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Abstract: We study the amplitude modulation of the radial electric field constructed from the Langmuir probe plasma potential measurements at the edge of MAST. The Empirical Mode Decomposition technique is applied, which allows us to construct fluctuations on temporal scales of plasma turbulence, the Geodesic Acoustic Mode and these associated with the residual poloidal flows. This decomposition preserves nonlinear character of the signal. Hilbert transform is then used to obtain the amplitude modulation envelope of the fluctuating radial electric field on these time scales. We find significant spectral coherence at frequencies between 1 – 5 kHz, in the turbulence and the GAM amplitude modulation envelopes and for the signal representing the low frequency zonal flows. We find evidence of local and nonlocal three wave interactions leading to coupling between the GAM and the low frequency part of the spectrum.

Keywords: Fusion plasma; radial electric field; temporal intermittency

1. Introduction

The edge region of tokamaks, defined by the steep pressure gradient, is dominated by turbulent structures of density, temperature and the electrostatic potential (see, for example, [1] and the references therein) arising from resistive and/or interchange plasma instabilities. These fluctuations are responsible for the intermittent turbulent radial transport, which drives core heat and particle losses. Thus, better understanding and control of the edge transport is fundamental to enabling enhanced plasma confinement scenarios of future fusion reactors. The anisotropic shear amplification of micro-turbulence Reynolds stress produces radially localised, toroidally and poloidally symmetric flows, called zonal flows (ZF). These flows are distinct from the residual poloidal flows, often called zero frequency zonal flows (ZFZF) [2,3]. Shearing associated with both types of zonal flows can nonlinearly modify stability threshold of unstable plasma modes [4–6] and reduces turbulence level by vortex stretching [7,9].

Zonal flows are axisymmetric electrostatic potential modes with zero poloidal and toroidal numbers, $m = n = 0$. In tokamak geometry, toroidal curvature couples ZF to the density perturbations with poloidal mode numbers $m \geq 1$ ($n = 0$), and with a finite frequency. This compressible component of ZF is called Geodesic Acoustic Mode (GAM). The local dispersion relation for the GAM has been derived from various plasma models and the leading term is $\omega_{GF} \sim c_s / R_0$, where $c_s$ is the local sound speed and $R_0$ is the major radius [10,11]. The amplitude of the density fluctuations varies with the poloidal angle $\theta$ as $A \sim \sin(\theta)$. Since its theoretical discovery, the GAM has been experimentally...
observed in many tokamaks [12–16]. While the local theory predicts a monotonic change of the
observed GAM frequency with a radius, due to temperature gradient, there is growing evidence
that the GAM is a global mode with a complex radial mode structure [17–20]. Driven by turbulent
fluctuations [21], ZF and GAM provide a natural sink for turbulent energy. The stability of these flows
is less understood. The GAM is damped by a Landau mechanism, in the region of low safety factor
$q(r)$. The ZF/GAM can decay via nonlinear tertiary K-H instability of small scale fluctuations [5,21].
Nonlinear advection of GAM pressure perturbations provides a mechanism for the energy transport
from ZF back to micro-turbulence scales [22,23].

Nonlinear interaction of ZF/GAM with turbulence are fundamental to our understanding of the
L-H transition [8,24]. In this context, GAM is a valuable tool in the experimental studies of ZF due to
its finite frequency, which allows easier identification of ZF in experimental data [19]. Here, we study
the coupling of GAM/ZF and turbulence by examining the low frequency amplitude modulations of
the oscillating radial electric field component on different temporal scales. It has been observed before
that the power associated with the radial electric field oscillations at GAM frequency is not uniform
in time. Instead, it shows strong temporal intermittency, that is, the power is concentrated in few
intense temporal regions, separated by intervals of low level activity. We use Hilbert transform based
techniques to extract radial electric field fluctuations on different temporal scales while preserving
their nonlinear character. This allows us to construct signals representing meso-scale turbulence, the
oscillatory GAM signal and the low frequency zonal flows, $ZFZF$. Hilbert transform gives nonlinear
envelopes for the turbulence and the GAM. The spectral coherence of the turbulence with the GAM
is then examined. We find that the amplitude modulation of the turbulence and the GAM have a
similar behaviour at low frequencies, between $2 – 5 \text{ kHz}$. The auto bi-coherence reveals nonlinear
self-interaction of GAM and the possible coupling to these low frequency components.

