2 \right), \quad (2)$$

where $\Theta = \theta_0 1 + T_i/T_e$ is the threshold triggering the instability.^[6] The term proportional to $\nabla p_i \cdot \nabla B_\phi$ makes interchange growth rate positive only on the tokamak's low-field-side region. Assuming $\frac{\nabla B_\phi}{B_\phi} \approx \frac{1}{RB_\phi}$, one can simplify $R^2 \frac{\nabla p_i \cdot \nabla B_\phi}{p_i B_\phi} = \frac{R}{\lambda_p}$, where λ_p is the pressure radial decay length.

One notices that in Equation 2 the term βT_e^2 has been added heuristically to model instabilities occurring at the core. It is regulated by the β parameter, which will be investigated in Section 3. Neglecting the transport terms and subtracting κ times the continuity equation, Equation 1 reduces to

$$\partial_t \kappa = \gamma \kappa - \Delta\omega \kappa^2. \quad (3)$$

At the steady state, the turbulence intensity is thus simply given by $\kappa = \gamma/\Delta\omega$. Transport diffusivities are derived from the turbulence intensity according to the expression proposed in ref. ^[3]:

$$D_{\kappa\varepsilon} = \frac{R}{c_s} \cdot \kappa = \frac{1}{\Delta\omega} \cdot \left(\gamma_0 \sqrt{\frac{R}{\lambda_p}} - \Theta + \beta T_e^2 \right). \quad (4)$$

In Equation 4, four free parameters have to be determined in order to close the problem: $\Delta\omega_0$, γ_0 , θ_0 , β . It is possible to group them as

$$D_{\kappa\varepsilon} = \alpha_0 \sqrt{\frac{R}{\lambda_p}} - \Theta + \zeta T_e^2, \quad (5)$$

where $\alpha_0 = \gamma_0/\Delta\omega_0$ and $\zeta = \beta/\Delta\omega_0$. We choose $\theta_0 = 5$ according to ref. ^[6].

3 | TCV REFERENCE TEST CASE AND EXTRACTION OF THE DIFFUSION COEFFICIENT

In order to determine the free parameters of the reduced turbulence model described in the previous section, we use a typical L-mode ohmic discharge on the TCV studied in ref. [2] (TCV #51333). The main plasma characteristics are given in Table 1.

The magnetic configuration of the discharge is shown in Figure 1.

TABLE 1 Main discharge characteristic of shot Tokamak à configuration variable (TCV) #51333

B_t (T)	I_p (kA)	P_{Ohm} (MW)	$n_{e,sep}$ (m ⁻³)
1.4	210	0.21	6.8×10^{18}

FIGURE 1 Magnetic configurations of the Tokamak à configuration variable (TCV) #51333 (by permission from ref. [2]). We consider the medium leg configuration (blue solid line). HRTS collection points are shown in red

Electron density and temperature profiles are obtained in the outboard mid-plane by high-resolution Thomson scattering. The sampling points are shown in Figure 1. Additionally, measurements of electron temperature and density are obtained on the divertor targets by Langmuir probes.

In order to estimate the intensity of the turbulent transport in the TCV shot described above, we analyse the experimental data with the transport code SolEdge2D-Eirene.^[7] As any transport code, SolEdge2D requires providing effective cross-field diffusivities to properly account for the turbulent transport. In a “forward” simulation, one provides to the code ad hoc values for the transport coefficients as input and obtains a 2D poloidal map of the plasma density and temperature as output. Experimental mid-plane profiles can thus be compared with simulated ones, and if the match is not satisfactory, one must iterate on the transport coefficient values to improve the agreement between the two. In this paper, we develop a procedure to simplify this iterative process. Instead of providing transport coefficients as input to the code, we directly provide mid-plane profiles of the density and temperature. The code will automatically adjust the transport coefficients so as to match in the end the simulated mid-plane profiles with the experimental ones. More precisely, one uses the proportional–integral feedback loop shown in Figure 2 to correct “on the fly” the transport coefficients while the temporal loop of the code is running.

The gain of the feedback loop τ_i has been adjusted to ensure the code stability, while the difference of signals ε is given by the following formula:

$$\varepsilon = D_{\perp} - D_{\perp 0} = \frac{1}{\tau_i} \int (\partial_r n_{XP} - \partial_r n_{simu}) dt + K(\partial_r n_{XP} - \partial_r n_{simu}), \quad (6)$$

FIGURE 2 Feedback loop used in the code to match simulated mid-plane profiles of density and temperature with the experimental ones. We set the gain and integration time in such a way that the dynamic of the feedback loop is of the order of the parallel plasma characteristic time

FIGURE 3 Left and middle: Mid-plane profiles of density and temperature (symbol: experimental data, lines: SolEdge2D-Eirene simulation results with feedback procedure) from Ref. [12]. Right: Cross-field transport coefficients obtained from the feedback procedure

where the subscripts “XP” and “simu” stand for “experimental” and “simulation”, respectively. Typically, the time constant associated with the integral correction must be larger than the parallel time $\tau_{\parallel} = L_{\parallel}/c_s (\sim 10^{-5} \text{ s})$. The technical details on the implementation of the procedure are beyond the scope of this paper.

Using this automatic feedback loop procedure, the way of running SolEdge2D is made simple for experimental interpretation. As input of the code, one directly provides experimental measurements of density and temperature in the mid-plane. As output, one still obtains a 2D poloidal map of the density and temperature but additionally gets a 1D profile of the cross-field diffusivities that were ad hoc earlier. The poloidal variation of these coefficients is, however, not taken into account. Since we set the 1D radial profiles of density and temperature at the outer mid-plane, we get 1D radial profiles of diffusivities. The diffusivities are assumed to be homogeneous on each flux surface. Figure 3 shows results of the feedback procedure on the transport coefficients applied to TCV #51333. In this simulation, we consider a pure deuterium plasma. Drift velocities are not taken into account.

