Order statistics and extremal properties of spatially smoothed laser beams in laser-plasma interaction

(short title: Order statistics of intense laser hot spots)

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Abstract
The order statistics of intense speckles or ‘laser hot spots’ are studied in the context of so-called “optically smoothed” light beams of laser-matter interaction. We investigate theoretically and by means of numerical simulations the distribution function for the \(k\)-th most intense speckle maxima in the upper tail speckle distribution. From these distributions for each order \(k\), a distribution functions for the intense speckles as a function of their peak intensity can be established, which allows to compute their impact on nonlinear processes, like parametric instabilities. This is done for the example of Stimulated Brillouin Scattering, using the so-called \textit{independent hot spot model}, for which the backscatter reflectivity level is computed, which proves to be in very good agreements with numerical simulations. This result is of great interest for nonlinear processes, like instabilities, where extreme speckles play an important role.

Order Statistics, Laser-Plasma Interaction, Extreme Theory, Statistical Optics, Speckles

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1 Introduction

We study the extremal properties of so-called speckle patterns of optically smoothed laser beams, in the context of laser-plasma interaction. Laser speckles are the small-scale structures in the intensity pattern in the focal volume of the laser beam. For a certain class of smoothing techniques, these speckles are coherent structures of homogeneous size with statistically distributed peak intensities. The most frequently used method for spatial smoothing is the so-called ‘Random Phase Plate’ (“RPP”) technique, which produces on the focal plane an intensity pattern consisting of a large number of small-size "speckles", also called 'hot spots'. (Ohtsubo & Asakura 1977), (Kato et al., 1984), (Obenschain et al., 1986).

We aim to describe the statistical properties of the most intense speckles. These extremal properties, in particular the fluctuation of intensity maxima are an important issue for nonlinear mechanisms which affect the beam propagation, like filamentation, or scattering instabilities (see e.g. Ref. (Rose & DuBois, 1994), (Hüller et al., 1998, Mounaix et al., 2000)).

Such instabilities are particularly sensitive to the features of the ‘upper tail’ of the speckle distribution function, in particular if they follow threshold like behaviour.

As an illustration of the relevance of these extremal properties ,we discuss the exemple of the reflectivity value of a parametric instability generated by SBS.

A typical “speckle distribution function” can be characterized by the “body”, namely the interval in the speckle intensity where the probability density for speckles peaks, and by the “tail” of the most intense speckles. The body of the distribution, located around the mean speckle intensity $I$, (Goodman 1984 & 1985), (Rose & DuBois, 1993a) is generally a well-known function, which can in the case of laser-smoothing easily be deduced from the experimental technique applied (Ohtsubo & Asakura 1977), (Kato et al., 1984), (Obenschain et al., 1986), (Goodman 1984 & 1985), (Biancalana & Chessa, 1994). The “body” of the speckle distribution function (DF) depends only very weakly on the quality of the light beam incident to the RPP; i. e. whether the beam is close to an ideal beam (e.g. a plane wave) or whether imperfections in the optical path and/or during the amplification deform the intensity profile and the wave phase front. The “tail” of the distribution, however, contains only a small number of intense speckles (for a finite-size speckle pattern), and hence, statistically, depends delicately on the beam quality and on the information in the RPP mask. The tail may therefore exhibit considerable fluctuations with respect to changes in the phase mask and/or in the optical path.
Theoretical modeling of laser-plasma interaction experiments under conditions, relevant to laser fusion, had been very difficult before single hot-spot experiments providing well-defined laser beam conditions were achieved at the end of the 1990’s (Montgomery et al., 1999), (Kline & Montgomery, 2005), with the exception of regimes with very strong beam self-focusing (Malka et al., 2000). ‘Generic’ high-power laser beams, without any sophisticated phase front correction, did not allow to incorporate reliably the necessary phase information in theoretical models. In contrast to this, the application of theoretical models incorporating optical smoothing methods, see Rose & DuBois, (1993a), has already been very successful earlier, see e.g. in work by Tikhonchuk et al. (1996), Tikhonchuk et al., (1997), Hüller et al. (1998), Mounaix et al. (2000), Myatt et al. (2001), Tikhonchuk et al. (2001), Pesme et al. (2002), Weber et al. (2004). Masson et al. (2006). The applicability of optically smoothed beams in laser-plasma theory is a consequence of the fact that each of the numerous speckles is no longer affected by imperfections originating from the laser itself. Ideally, the phase information of a speckle can be assumed to be known. From this point of view, it is relatively irrelevant to the statistics and modeling of the speckle pattern in the focal plane, whether the ‘generic’ laser beam, before entering into smoothing optical element, is well controlled in its phase or not. This is the reason why good agreement between theory and experimental results has been found using RPP-smoothed beams where the physics was governed by contributions from speckles in the body of the DF.