Figure 1. (a) Langmuir Mach probe with separations of different pins in millimetres. Pins 1, 2 and 3 are
radially offset by 8mm from all other pins.

2. Experimental setup and data

The Mega Amp Spherical Tokamak (MAST) has a major radius $R_0 \approx 0.85 \text{ m}$ and a minor radius
of $a \approx 0.65 \text{ m}$. The magnetic field strength is about 0.5T with the toroidal, $B_T$, and the poloidal, $B_P$,
field components giving a pitch angle of about 22° at the edge of the device. We analyse data from an
Ohmic plasma discharge numbered 29150, with a line average number density, \( n \approx 1.47 \times 10^{19} \text{ m}^{-3} \),
and plasma current \( I_p = 0.43 \text{MA} \). No additional heating power was applied during the discharge.
Magnetic configuration was a double null.

The data was collected using a Mach type reciprocating Langmuir probe [25], on the outboard
mid-plane, measuring floating potential, \( \tilde{V}_f \) as well as a set of ion saturation currents (pins 2, 5 and
8). Figure 1 shows the schematic of the probe, with pin numbers and the relative distances between
them. Pins (1,3) are positioned 0.8 cm behind pin pairs (4,6) and (7,9). We assume that the floating
potential \( \tilde{V}_f \) is a good proxy of the plasma potential \( \tilde{V}_p \). These are related by \( \tilde{V}_p = \tilde{V}_f + \Lambda \), where \( \Lambda \)
is the sheath potential drop, which is a slowly varying function of the electron and ion temperatures
and is usually approximated by \( \Lambda \approx 2.5 T_e/e \) [26]. It is assumed that the electrostatic potential
fluctuations are larger then these due to electron/ion temperature fluctuations. The sheath potential
of the Mach barrier can modify plasma flows as well as the electron and the ion temperatures, but
the measurement of \( \tilde{V}_f \) is based on the ion and electron currents balance, which in principle should
depend only on the plasma temperature. The high values of temporal correlations on all pin pairs,
with correlation coefficients between 0.65 and 0.95 are consistent with these assumptions. We construct
radial electric field component \( E_r \) by differencing the floating potential values on pin pair (1,9),
\( E_r = \nabla_r \tilde{V}_p \approx (\tilde{V}_p^{(1)} - \tilde{V}_p^{(9)})/d_r^{(1,9)} \), where the superscripts on the floating potential indicates a pin and
\( d_r^{(1,9)} = 0.8 \text{ cm} \) is a radial separation of these pins. We note that the poloidal separation of the pin
pair (1,9), \( d_\theta^{(1,9)} = 3.8 \text{ cm} \), is much larger than the radial separation. This minimises the impact of
large poloidal wave numbers, associated with turbulence, on the radial electric field estimation. We
have chosen time interval of 0.315 – 0.33, during which the intermittent character of the fluctuations is
clearly present.

Figure 2 presents a summary of the data. Panel (a) shows the time series of the radial electric field
\( E_r(t) \), containing approximately 7500 samples. Assuming that the toroidal magnetic field is dominant,
the radial electric field gives the poloidal flow speed \( v_p = (E_r B_\phi)/B^2 \). Panel (b) shows the distance of
the Mach probe in relation to the last closed flux surface (LCFS) during this time interval. The probe is
inside the plasma its radial depth varies between ~4 and ~5 cm. Panel (c) of Figure 2 shows electron
temperature from Thomson scattering diagnostic at the time of interest. It gives the median of electron
temperature at the probe location of \( T_e \approx 14 \text{ eV} \). We take the electron temperature at \( T_e = 10 \text{ eV} \) in all
calculations that follow. The proton gyroradius at this electron temperature is \( \rho_p = 0.15 \text{ cm} \).

