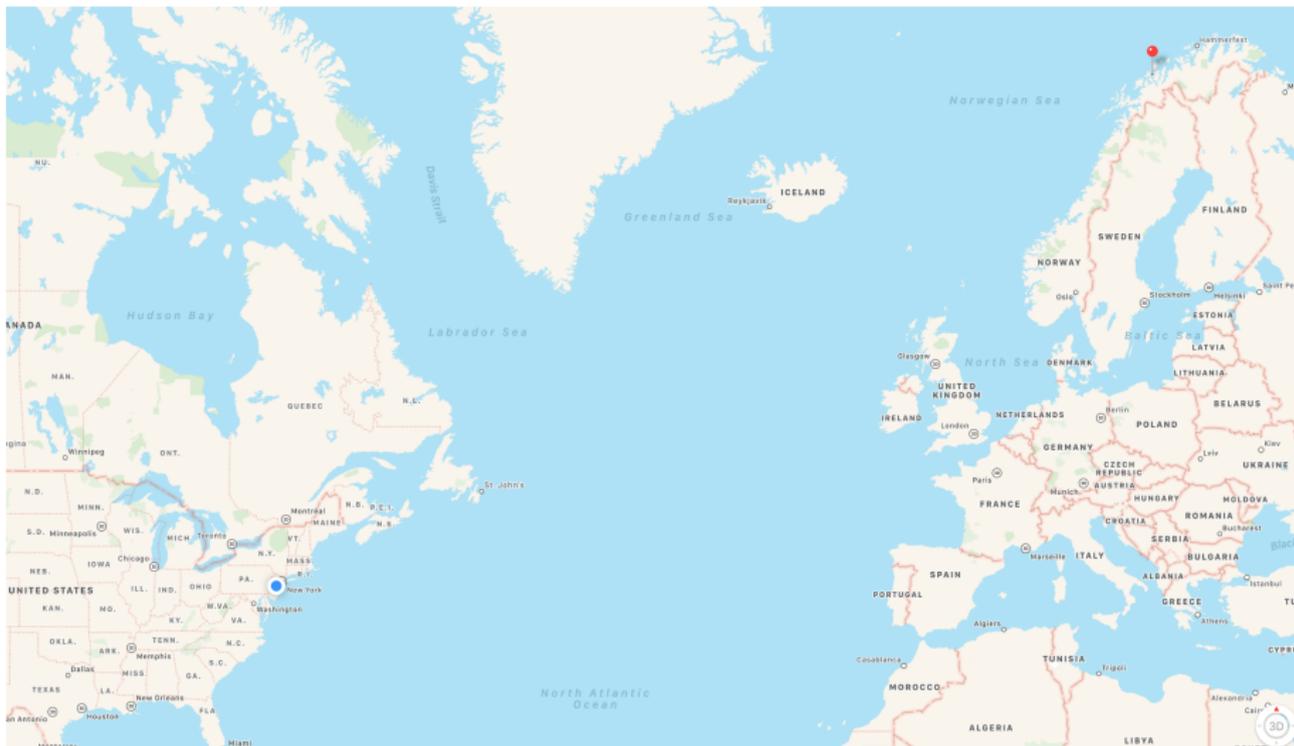


# Stochastic modeling of scrape-off layer fluctuations

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# Bursts in single point measurements correspond to traversing blobs

- 1 Stochastic model of data time series
- 2 Comparison to experimental measurements
- 3 Conclusions

# Superpose uncorrelated pulses to model data time series

Superposition of  $K$  pulses in a time interval  $[0 : T]$

$$\Phi_K(t) = \sum_{k=1}^{K(T)} A_k \phi \left( \frac{t - t_k}{\tau_d} \right)$$

where  $k$  labels a pulse and

- $A_k$  denotes the pulse amplitude
- $t_k$  denotes pulse arrival time
- $\phi$  denotes a pulse shape
- $\tau_d$  denotes pulse duration time

Intermittency parameter:  $\gamma = \tau_d / \tau_w$

# Pulses arrive uncorrelated and form a Poisson process

Choose distribution for all random variables

- $P_K(K|T)$  gives the number of bursts in time interval  $[0; T]$
- $P_A(A_k) \rightarrow$  distribution of pulse Amplitudes.
- $P_t(t_k) \rightarrow$  distribution of pulse arrival times.

Consider a Poisson process:

- 1 Pulses arrive uncorrelated:  $P_t(t_k) = 1/T$
- 2 Avg. rate of pulse arrival is  $1/\tau_w$