For processes, however, with critical dependence on high intensities (Rose & DuBois, 1994), potentially situated in the tail of the speckle DF, theoretical modeling is often difficult because of the incomplete information from experiments. A statistical treatment of the fluctuations in the tail of the speckle DF is therefore necessary, and the study of extremal properties in the tail of the DF is very valuable for such critical processes.

As we have shown in recent work (Porzio & Hüller, 2010), the statistics of the most intense speckle in a speckle pattern of a smoothed laser beam is essentially governed by the ratio between the focal volume $V_{\text{foc}}$, in which the considered speckle pattern is defined, and the specific volume of a single speckle, $V_{\text{sp}}$. For the method of RPP beams, the specific speckle volume is, in principle, the same for all speckles, so that the ratio $V_{\text{foc}}/V_{\text{sp}}$ simply indicates the number of potential speckles in $V_{\text{foc}}$. While there are other smoothing techniques where this feature does not necessarily apply, we restrict ourselves in this work to the case of a unique size/volume of speckles.

The paper is organized as follows: In the following section 2 we develop the theory for the most intense
speckle and the successive orders. These results are in good agreement with numerical simulations, which also confirm that the spacing in intensity between successive orders can be used to define a histogram for the speckle maxima abundance of the most intense speckles. By means of numerical simulations using the properties of optically smoothed beams, we show in section 3 that the theoretical results on order statistics allow to compute the impact of the most intense speckles to critical nonlinear processes like stimulated scattering. We discuss the limit of the model and aspects going beyond the scope of this work in the concluding section 4.

2 Order statistics

While the “body” of the speckle distribution function shows almost no fluctuations from one realisation to another, the most intense speckles in a pattern show fluctuations in their peak intensity value from one to another realisation for which one wants to quantify the relevant intensity interval. Let us explain the meaning of ‘realisation’ in this context: a realisation of a speckle pattern corresponds either to a numerically generated pattern by simulations, or by a speckle pattern associated with an experimental setup that may vary the properties in the optical path, and which cannot be reproduced identically from one laser shot to another.

It is intuitive that the fluctuations of intense speckles should decrease successively and inversely with the order (the most intense being the first maximum). For a speckle field pattern of a smoothed laser beam, it is however interesting to investigate the joint occurrence of the first $k$ successive maxima following the most intense one.

We do this in the following by looking for the asymptotic behaviour of the joint probability distribution of the most intense, the second, up to the $k$-th maximum, each maximum – this has to be underlined – being the maximum of a number of local intensities measured in a box approximately of the size of a speckle.

The method used here to determine the successive orders $M_1 > \ldots > M_{k-1} > M_k > M_{k+1}$ of maxima consists in dividing the considered focal volume $V_{foc}$ in a number of $n_{sp}$ sub-cells and to find the maxima in each sub-cell. The number $n_{sp}$ indicates the number of potential speckles to be found in the considered volume.

Note that to order the maxima $M_k$ does not mean that we are ordering the strict intensity maxima
of the field! Ordering the maxima of different boxes has the meaning of ordering the power contained in different speckles. Of course, in the pure “Gaussian” case, where individual speckles do not exhibit secondary peaks, these two approaches, namely ordering the maxima and ordering the intensity peaks of the field, coincide.