3. Methods

The methods based on the Hilbert-Huang transform (HHT) offer a natural approach to studying
temporal intermittency in a time series. The intermittency is interpreted as an amplitude modulation
of a mode, or a group of modes, of interest. Here, we employ this method as an effective filter, which
allows us to construct signals representing different dynamical temporal scales. The HHT performs
well when analysing non-stationary and non-harmonic fluctuations arising in nonlinear systems.
Fourier-based spectral techniques as well as the wavelet transform are unsuitable for such time series
if the principal aim is to preserve the nonlinear nature of the wave trains. The HHT makes use of the
Empirical Mode Decomposition (EMD) [27], which expands the input signal onto a set of intrinsic
mode functions (IMFs) derived directly from the data.

In practice, the iterative sifting process is performed as follows: firstly, the maxima and minima of
the signal are separately connected using cubic splines to form two envelopes of the data; one that
contains all of the maxima and the other, the minima. The mean of the maximum and minimum
envelopes, \( m_1 \), is calculated. For an input signal \( S(t) \), the difference, \( h_1 = S(t) - m_1 \) gives the first estimation of the envelope of \( S(t) \). However, this envelope’s mean is, in general, not equal to the
true local mean, especially if the data is nonlinear. The process is therefore repeated \( k \) times until the
resultant, \( h_{1k} \), satisfies the requirement for an IMF, \( h_{1(k-1)} - m_{1k} = h_{1k} \), where \( h_{1k} \) and \( m_{1k} \) are the \( k \)th
Figure 2. Summary of the data. (a) Wavelet power at lower frequencies. (b) Mach probe distance from the last closed flux surface (LCFS); negative values indicate position inside the plasma. Horizontal red lines mark three time intervals analysed here. (c) Electron temperature profiles from Thompson scattering system between 0.32 – 0.36 seconds.

envelope and its mean after kth sifting iteration, accordingly. We then designate $s_1(t) = h_{1k}$ as the first IMF component of the data, containing the shortest period of the signal. Fluctuations at this scale are removed from the data to obtain a residual $r_1 = S(t) - s_1(t)$. The procedure is then repeated for the residual $r_1$, treated as a new input signal. The decomposition is stopped either when the component $s_i$, or the residue $r_i$, become too small to be of interest, or when the residue, $r_i$, becomes a monotonic function from which no more IMFs can be extracted. For data with a trend, the final residue should be that trend. When the process is finished, we obtain the decomposition of a signal $S(t)$ into IMFs $s_i$ and the final residue:

$$S(t) = \sum_{i=1}^{N} s_i(t) + r_N. \quad (1)$$

The IMFs may contain oscillations with different periods in one mode, and different modes can contain similar periods. This spectral leakage, or mode mixing, can be an issue, especially for short and intermittent data. We incorporate the ensemble empirical mode decomposition (EEMD) \cite{28,29} to reduce the impact of mode mixing. This noise-assisted method adds white noise to the original data before the sifting process starts. The EMD modes are computed as normal until all of the IMFs are calculated. The original data is then reprocessed with a different noise realisation and the final IMF is averaged over all ensembles. In this work we use EEMD to decompose the radial electric field time series into a number of IMFs. We are interested in the amplitude modulation of turbulence and the
Figure 3. (a) Wavelet dynamic spectrum of $E_r(t)$. (b) Integrated wavelet power spectrum (black) and Fourier spectrum estimate (red). A significant spectral peak at $\sim 10$ kHz is clearly seen. Fourier spectrum shows an internal structure of the GAM peak, with multiple modes separated by $\sim 1$ kHz. A possible second harmonic is also present at $\sim 20$ kHz.

Geodesic Acoustic Mode (GAM). The modes with the periods shorter than the GAM are interpreted as turbulence. An envelope of a modulated signal can be constructed using analytic signal $S_a$:

$$S_a(t) = S(t) + i\mathcal{H}[S(t)] = E(t) \exp[i\phi(t)],$$

where $\mathcal{H}[S(t)]$ indicates Hilbert transform of the signal $S$, and $i^2 = -1$. For a slowly modulated signal the modulus of $S_a$ corresponds to the amplitude modulation envelope $E(t)$. The frequency of $S_a$, which may not be constant in time, can be obtained from the mean of the instantaneous phase change $f = \langle d\phi/dt \rangle$.