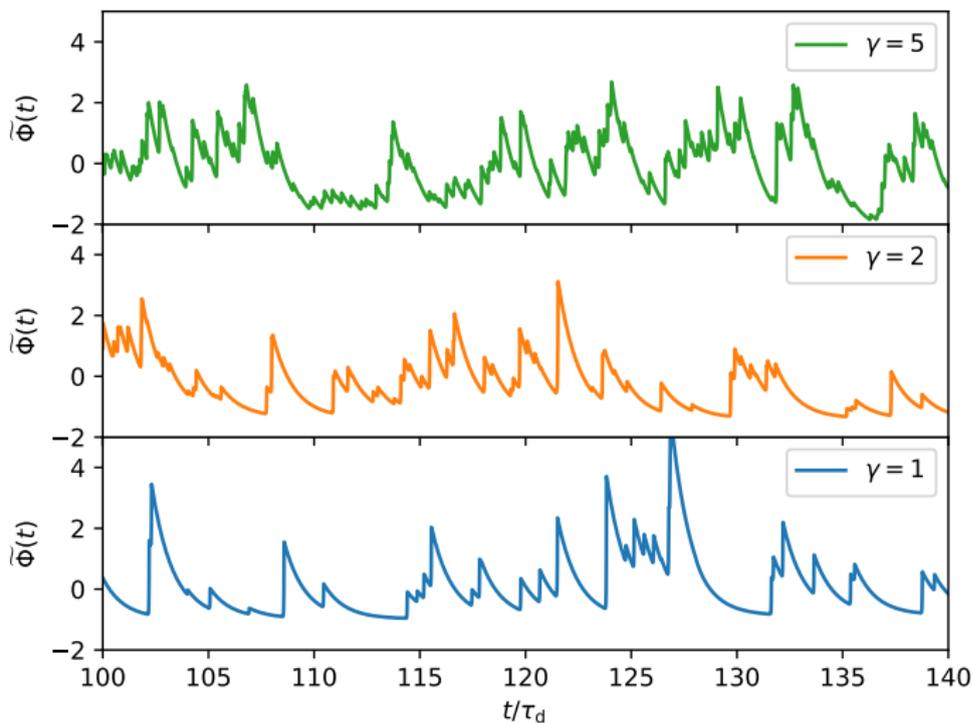
$$P_K(K|T) = \exp\left(\frac{-T}{\tau_w}\right) \left(\frac{T}{\tau_w}\right)^K \frac{1}{K!}$$

Exponentially distributed pulse amplitudes:  $\langle A \rangle P_A(A_k) = \exp(-A_k/\langle A \rangle)$

We often normalize the process as

$$\tilde{\Phi} = \frac{\Phi - \langle \Phi \rangle}{\Phi_{\text{rms}}}$$

# Intermittency parameter governs pulse overlap



## Model experimental data with double-exponential pulses

Experimental data is approximated by a double-exponential pulse shape

$$\phi(\theta) = \Theta(-\theta) \exp\left(\frac{\theta}{\lambda}\right) + \Theta(\theta) \exp\left(-\frac{\theta}{1-\lambda}\right)$$

In physical units:  $\theta = (t - t_k)/\tau_d$ ,  $\tau_d \approx 10\mu\text{s}$ .

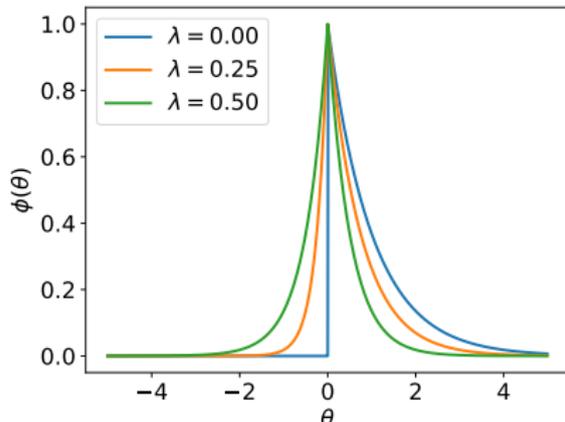
$\lambda$  defines pulse asymmetry:

$$\tau_r = \lambda\tau_d$$

$$\tau_f = (1 - \lambda)\tau_d$$

Notation:  $I_n = \int_{-\infty}^{\infty} d\theta [\phi(\theta)]^n$

Normalization:  $I_1 = 1$



# Correlation and power spectral density depend on pulse asymmetry

Correlation function of the pulse shape is given by

$$\begin{aligned}\rho_\phi(\theta) &= \frac{1}{l_2} \int_{-\infty}^{\infty} d\chi \phi(\chi) \phi(\chi + \theta) \\ &= \frac{1}{1 - 2\lambda} \left[ (1 - \lambda) \exp\left(-\frac{|\theta|}{1 - \lambda}\right) - \lambda \exp\left(-\frac{|\theta|}{\lambda}\right) \right]\end{aligned}$$

Wiener-Khinchin theorem states that the power spectral density is the Fourier-transform of the autocorrelation function

$$\begin{aligned}\sigma_\phi(\omega) &= \int_{-\infty}^{\infty} d\theta \rho_\phi(\theta) \exp(-i\omega\theta) \\ &= \frac{2}{[1 + (1 - \lambda)^2 \omega^2][1 + \lambda^2 \omega^2]}\end{aligned}$$

## The mean of the process can be computed analytically

Averaging the process over all random variables and neglect finite box effects by extending time integration to the entire real axis:

$$\begin{aligned} \langle \Phi_K \rangle &= \int_{-\infty}^{\infty} dA_1 P_A(A_1) \int_{-\infty}^{\infty} \frac{dt_1}{T} \dots \int_{-\infty}^{\infty} dA_K P_A(A_K) \int_{-\infty}^{\infty} \frac{dt_K}{T} \sum_{k=1}^K A_k \phi \left( \frac{t - t_k}{\tau_d} \right) \\ &= \frac{K}{T} \tau_d \langle A \rangle \end{aligned}$$

Average over number of pulses  $K$ :

$$\langle \Phi \rangle = \frac{\tau_d}{\tau_w} \langle A \rangle$$

Mean value of the process increases with pulse overlap and average pulse amplitude.