For this reason the correlation properties inside an individual speckle, depending on the method generating the pattern, are important. Several techniques to generate RPP speckle patterns are known, each giving rise to more or less oscillatory structure of the speckle field, both along and across the propagation axis. However, we are interested in the speckle maxima so that subsidiary maxima of the same speckle have to be excluded when considering speckle statistics. For this reason the characteristic dimensions of speckles have to be determined from the speckle correlation functions, as done in Appendix B. An unsuitable choice of typical speckle dimensions can henceforth lead to the wrong number of the $V_{\text{foc}}/V_{\text{sp}}$ ratio. However, the knowledge of the correct value $V_{\text{foc}}/V_{\text{sp}}$ is of particular importance, as we see in the following, to establish an order statistics of the most intense speckles, because one uses a method to subdivide the speckle pattern in sub-volumes of the size of an individual speckle.

### 2.1 The principal maximum $M_1$

Before determining the order statistics, we first recall the notion for the extremal statistics of the highest intensity of a speckle field, which is the 1st order maximum $M_1$ (Embrechts et al., 1997), (Porzio & Hüller, 2010): let us denote with $F()$ the common probability distribution function for the sequence $\{I\}_n \equiv I_1, ..., I_n$ of intensity values in a speckle pattern. Then $F(u_n)$ is the probability $P()$ that an intensity value $I_j$, being one among the sequence $\{I\}_n$ of $n$ known intensity values, can be found below the level $u_n$, namely $P(I_j < u_n)$. Using that $F(u_n)$ is common for all values in $\{I\}_n$, the mean number of exceedences of the level $u_n$ by all $n$ values in $\{I\}_n$ is given by $n(1 - F(u_n))$. If the value of $n(1 - F(u_n))$ stabilizes toward a positive real value $\tau$ for $n \to \infty$, a limit law exists that can describe the extremal properties of speckle maxima (Embrechts et al., 1997). We first assume that this is the case, $\lim_{n \to \infty} n(1 - F(u_n)) = \tau$, and approximate $F(u_n)$ by $F(u_n) = 1 - \tau/n + o(1/n)$ with a residual error $o(1/n)$ converging faster than $1/n$. The latter allows to calculate the probability that the maximum $M_1$ of all $n$ values of $\{I\}_n$ stays below $u_n$, which is given by

$$P(M_1 < u_n) = \prod_{j=1}^{n} F(u_n) = (F(u_n))^n,$$
under the condition that all values of \( \{I\}_n \) are statistically independent. With the above approximation for \( F(u_n) \) one thus obtains \( P(M_1 < u_n) \to e^{-\tau} \) for \( n \to \infty \). In the case of speckle fields generated by RPP smoothing techniques, the probability distribution is generally an exponential, such that \( F(u) \equiv 1 - \exp(-u) \), with \( u \) being the intensity normalized to the mean intensity. Therefore the probability that the maximum \( M_1 \) is found below the level \( u_n \) finally reads

\[
P(M_1 < u_n) = \exp\{-ne^{-u_n}\} = \exp\{-e^{-u_n} - \ln n\} ,
\]

with \( \tau \to n(1 - F(u_n)) = ne^{-u_n} \). This is a double exponential distribution, identical to the so-called Gumbel distribution with the transform \( u_n = a_n + b n x \) and the scaling factors \( a_n = b_n \ln n \) and \( b_n = \langle I \rangle \), the latter being \( \langle I \rangle \equiv 1 \) for intensities normalized to the mean value. The above mentioned condition for the stabilisation toward a limit law \( H() \) can only be met if this transform between a finite value \( n \) and \( n \to \infty \) exists such that \( H(u_n) \equiv \lim_{n \to \infty} P(M_1 < u_n) \) is equivalent to \( H(x) \equiv \lim_{n \to \infty} P([M_1 - a_n]/b_n < x) = \exp(-e^{-x}) \). The transform between the level \( x \) and \( u_n \) corresponds to a translation of the probability distribution in the intensity axis, which has an important physical meaning.

