

DE LA RECHERCHE À L'INDUSTRIE

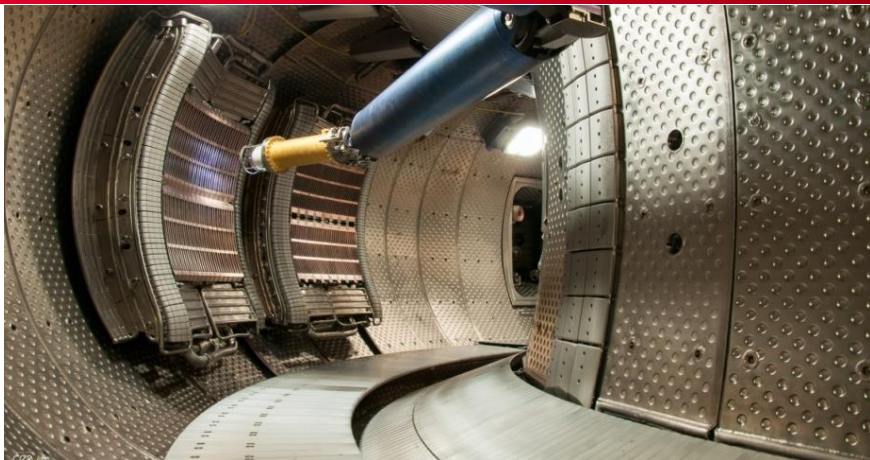


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First WEST experimental data analysis using machine learning algorithms

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I_p ($q_{95} \sim 2.5$)	1 MA
B_T	3.7 T
R	2.5 m
a	0.5 m
A (aspect ratio)	5 – 6
n_{GW} (1MA)	$1.5 \cdot 10^{20} \text{m}^{-3}$
P_{ICRH}	9 MW
P_{LHCD}	7 MW
time_{flattop} (0.8 MA)	1000 s

Diagnostics organized into groups related to their measurement principle

Magnetics

Neutrons

X-Ray measurements

- Soft X-ray (impurities)
- Hard X-ray (fast electron)

Optical

- Interferometry (density)
- Polarimetry
- Thomson scattering
- Visible cameras
- Infrared cameras

Spectroscopy

- Visible (plasma edge)
- UV (plasma edge-core)
- 2D X-ray (T ions at plasma core)

Bolometry (plasma radiation)

Microwave

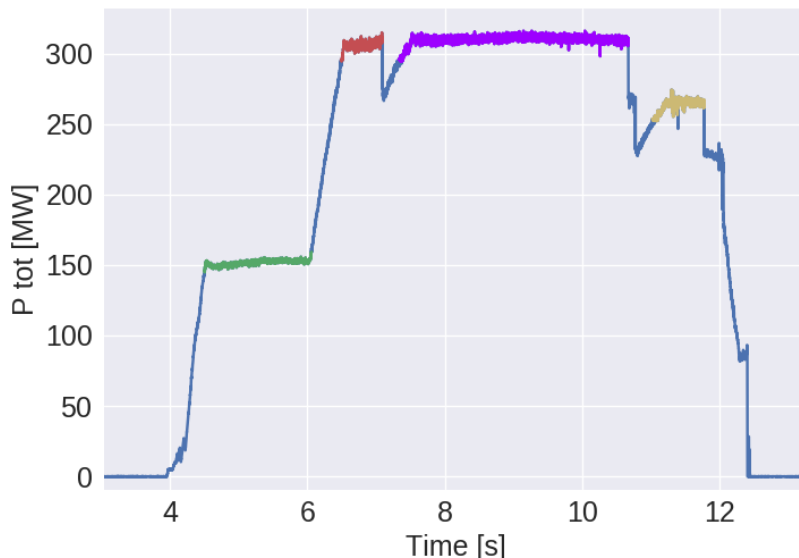
- Reflectometers (density)
- Electron cyclotron emission (T electrons)