# The variance can be computed analytically

$$\langle \Phi_K^2 \rangle = \int_{-\infty}^{\infty} dA_1 P_A(A_1) \int_{-\infty}^{\infty} \frac{dt_1}{T} \dots \int_{-\infty}^{\infty} dA_K P_A(A_K) \int_{-\infty}^{\infty} \frac{dt_K}{T} \\ \sum_{k=1}^K A_k \phi\left(\frac{t-t_k}{\tau_d}\right) \sum_{l=1}^K A_l \phi\left(\frac{t-t_l}{\tau_d}\right)$$

This results in  $K(K-1)$  terms with  $k \neq l$ ,  $K$  terms with  $k = l$ .

$$\langle \Phi_K^2 \rangle = \tau_d I_2 \langle A^2 \rangle \frac{K}{T} + \tau_d^2 I_1^2 \langle A \rangle^2 \frac{K(K-1)}{T^2} \\ \Rightarrow \langle \Phi^2 \rangle = \frac{\tau_d}{\tau_w} I_2 \langle A^2 \rangle + \langle \Phi \rangle^2$$

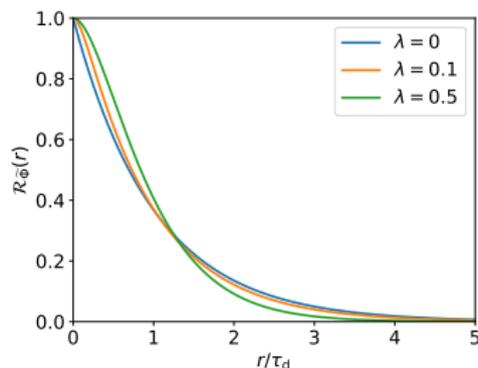
where  $\langle K(K-1) \rangle = \langle K \rangle^2$  has been used.

# Auto-correlation is determined by the pulse shape

Auto-correlation function is computed from  $\langle \Phi(t)\Phi(t+k) \rangle$

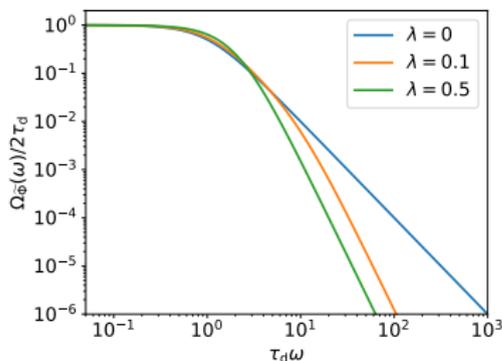
$$R_{\Phi}(r) = \langle \Phi \rangle^2 + \Phi_{\text{rms}}^2 \rho_{\phi} \left( \frac{r}{\tau_d} \right)$$

$$= \langle \Phi \rangle^2 + \frac{\Phi_{\text{rms}}^2}{1 - 2\lambda} \left[ (1 - \lambda) \exp \left( -\frac{|r|}{(1 - \lambda)\tau_d} \right) - \lambda \exp \left( -\frac{|r|}{\tau_d} \right) \right]$$



# Power spectral density

$$\begin{aligned}\Omega_{\Phi}(\omega) &= 2\pi\langle\Phi\rangle^2\delta(\omega) + \Phi_{\text{rms}}^2\tau_d\sigma_{\phi}(\tau_d\omega) \\ &= 2\pi\langle\Phi\rangle^2\delta(\omega) + 2\Phi_{\text{rms}}^2\frac{\tau_d}{\left[1 + (1 - \lambda)^2\tau_d^2\omega^2\right]\left[1 + \lambda^2\tau_d^2\omega^2\right]}\end{aligned}$$

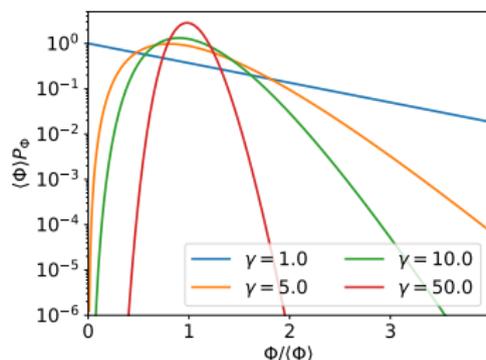


- $\lambda = 0$ : Power law tail,  $\sim \omega^{-2}$
- $\lambda = 1/2$ : Power law tail,  $\sim \omega^{-4}$
- Else: broken power law, curved spectrum.

# Probability distribution function

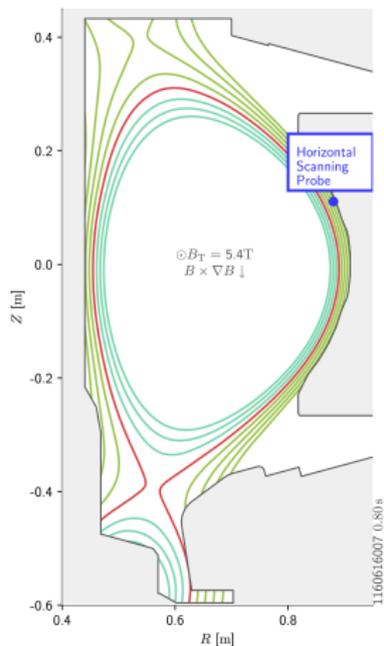
For exponentially distributed amplitudes and exponential wave forms is the process Gamma distributed:

$$\langle \Phi \rangle P_{\Phi}(\Phi) = \frac{\gamma}{\Gamma(\gamma)} \left( \frac{\gamma \Phi}{\langle \Phi \rangle} \right)^{\gamma-1} \exp \left( -\frac{\gamma \Phi}{\langle \Phi \rangle} \right)$$