In order to quantify the nonlinear interactions between different modes we use the wavelet bi-coherence defined as

$$b^2(f_1, f_2) = \frac{|\langle \hat{S}(f_1, \tau) \hat{S}^*(f_2, \tau) \hat{S}^*(f_1 + f_2, \tau) \rangle|^2}{\langle |\hat{S}(f_1, \tau)\hat{S}^*(f_2, \tau)|^2 \rangle \langle |\hat{S}(f_1 + f_2, \tau)|^2 \rangle},$$

where $\hat{S}$ is a wavelet coefficient at a scale associated with a period $1/f$ and at time $\tau$. For a signal $S(t)$, the wavelet coefficients are given by

$$\hat{S}(s, \tau) = \int_{-\infty}^{\infty} dt S(t) \frac{1}{\sqrt{s}} \psi^*\left(\frac{t - \tau}{s}\right),$$

where $s$ is a temporal scale, $\tau$ is a new time label and $\psi(t)$ is the analysing wavelet. We use Bump wavelets [30], which have better frequency resolution, but poorer time localisation compared with a standard Morlet wavelet. Given a set of wavelet coefficients $\hat{S}(f, \tau)$ the wavelet spectrum estimate is given by

$$P(f) = \langle \hat{S}^*(v, \tau) \hat{S}(v, \tau) \rangle_{\tau}.$$
Wavelet estimates of the bi-coherence are superior to those obtained from the Fourier transform for shorter and non-stationary data sets. Fourier-based bi-coherence requires the averaging over many realisations of the same data, while the averaging indicating by $\langle \ldots \rangle$ in (3) is over time.

![Figure 4](image)

**Figure 4.** Signals constructed from EMMD produced IMFs: turbulence $S_T$ (lower, black), the GAM $S_G$ (upper black) low frequency zonal flows $S_Z$ (blue). Envelopes of turbulence $E_T$ (red) and of the GAM (green) obtained from the analytic signal method. Both, the GAM and ZFZF were offset vertically for clarity.

4. Results

A strong oscillatory component has been previously identified in MAST edge plasma density and electrostatic potential fluctuations measured by the reciprocating Langmuir probes [15,20]. This mode shows a spectral power peak at the frequency of $\sim 10$ kHz, in a reasonably good agreement with the theoretical and numerical predictions for GAM frequency in MAST L-mode edge plasma [15,16]. Panel (a) of figure 3 shows the wavelet transform dynamic spectrum for the radial electric field. The intermittent series of power maxima are clearly visible around the predicted GAM frequency of $\sim 10$ kHz. The integrated wavelet spectrum of electric field fluctuations is shown in the panel (b) of the same figure. The spectral peak at $\sim 10$ kHz is approximately 4 times above the power level of turbulent fluctuations at neighbouring frequencies. We note an apparent second harmonic peak at $\sim 20$ kHz and a complex number of smaller maxima at frequencies below $\sim 5$ kHz. We also show the Fourier power spectrum estimation for the same signal in the red trace, which reveals multiple peaks within a single broad spectral peak of a wavelet spectrum estimate. These additional spectral peaks in the Fourier-based spectrum are separated from the main peak by no more than 1 kHz. Allowing non-linear interactions between the modes represented by the peaks clustering around $\sim 10$ kHz, could lead to low-frequency modulation in the electric field signal.

In order to study the GAM amplitude modulation and its possible impact on turbulence and ZFs, we decompose our data into three components with distinct temporal scales: turbulence, GAM and low frequency zonal flow (ZFZF). We use the EEMD technique to generate 25 IMFs from the original, $E_r$ data, with the largest frequency of about $\sim 115$ kHz and the smallest frequency at $\sim 70$ Hz. The largest frequency is treated as the residual noise in the data and is discarded. Similarly, we discard he
smallest frequency mode, which is a nonlinear trend very close to zero. We then combine the IMFs at
three different frequency ranges to obtain signals of interest. The turbulence, $S_T(t)$, is a superposition
of IMF 2 – 4, corresponding to mean instantaneous frequencies between $\sim 25$ kHz and $\sim 66$ kHz. The
GAM is a single IMF with the mean frequency of 10 kHz, $S_G(t)$. Finally, the ZFZF signal $S_Z(t)$ is
obtained by summing modes 8 – 20, with mean frequencies in the range 77 – 5000 Hz. We apply
an analytic signal approach to $S_T$ and $S_G$, in order to obtain their amplitude modulation envelopes
$E_T$ and $E_G$, respectively. The turbulence envelope is smoothed over 50 neighbouring points. Figure
4 shows signals and their upper envelopes. The GAM, its envelope and the ZFZF signal have been
shifted vertically and their amplitudes were modified for clarity. A close visual inspection is sufficient
to see that there is no phase coherence between turbulence envelope, GAM envelope and the ZFZF
signal. This is confirmed by a linear cross correlation coefficients calculated for the pairs ($E_T, E_G$),
($E_T, S_Z$) and ($E_G, S_Z$), which had values at around or below 0.25, at different non-zero time lags.