The Gumbel limit law, as deduced above, strictly applies only to independent values in the set \( \{I\}_n = I_1 \ldots I_n \). If among the \( n \) values only a portion \( \theta < 1 \) meets this criterion, the limit law reads

\[
H^\theta(x) = \exp\{-\theta \exp(-x)\} .
\]

This means that a fraction \( 1 - \theta \) of the evaluated points is “clustered”, so that the shift of the distribution is limited to \( \ln(n/\theta) \) instead of \( \ln n \) for a completely unclustered set \( \{I\}_n \) (which is usually the case for a resolution with a number of points \( n \) smaller than the number of potential speckles). To determine the intensity maximum of a speckle field, the number of points required is naturally considerably higher than the number of potentially independent values, which is in fact the number of speckles \( n_{sp} \) in a volume. The shift has consequently a strict physical meaning, and the probability distribution centers around the normalized intensity equal to \( \ln(n/\theta) = \ln n_{sp} \). This is intuitive because the more speckles are found in a volume, the the more likely it is to find a high-intensity maximum in the pattern.

The number \( n_{sp} \) is an important parameter, also in what follows later. It depends on the specific speckle volume \( V_{sp} \) which can be derived from the speckle correlation function. Characteristic sizes are deduced from the method generating the speckle pattern. This is done in the Appendix B. Table 1 lists the typical speckle volumes for different generating algorithms for Random Phase Plates. It is also related to the spatial dimension \( (d = 2 \text{ or } 3 \text{ for } 2D \text{ or } 3D) \) in which numerical simulations of speckle pattern are
performed. The dimension $d$ therefore naturally enters in the ratio between the focal volume and the specific speckle volume.

<table>
<thead>
<tr>
<th>Method</th>
<th>Specific Speckle Volume</th>
</tr>
</thead>
<tbody>
<tr>
<td>2D Gaussian</td>
<td>$l_z l_z$</td>
</tr>
<tr>
<td>Flat-top</td>
<td>$\sqrt{\ln 2} k_0 w_{RPP} = 1.33 f_#^3 \lambda_0^3$</td>
</tr>
<tr>
<td>mod. RPP Flat-top</td>
<td>$6.2 k_0 / q^3 = 1.23 f_#^4 \lambda_0^3$</td>
</tr>
<tr>
<td>3D cyl. Gaussian</td>
<td>$2.9 k_0 / q^3 = 0.69 f_#^4 \lambda_0^3$</td>
</tr>
<tr>
<td>cyl. Flat-top</td>
<td>$14.3 k_0 / q^4 = 0.90 f_#^4 \lambda_0^3$</td>
</tr>
</tbody>
</table>

Table 1: Summary of specific volume of speckles in a RPP pattern, generated by different methods in the near field. The quantities $f_#$, $q$, and $\lambda_0$ are the optical f-number, the corresponding cutoff-wavenumber ($q \equiv 2\pi / \lambda_0 f_#$), and the laser wavelength in vacuum, respectively.

### 2.2 The higher order maxima $M_k < M_1$

We now proceed beyond this feature of a Gumbel law for the principal intensity maximum $M_1$ in a volume, by investigating the order statistics properties of the next intense maxima denoted as $M_k < M_{k-1} \ldots < M_1$ for $k > 1$. To obtain an adequate formulation for the probability distribution of the order statistics, one has to define the manner how the maxima are determined.

Since a speckle itself has a correlated structure, the points in the neighbourhood of a speckle maximum do consequently not form a set of independent intensity values. As discussed above the number of independent points is hence limited by the number $n_{sp}$ of speckles in the volume, of which we seek the maximum values and sort them in intensity. The most efficient method, which also can be adequately described by a probabilistic procedure, is to divide the volume into subcells of the size of a specific speckle volume, $V_{sp}$ (see Appendix B), and to determine the maximum of each subcell. Such a method is of course not free of errors which have to be handled so that the number of incorrectly taken maxima stays small.
with respect to the ensemble of \( n_{sp} \) maxima. For the details of the quite technical procedures choosing the maxima we refer to Appendix A.