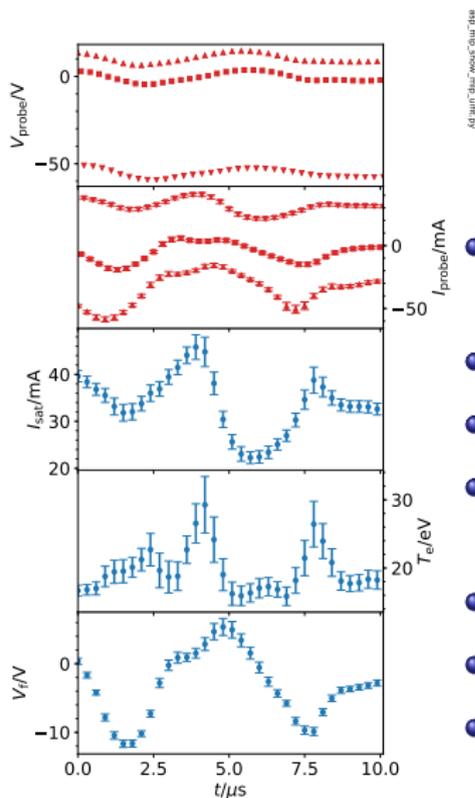


# SOL fluctuations measured in a density scan



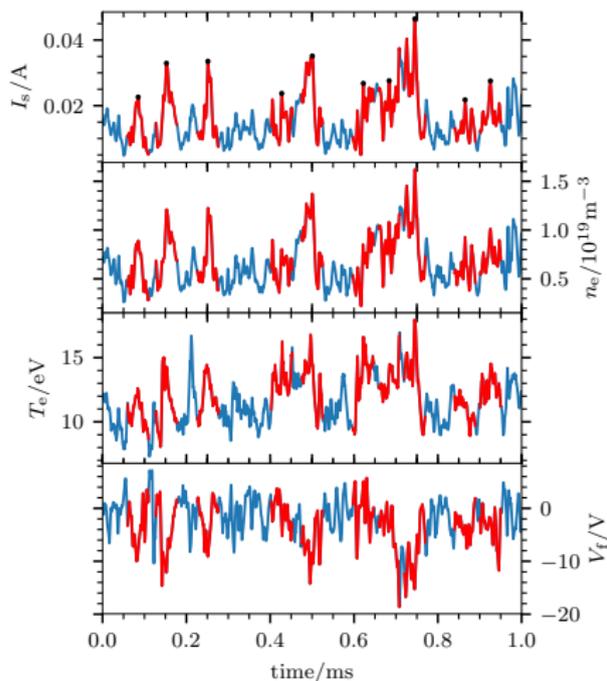
- Ohmic L-mode plasma
- Lower single-null magnetic geometry
- Density varied from  $\bar{n}_e/n_G = 0.12..0.62$
- Probe head dwelled at the limiter radius
- 4 electrodes with Mirror Langmuir probes
- Approximately 1s long data time series in steady state

# Mirror Langmuir Probe allows fast $I_s$ , $T_e$ , and $V_f$ sampling



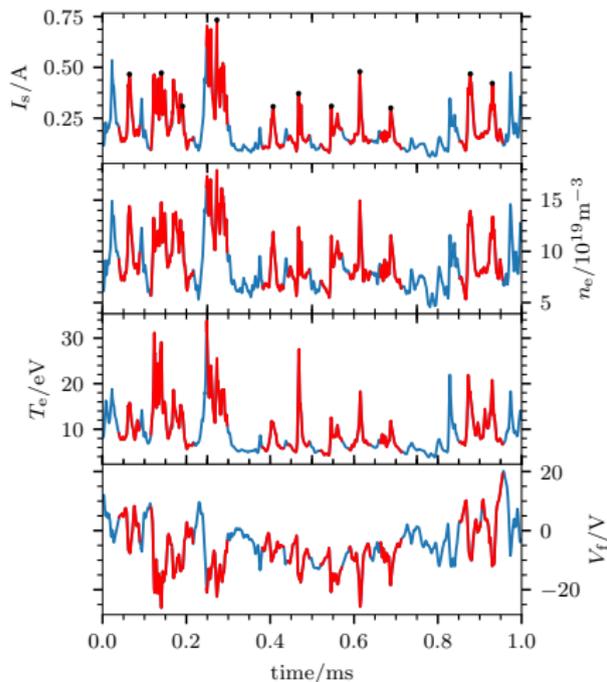
- MLP biases electrode to 3 voltages per microsecond.
- Voltage range is dynamically adjusted
- Probe current measured in each voltage state
- Fit input voltage and current is subject to 12pt smoothing (running average)
- Fit U-I characteristic on (U,I) samples
- Largest error on  $T_e$ .
- Resolves fluctuations on  $\mu\text{s}$  time scale

# Low density discharge, $\bar{n}_e/n_G = 0.12$



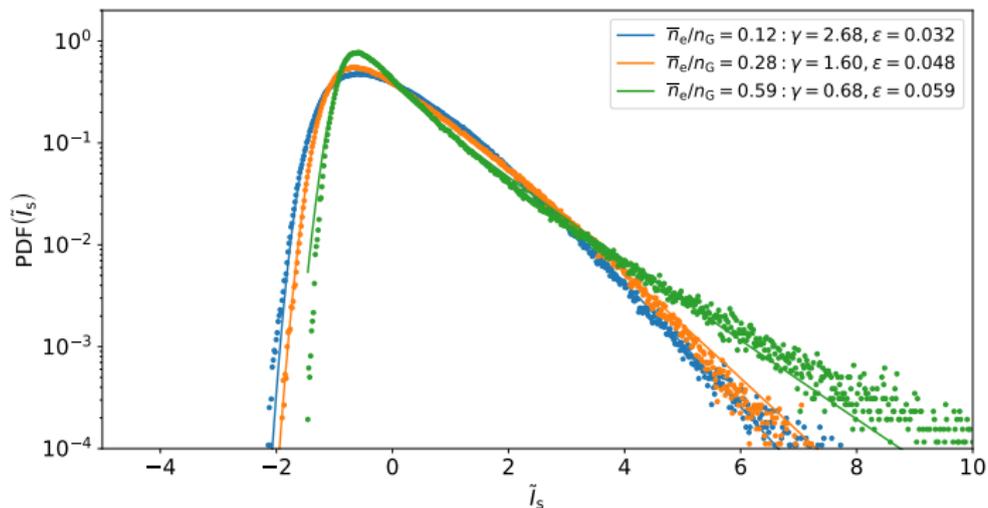
- Intermittent, large amplitude bursts in  $I_s$ .
- Bursts in  $n_e$  and  $T_e$  appear correlated
- Timescale approximately  $25\mu\text{s}$
- Irregular potential waveform