Plasma Wall Interaction

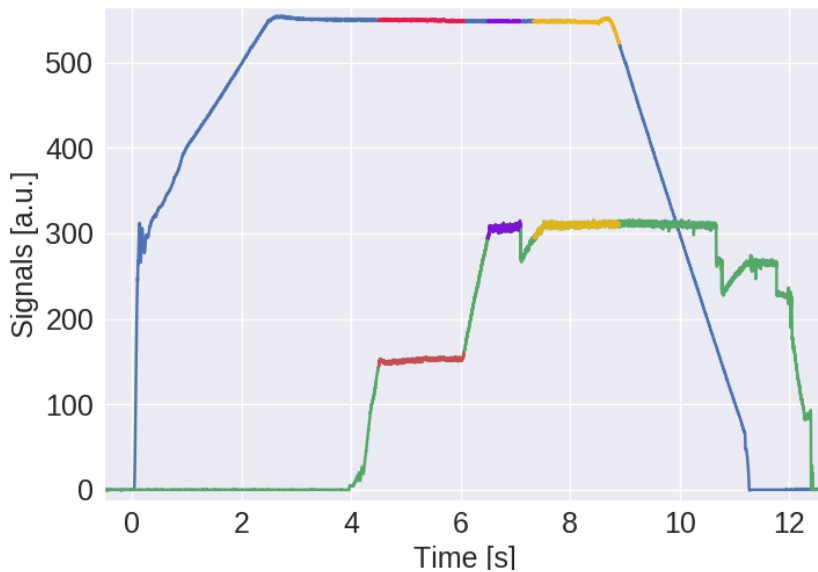
- Reciprocating probes
- Pecker probes
- Langmuir. probes
- Calorimetry
- Thermocouples
- Barometry
- Mass spectroscopy

Total **50 diagnostics**

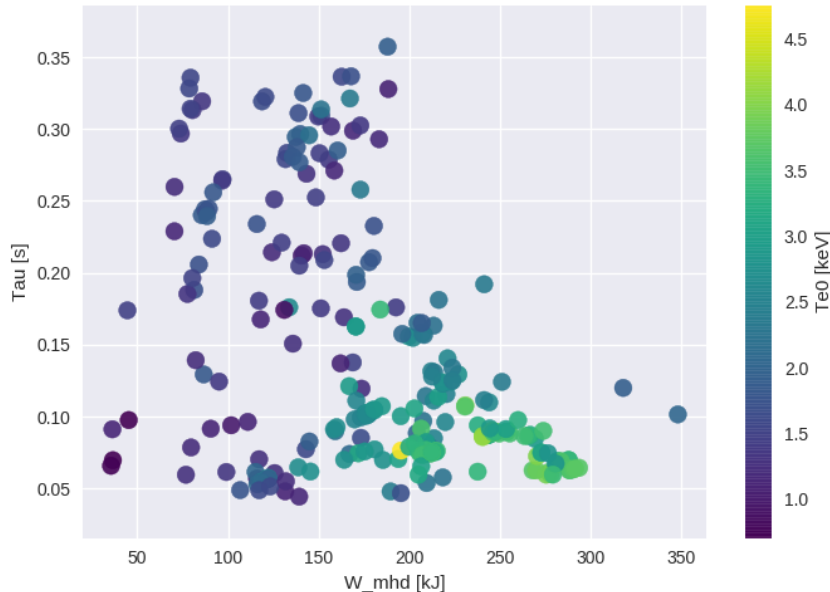
- We are interested on **steady-states** reached by the plasma ► need for a **plateau recognition algorithm**
- In this case we use a Kernel density linear estimator provided by scikit-learn



- We are interested on **steady-states** reached by the plasma ► need for a **plateau recognition algorithm**
- In this case we use a Kernel density linear estimator provided by `scikit-learn`
- Also we need to find steady-states taking into account more than one single signal



- Using the plateau recognition algorithm a reduced WEST database is constructed from statistical moments calculated at each detected plateau
- We are interested in studying the confinement quality using the plasma **stored energy** W_{mhd} and the **confinement time** $\text{Tau} = \frac{W_{\text{mhd}}}{P_{\text{tot}}}$ (main target features)
- First constructed reduced database: **36 features** (diagnostic measurements) and **254 samples** (detected plateaus)



Plot example of mean quantities at detected plateaus

Three different feature selection algorithms are tested:

- **Pearson correlation:** measures the linear correlation between variables
- **Mutual information:** measures linear and non-linear dependences between variables
- **Random forest:** feature importance estimated by how much the tree node that use that feature reduce impurity on average in all decision trees