While there is no phase coherence in the low-frequency behaviour of $E_T$, $E_G$ and $S_Z$, there is a
considerable spectral coherence for these time traces. Figure 5 shows Fourier spectral estimate of power
for $E_T$, $E_G$ and $S_Z$ in panels (a), (b) and (c) respectively. Note that all spectra have been normalised to
their respective maxima, which had a value of 0.05 for turbulence envelope, 0.28 for the GAM envelop
and 0.008 for ZFZF signal. All spectra have pronounced peaks at about 1 kHz and at 2 kHz. The GAM
envelope spectrum shows a number of peaks at relatively constant increments, positioned at $\sim 0.55$,
$\sim 0.75$, 1 and 1.5 kHz. We also note that, the ZFZF spectrum shows a broad spectral power between
frequencies 3 – 5 kHz. This is a strong suggestion that there are nonlinear interactions between various
modes, close to the GAM frequency, and also non-local between GAM/turbulence and ZFZF.

The resonant tree wave interaction process is considered as a model of coupling between different
modes present in the radial electric field time series. For a single point time series measurements,
we can only consider frequency resonances $f_3 = f_1 + f_2$. The strength of these interactions is then
approximated by a bi-coherence, which we have calculated using wavelet coefficients and averaged
over all times. Figure 6 shows only positive frequency part of the bi-coherence, which was thresholded
at a relatively high value of 0.7 to emphasise the most relevant interactions. We find the signature
of strong local interactions at the GAM frequencies $f_1 \approx f_2 \approx 10$ kHz, positioned on the diagonal
line, as well as non-local interactions with the low frequency modes such as $f_1 = 8.4$ kHz, $f_2 = 1.2$
kHz. Interestingly, the bi-coherence also reveals the importance of a mode with $f \approx 4$ kHz, which also
self-interacts and couples to low frequency modes.

5. Discussion

We have performed the analysis of the fluctuations in the radial electric field component obtained
from the reciprocating Langmuir Mach probe at the mid plane of MAST. Wavelet dynamic spectrum
reveals temporal power intermittency of the GAM, while Fourier spectrum estimate revealed multiple
spectral peaks in the vicinity of 10 kHz. We use Hilbert-Huang transform based technique, EEMD,
to extract radial electric field fluctuations on temporal scales of turbulence, the GAM and ZFZF.
Envelopes of turbulent signal and the GAM were constructed using analytic signal approach. We
found a significant spectral coherence for the turbulent envelope, GAM envelope and the low frequency
component. The bi-coherence revealed strong nonlinear interactions, local self-interactions near the
GAM frequency [31], and non-local interactions with low frequency mediated by the GAM. This is
broadly consistent with previous results presented in the literature [19,32].

Our findings may be of particular importance for better understanding how the presence of the
GAM alters the physics of L-H transition. It has been reported that the low frequency component,
which we have termed ZFZF in this work, increases significantly at the expense of the GAM during
the transition [24]. The energy flow for the three component system, turbulence, GAM and ZF, is often
modelled using nonlinear predator-prey type formulation [2,33]. These models incorporate all key
physical interactions important for the dynamics of the system, but retain simplicity that allows better
understanding of how each component effects their collective complex dynamics, for example, the L-H
transition. The behaviour of these models is strongly influenced by the included interactions between various components. This work clearly shows that in addition to the linear impact of the GAM on ZFs [34], the nonlinear interactions are also important.

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**References**


Figure 6. The wavelet bi-coherence for the entire data set $E_r$. 


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