Once the code’s free parameters (namely transport coefficients) are set by the feedback procedure, we know that experimental and simulated mid-plane quantities are in rather good agreement.

4 | REDUCTION AND OPTIMIZATION OF THE DEGREES OF FREEDOM OF THE MODEL

After running SolEdge2D-Eirene as an interpretative tool on TCV #51333, one obtains 2D poloidal maps of density and temperature as well as an estimation of the transport coefficients required to describe the turbulent transport of the discharge in a diffusive fashion. From these data, we propose in this section to optimize the free parameters of our turbulence reduced model. To do so, we apply Equation 4 to the maps of density and temperature computed by SolEdge2D-Eirene for different values of the free parameters of the reduced turbulence model, and compare the diffusivity maps with the transport coefficients obtained from the experiment with the feedback procedure.

As seen in Section 1, we can rewrite Equation 5 as a function of two free parameters: α_0 and ζ .

A scan is performed on these two quantities in order to minimize the error between the diffusivities computed with the reduced κ - ε model ($D_{\perp, \kappa \varepsilon}$) and those obtained from the experimental analysis (feedback procedure, $D_{\perp, fb}^{1D}$). In Figure 4, the absolute error $Err(\alpha_0, \zeta) = \|D_{\perp, \kappa \varepsilon} - D_{\perp, fb}^{1D}\|_2$ is parameterized with respect to the two free parameters. The minimum error is found to be for $\alpha_0 = 1.33 \times 10^3 \text{ (m}^2/\text{s}^2)$ and $\zeta = 1.3 \times 10^{-3} \text{ (m}^2 \text{ s}^{-1} \text{ eV}^{-2})$. The minimum absolute error is found to be $Err_{min} = 2 \times 10^{-2} \text{ (m}^2/\text{s)}$,

FIGURE 4 Scan of the relative error $Err(\alpha_0, \zeta) = ||D_{\perp, \kappa\epsilon} - D_{\perp, fb}^{1D}||$ with respect to the free parameters

FIGURE 5 Left: Feedback loop $D_{\perp, fb}^{1D}$ vs. $D_{\perp, \kappa\epsilon}$ vs. Bohm diffusivity D_{Bohm} (the dotted line is the position of the separatrix): flux surface-averaged profiles. Although usually considered as the standard scaling for diffusion coefficients, and in particular an upper limit, D_{Bohm} is actually too small to account for blob convective transport particularly effective in the SOL region,^[8] as one can see in the plot. Right: 2D map of diffusivity self-consistently calculated with the $\kappa-\epsilon$ model. Strange behaviour in the far SOL is due to numerical effects (thresholding of the density)

FIGURE 6 Mid-plane profiles of density and temperature (*symbol*: experimental data, *lines*: SolEdge2D-Eirene simulation profiles auto-consistently calculated)

which gives a relative error $Err_{rel, min} = 8.6\%$. Figure 5 shows 2D poloidal maps and flux surface-averaged profile of the cross-field diffusivities obtained from the reduced turbulent model for the optimum values of the free parameters α_0 and ζ . It is important to stress that the diffusivity is determined for the whole geometrical domain and that the model captures the poloidal asymmetry of the plasma transport due to interchange instability; in particular, the ballooning effect at the low magnetic field side (LFS) is recovered for a divertor geometry as a strong poloidal variation of turbulence transport.^[3]

We have implemented the optimized $\kappa-\epsilon$ model into SolEdge2D. The results of this simulation are shown in Figure 6.

To run the simulation, ion and electron thermal conductivity have been set proportional to the self-consistently calculated diffusivity. This is an assumption whose validity will not be discussed in this paper but will be investigated in the future work.

FIGURE 7 Profiles of electron density and temperature at the outer target compared to data collected by the Langmuir probes for shot #51333. Here, s is the curvilinear abscissa along the wall of the vessel, while s_{SP} is the position of the separatrix at the outer target

FIGURE 8 Flow chart representing the loop to infer a self-consistently calculated diffusivity into SolEdge2D. The green path shows the feedback procedure, the red one is the $\kappa-\epsilon$ optimization strategy, with $Err(\alpha_0, \zeta) = ||D_{\perp, \kappa\epsilon} - D_{\perp, fb}^{1D}||$

The outer mid-plane profiles in Figure 6 resulting from the simulation with self-consistently calculated and optimized diffusivity show good accordance with the experimental data. A qualitative accordance is also found for the outer target profiles of electron density and temperature, as can be seen from Figure 7. The discrepancy observed for the density target profile is under investigation.

5 | CONCLUSIONS

A $\kappa-\epsilon$ model has been introduced with the aim of solving the Reynolds averaged Navier–Stokes equations for the SOL plasma. A method to tune the free parameters of the model from experimental data has been presented as well. A summary of this work is sketched in Figure 8. The green line represents the feedback loop between experimental data and SolEdge2D free parameters (anomalous diffusivities). The red line loop represents the strategy consisting of optimizing the $\kappa-\epsilon$ free parameters to recover diffusivity profiles and perform a more consistent comparison between simulation and experimental data, at the outer mid-plane as well as at the outer target. This $\kappa-\epsilon$ procedure allows us to reduce the original number of free parameters and capture the poloidal asymmetries of turbulence, which is a promising first step towards predictive transport models for tokamak plasmas.

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