First data analysis

Main feature detection for W_{mhd} target

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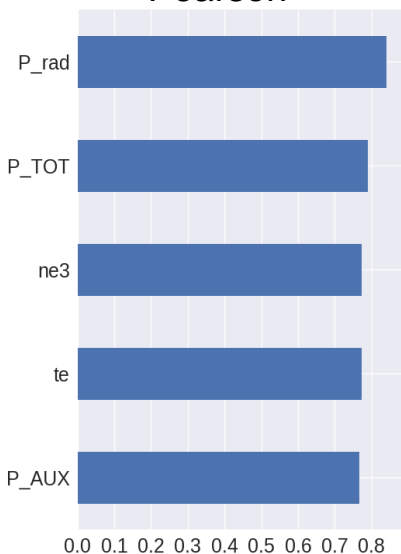
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P_AUX	1	0.99	1	0.98	0.77	0.66	0.67	0.77
P_COND	0.99	1	0.99	0.95	0.71	0.62	0.64	0.72
P_TOT	1	0.99	1	0.99	0.79	0.67	0.68	0.76
P_rad	0.98	0.95	0.99	1	0.84	0.7	0.69	0.77
eq_w_mhd	0.77	0.71	0.79	0.84	1	0.77	0.74	0.77
ne3	0.66	0.62	0.67	0.7	0.77	1	0.99	0.44
neVol	0.67	0.64	0.68	0.69	0.74	0.99	1	0.4
te	0.77	0.72	0.76	0.77	0.77	0.44	0.4	1
	P_AUX	P_COND	P_TOT	P_rad	eq_w_mhd	ne3	neVol	te

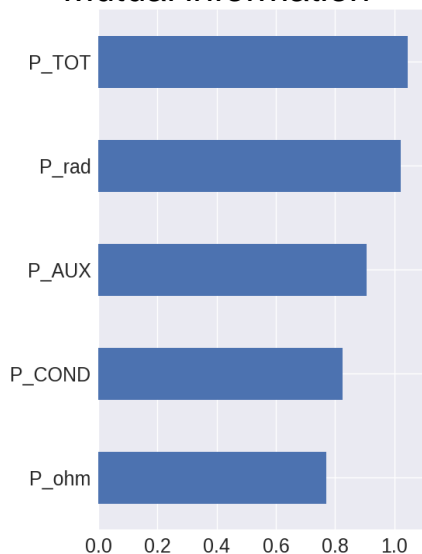
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- **Pearson correlation:** measures the linear correlation between variables
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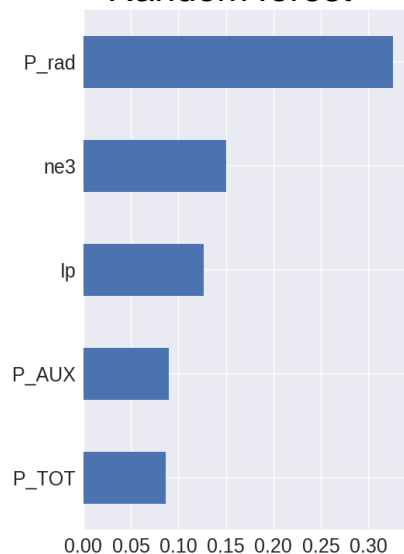
Pearson