We proceed in the way that the focal volume with the speckle pattern, evaluated at \( n \) mesh points, is sub-divided \( n_s \) sub-cells, \( J_1, \ldots, J_{n_s} \). We take \( n_s = n_{sp} \) sub-cells each containing \( m = [n/n_{sp}] \) elements (evaluated intensity values), where \( n \) is the total number of points of the mesh. Then, if we choose \( m \) such as each sub-cell contains a number of points \( m \) going to infinity when also \( n \to \infty \), if also the sub-cells are disjoint and sufficiently separated for large values of \( n \), we can consider that the maxima of the sub-cells are independent, and distributed following the Gumbel law with the shift parameter \( \theta \), where \( \theta \) is the shift parameter found in the above mentioned limit law Eq. (2) of a sequences of “clustering” exponential intensities.

Based on the work developed in Refs. (Leadbetter, 1982 & 1983) and (Embrechts et al., 1997), we find that \( M^k_m \), denoting the maximum of the \( k \)-th cell, follows a Gumbel law with the “clustering” parameter \( \theta \) (Middleton, 1960), (Leadbetter, 1982 & 1983)

\[
P(M^k_m \leq a^k_m + b^k_m x) \to H^\theta(x) = \exp(-\theta e^{-x})
\]

For \( M^1_m, \ldots, M^k_m, \ldots M^{n_s}_m \) being the maxima on the \( J_1, \ldots, J_{n_s} \) (asymptotically) independent sub-cells, the law for the joint probability, that the maxima \( M^k_m \) stay below the level \( u_n = a^k_m + b^k_m x \) for several maxima at the same time, is given by

\[
P(M^1_m \leq a^1_m + b^1_m x, \ldots, M^{n_s}_m \leq a^{n_s}_m + b^{n_s}_m x) \to \exp(-n_s \theta e^{-x}),
\]

because of the asymptotic independence of maxima in disjoints boxes.

We turn now to the order statistics of the maxima in the boxes. The result of Eq.(4) below is motivated by the following argument: We know that the exceedence of the level \( u_n \) by the sequence \( M_1, \ldots, M_n \) occurs with probability \( \hat{F}(u_n) \), if \( F \) is the common distribution function of the independent sequence \( M_1, \ldots, M_n \), this time of the maxima, with \( n = n_s \). If the mean number of exceedences, i. e. \( n_s \hat{F}(u_{n_s}) \), tends to \( \tau \), then the number \( S_{n_s} \) of exceedences of the level \( u_{n_s} \) by the sequence \( M_1, \ldots, M_n \) has limit distribution

\[
P(S_n \leq k) \to e^{-\tau} \sum_{h=1}^{k} \frac{\tau^h}{h!}
\]

Indeed, \( S_n = S_{n_s} \) is binomial with parameters \( (n, \hat{F}(u_n)) \), so that the result is just the Poisson limit for binomial distribution when \( n\hat{F}(u_n) \to \tau \).
We denote by $M_{k,n}$ the $k$-th maximum in the ordered sample of the maxima. Now, to evaluate the probability that $P(M_{k,n} \leq u_n) = P(S_n \leq k)$ we use Eq. (4) where the the explicit value for $\tau$ is taken as $\tau = -\ln H^\theta$. Indeed, by ordering the ensemble of maxima corresponding to the $n_s$ sub-cells, $M_{n_s, n_s} \leq ... \leq M_{1, n_s} = M_1$ denotes the order statistics of the sample (pattern), and we readily obtain from equation Eq (4)