# High density discharge, $\bar{n}_e/n_G = 0.62$



- Bursts appear more isolated
- Average density larger by factor of 10
- Average electron temperature approx. 8eV

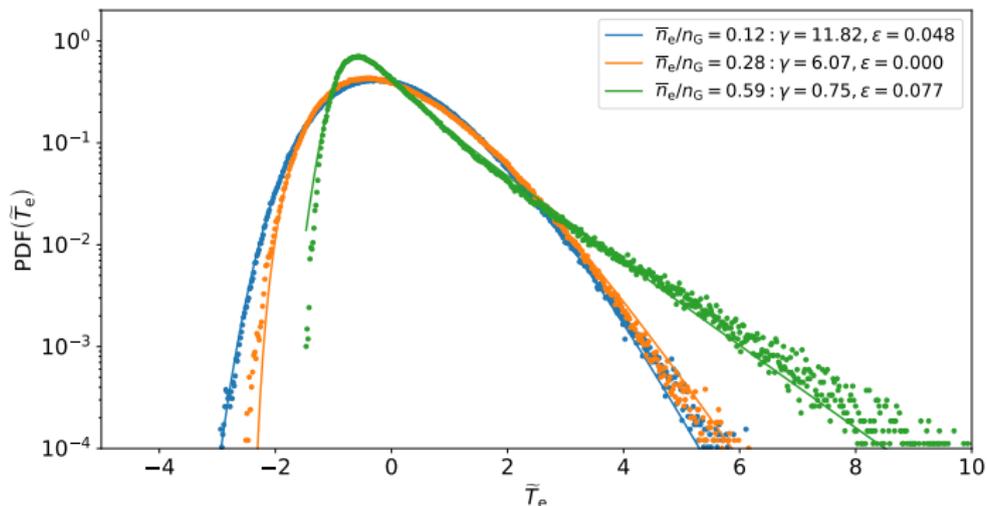
# Ion saturation current histograms are well described by a Gamma distribution



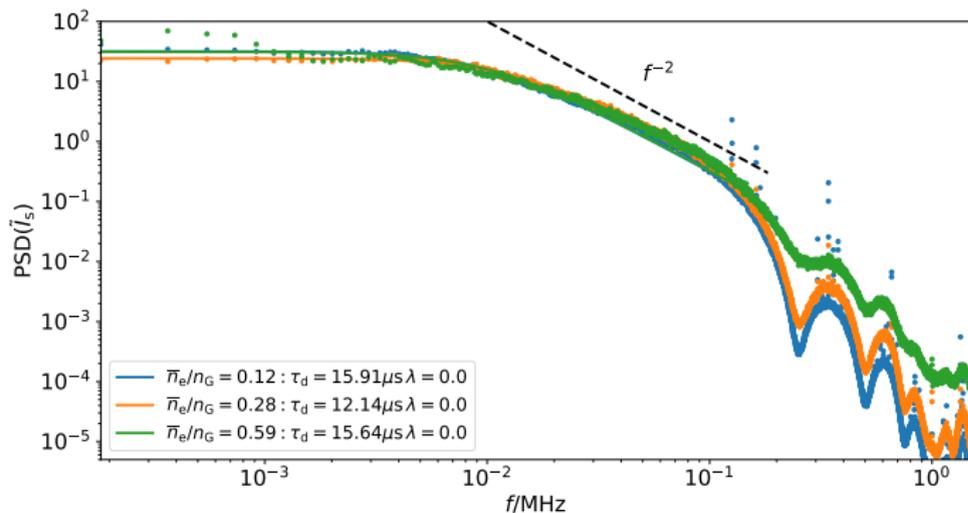
A.

Theodorsen, O.E. Garcia, and M. Rypdal, Phys. Scr. **92** 054002 (2017)

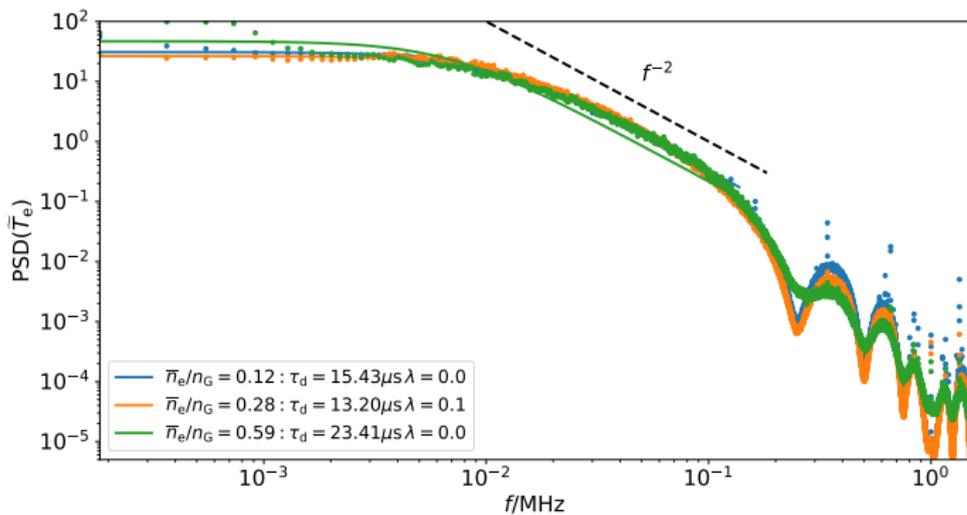
# Electron temperature histograms are well described by a Gamma distribution



