

4<sup>th</sup> Asia-Pacific Conference on Plasma Physics, 26-31Oct, 2020, Remote e-conference **Mapping Turbulence and its consequences around a reconnection region** 

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Reconnection is often accompanied by turbulence. The two seem to be disjoint only in the ideal world of 2D models and simulations but in 3D their relationship becomes stronger. An example from a massively parallel simulation with iPic3D is reported below [1].



Figure 1: Two sections that show the turbulent outflow originating from the right end of the domain shown (it is only half of the computational box) at x=20. The false colours show the current density. Turbulence is clearly present in the outflow.

Turbulence alters the evolution of reconnection and reconnection causes flows that shape the regions of turbulence and further promote turbulence. To understand these processes, we need to design methods that can measure quantitatively the processes of reconnection and turbulence in 3D [1-2]. This is far from trivial. In 3D even defining what is and what isn't reconnection can be a challenge.

Within the project European Commission funded Horizon 2002 project AIDA (<u>www.aida-space.eu</u>), we have recently developed a combination of deterministic physics-based methods and statistical machine learning techniques to investigate turbulence and reconnection.

We report here the combined use of physics-based methods (e.g. topological indicators, agyrotropy, distribution functions, structure functions) and unsupervised machine learning methods (e.g. DBSCAN [2-3], Gaussian Mixtures [4], SOMs [5]) to identify the role of turbulence and that of reconnection in the energy transformations developing in regions of reconnection

characterized by varying degrees of guide filed, representative of conditions in the tail, in the dayside and in the solar wind.

References

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