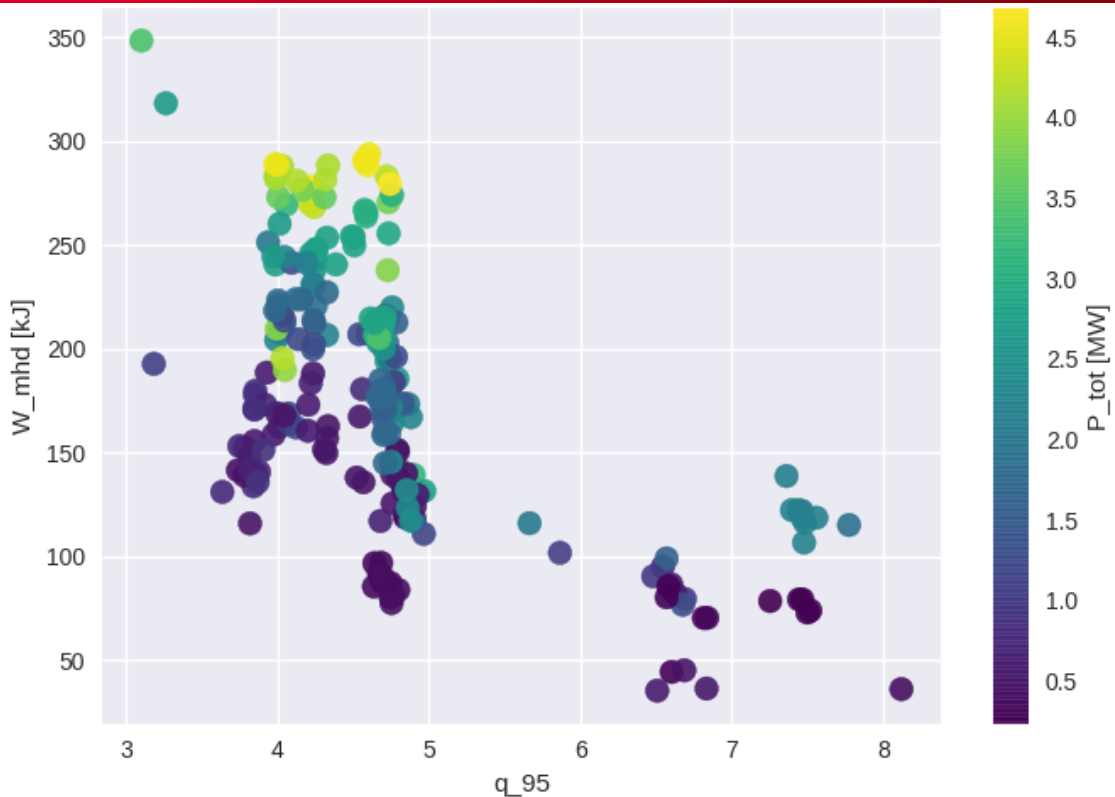


Mutual information



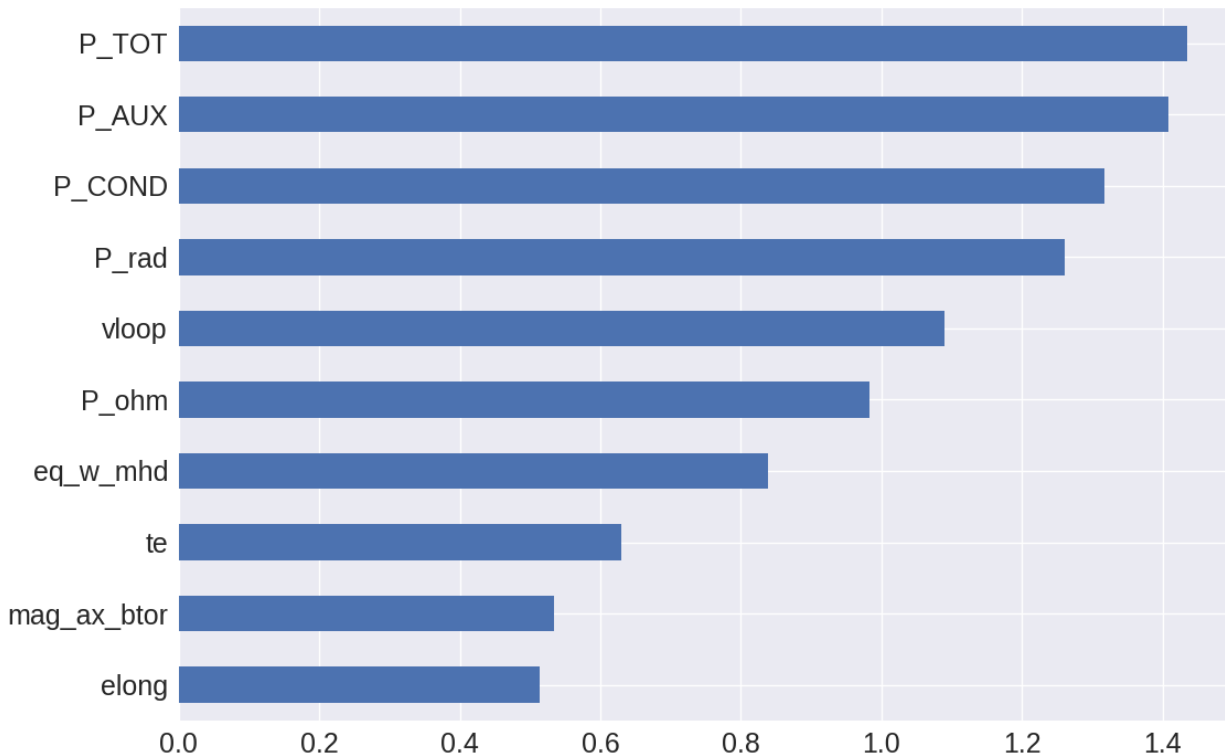
Random forest





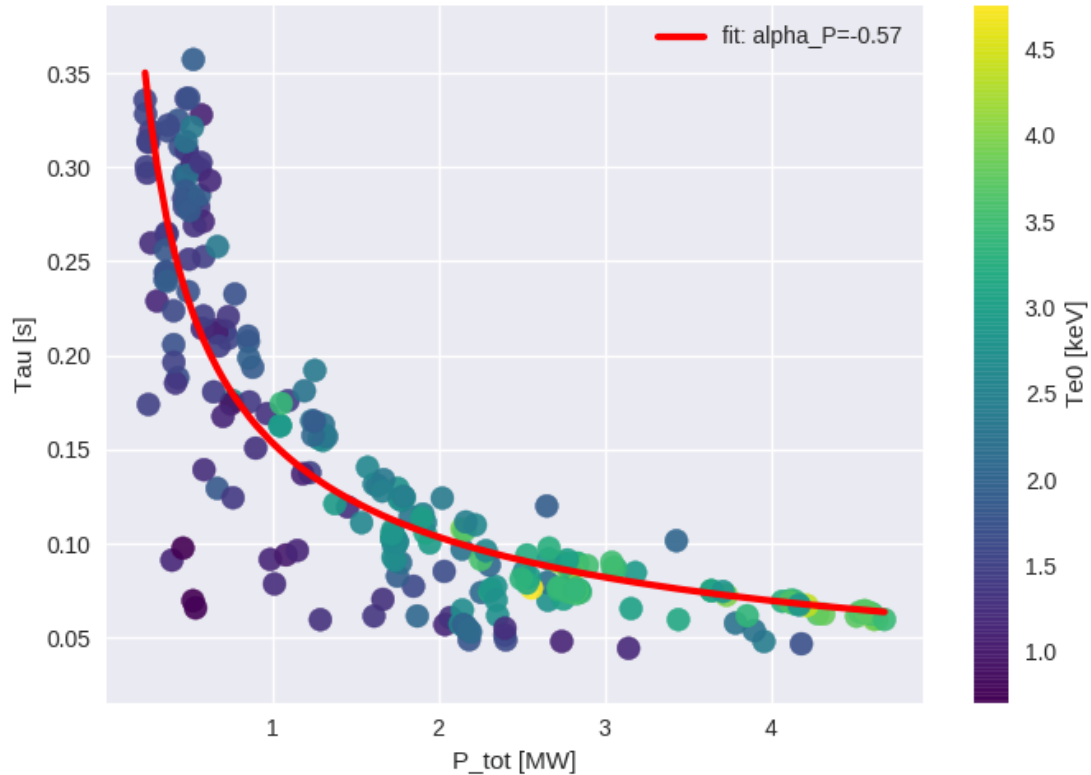
Stored plasma energy decreases as a function of q_{95} (edge safety factor) for a given total power level

Mutual information is used to detect nonlinear relations between variables



First data analysis

Confinement time scaling with total power



In the literature total power exponents in scaling laws vary between: -0.73 and -0.5 (our value: -0.57)

Summary

- To carry out plasma steady-state analyses a WEST reduced database is created. To this aim a plateau recognition algorithm for time signals is developed
- The plasma energy content measured by W_{mhd} and the confinement time (Tau) are the selected targets
- To detect the principal features impacting the targets three feature selection algorithms are tested (Pearson correlation, mutual information and random forest)
- The evolution of the target variables as a function of the main features is analyzed. Comparison with literature shows good agreement in the total power exponent computed for Tau scaling

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Perspectives

- More diagnostics data will be added to reduced database
- Failure and error detection algorithms for diagnostics
- Improve main feature selection, regression for targets of interest, WEST pulses classification... More ideas are more than welcomed!