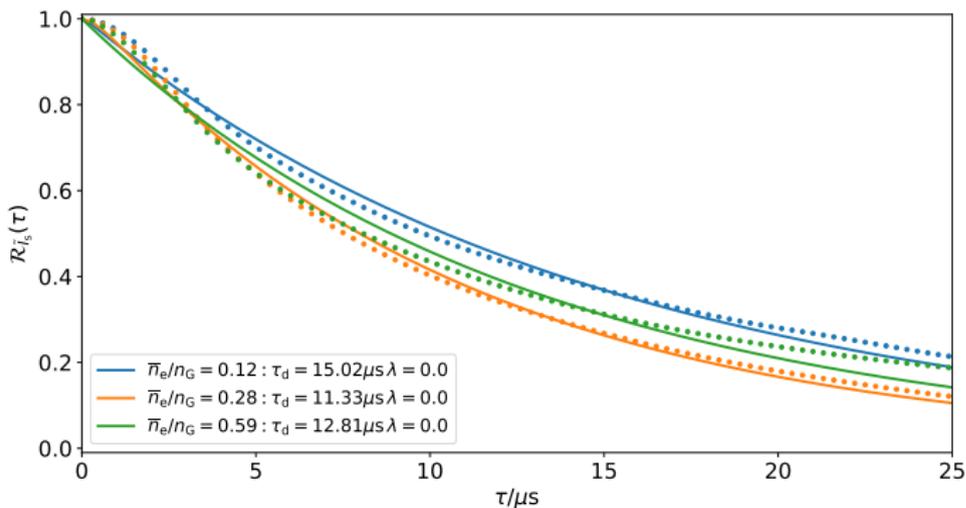
# PSD of $I_s$ shows broken power law



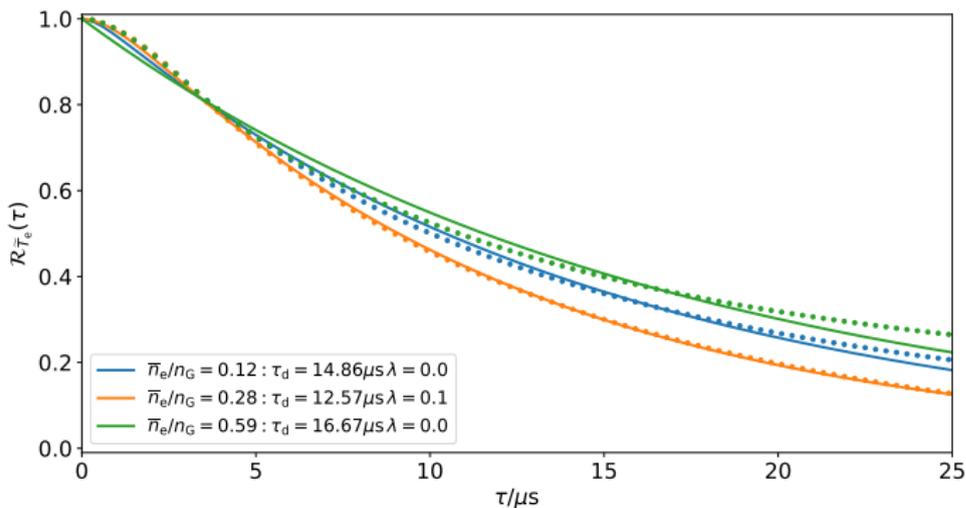
# PSD of $T_e$ shows broken power law



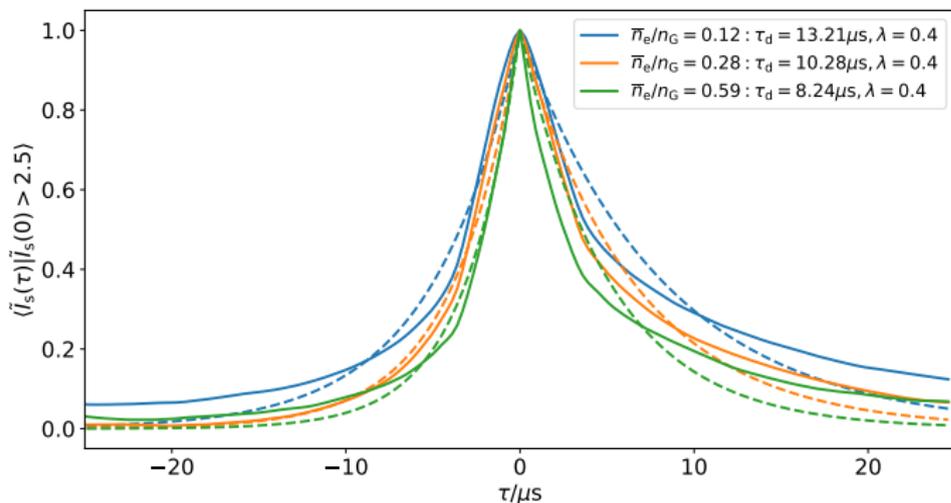
# $I_S$ shows exponential autocorrelation function



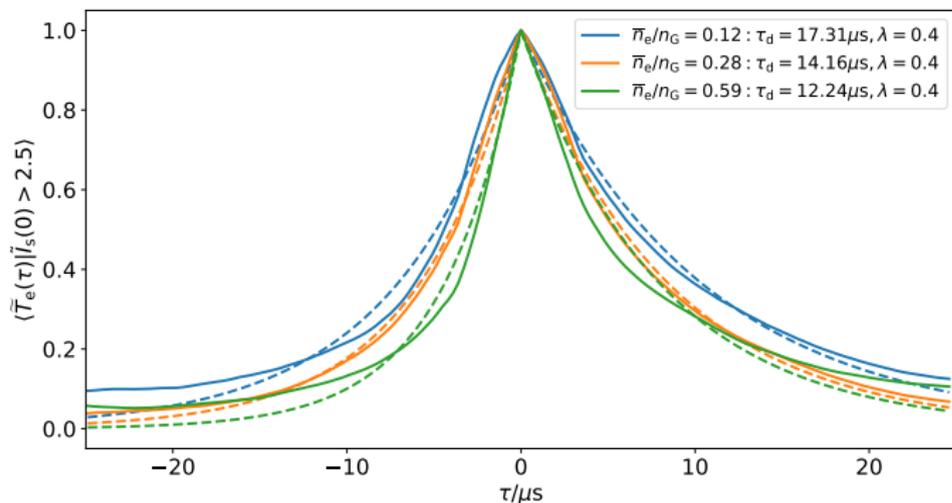
# $T_e$ shows exponential autocorrelation function



# Bursts in $I_S$ are approximated by double-exponential waveform

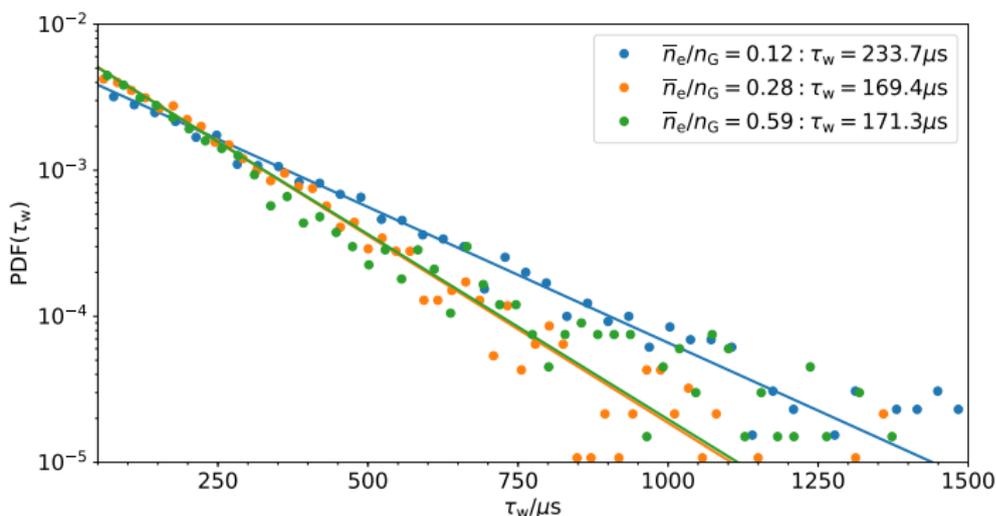


# Bursts in $T_e$ are approximated by double-exponential waveform

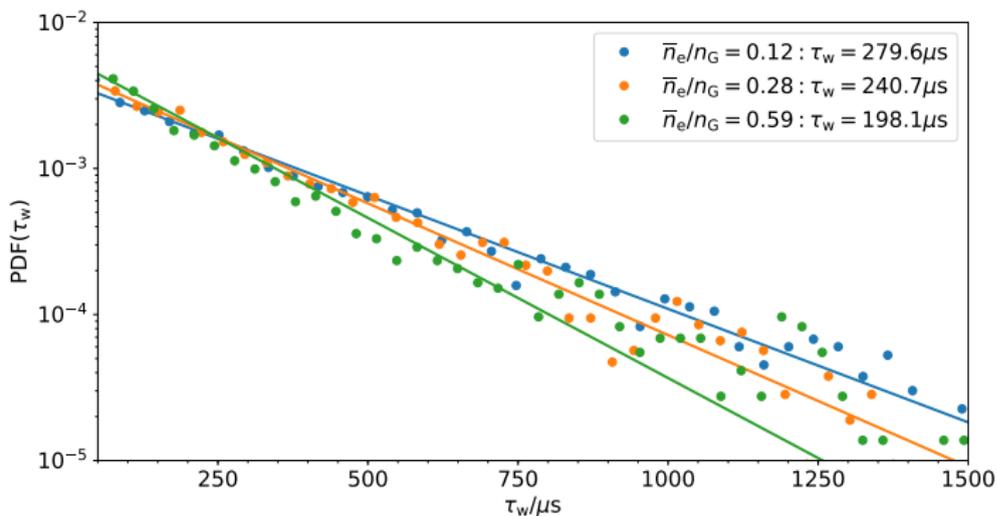


# Time between bursts in $I_s$ signal is exponentially distributed

Exponential distribution describes the time between events in a Poisson process.

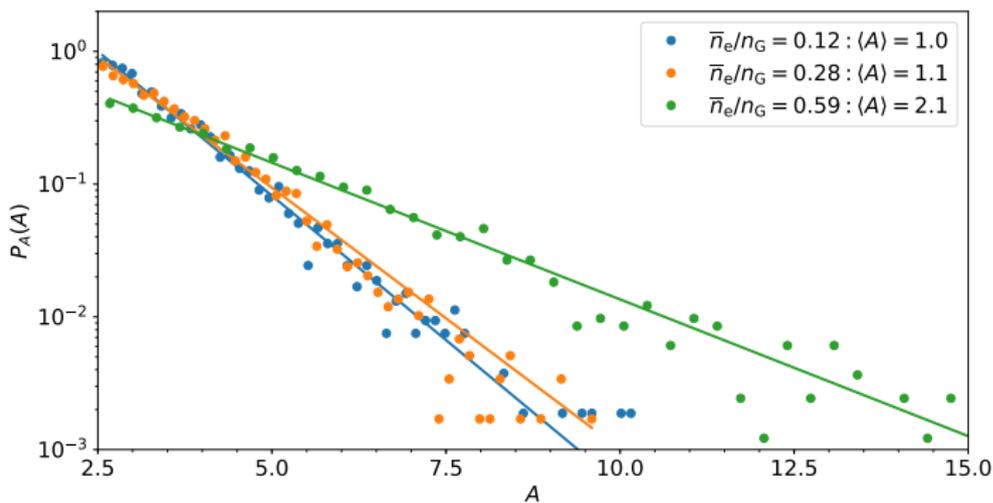


# Time between bursts in $T_e$ signal is exponentially distributed



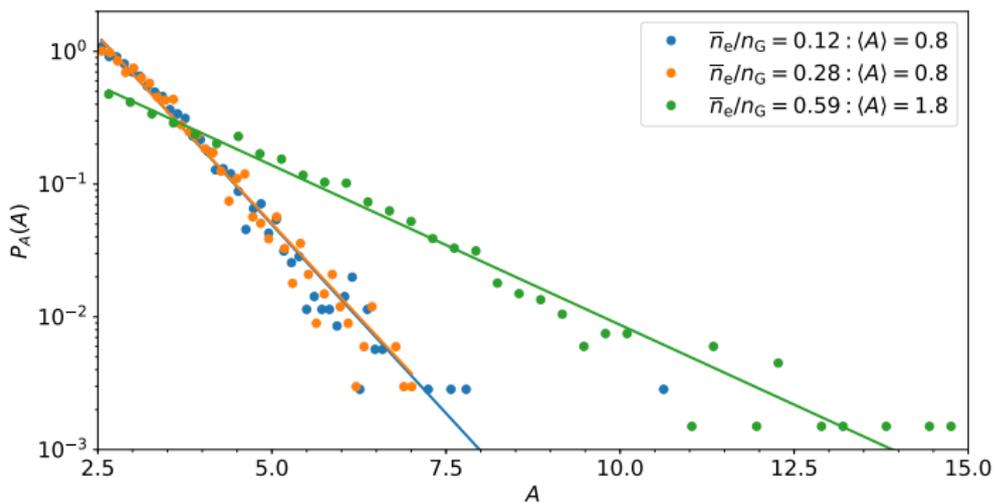
Advanced diagnostics for ITER: Diagnostic design

## Burst amplitude distribution - Isat



asp\_mfp\_000001\_tauwait\_burstamp\_scan.py

## Burst amplitude distribution - Te



asp\_mfp\_000001\_tauwait\_burstamp\_scan.py

# Conclusions

# Overview of estimated parameters

	$\frac{\bar{n}_e}{n_G}$	$\gamma$ (PDF)	$\gamma \left( \frac{\Phi_{\text{rms}}}{\langle \Phi \rangle} \right)$	$\tau_d$ (PSD)	$\tau_d, \mathcal{R}$	$\tau_d$ (CA)	$\tau_w$	$\langle A \rangle$
$I_s$	0.12	2.68	8.0	$15.0 \mu\text{s}$	$15.0 \mu\text{s}$	$13.2 \mu\text{s}$	$234 \mu\text{s}$	1.0
$I_s$	0.28	1.60	5.7	$12.1 \mu\text{s}$	$11.3 \mu\text{s}$	$10.3 \mu\text{s}$	$169 \mu\text{s}$	1.1
$I_s$	0.59	0.68	4.4	$15.6 \mu\text{s}$	$12.8 \mu\text{s}$	$8.24 \mu\text{s}$	$171 \mu\text{s}$	2.1
$T_e$	0.12	11.82	25	$15.4 \mu\text{s}$	$14.9 \mu\text{s}$	$17.3 \mu\text{s}$	$280 \mu\text{s}$	0.8
$T_e$	0.28	6.07	13	$13.2 \mu\text{s}$	$12.6 \mu\text{s}$	$14.2 \mu\text{s}$	$241 \mu\text{s}$	0.8
$T_e$	0.59	0.75	4.6	$23.4 \mu\text{s}$	$16.7 \mu\text{s}$	$12.2 \mu\text{s}$	$198 \mu\text{s}$	1.8

# Conclusions

Theory	Experimental data
Process is Gamma distributed	$I_s$ and $T_e$ time series are Gamma distributed
Pulses arrive uncorrelated	Waiting time between bursts in $I_s$ and $T_e$ is exponential distributed
Exponential distributed pulse amplitude	Burst amplitudes in $I_s$ and $T_e$ are expon. distributed
Double-exponential pulse shape	PSD, autocorrelation function and cond. avg. of $I_s$ and $T_e$ time series agree

- Less burst overlap at high densities
- Burst duration time changes little with  $\bar{n}_e/n_G$ .
- Burst amplitude increases with  $\bar{n}_e/n_G$

Thank you for your attention.