$$P( M_{k,n_s} \leq a_{n_s} + b_n x ) \rightarrow H^\theta(x) \sum_{l=0}^{k-1} \left( \frac{-\ln H^\theta(x)}{l!} \right)^l = \frac{1}{(k-1)!} \Gamma_k(-\ln H^\theta)$$

where $\Gamma_n(x) = (n-1)!e^{-x} \sum_{k=0}^{n-1} (x^k/k!)$ is the incomplete Gamma function, and $H^\theta(x) \equiv \exp(-\theta \exp(-x))$ is the Gumbel (double exponential) law with the parameters $a_n = 1$ (or $a_n \to 1$) and $b_n = \ln n$, and $\theta$ as the (clustering) shift parameter.

The joint distribution of several large maxima concerns the simultaneous exceedence of multiple hierarchic levels $u_{1,n_s}, u_{2,n_s}, \ldots, u_{r,n_s}$ by the sequence $M_1, ... M_{n_s}$. Again, searching for the number $S_{k,n}$ of the exceedences of the level $u_{k,n_s}$ by the sequence, we obtain, see below, Eq. (6)

In fact, under the same hypothesis it is possible to write down the joint law of the maxima between order 1 and $k$ among the $n_s$ maxima, in decreasing order of intensity (see Refs. (Leadbetter, 1982 & 1983) or (Embrechts et al., 1997)). With $M_{n_s,k} \leq ... \leq M_{n_s,1} = M_1$ and $M_{k,s}$ denoting $k$-th most intense maximum, and $u_{n_s,k} \leq ... \leq u_{n_s,1}$, following $u_{n_s,k} = a_{k,n} + b_{n,k} x_k = \ln n + x_k$, one obtains for the joint probability for the $k$ most intense maxima

$$F^{(k)}(x_1, ... x_k) \equiv P(M_{1,n} \leq u_{n,1}, \ldots, M_{k,n} \leq u_{n,k}) =$$

$$= P \left( \frac{M_{n,1} - a_{n,1}}{b_{n,1}} \leq x_1, \ldots, \frac{M_{n,k} - a_{n,k}}{b_{n,k}} \leq x_k \right)$$

$$\rightarrow H^\theta(x_k) \sum_{l_2=0}^{\ldots} \ldots \sum_{l_{k}=0,1 \ldots k-1} \prod_{i=1}^{l_{i+1}} \frac{1}{l_{i+1}!} \left( \frac{\ln H^\theta(x_i)}{H^\theta(x_{i+1})} \right)^{l_{i+1}}.$$  

Let us explicitly express the result of $F^{(k)}(x_1, ... x_k)$ for the first two maxima:

$$P( \frac{M_{n,1} - a_{n,1}}{b_{n,1}} \rightarrow M_1, \frac{M_{n,2} - a_{n,2}}{b_{n,2}} \leq x_2 ) \rightarrow F^{(2)}(x_1, x_2) = \exp(-\theta e^{-x_2}) \left[ 1 + \theta (e^{-x_2} - e^{-x_1}) \right].$$

This allows us to evaluate the joint level of occurrence of the two most intense speckles, $M_{n_s,1} = M_1$ and $M_{n_s,2} = M_2$, that is the probability $P[M_{n,1} < M_1, M_{n,2} < M_2]$$ \equiv F^{(2)}(M_1, M_2)$, where of course
Figure 1: Joint distribution of the first two intensity maxima, $M_1 > M_2$, with $\ln n_{sp} \simeq 5.29$. Intensities are normalized to the mean intensity. The steepest gradient of the distribution in $M_2$ is asymptotically, for $M_1 \gg \ln n_{sp}$, found at $M_2 = M_1 - \ln 2 \simeq 4.6$. (a) surface 3D graphics representation, (b): contours (in reduced interval).

Figure 2: Derivative $\partial_{M_2} F^{(2)}(M_1, M_2)$, with respect to $M_2$, of the joint distribution $F^{(2)}(M_1, M_2)$ of the first two intensity extrema, $M_1 > M_2$, corresponding to what is shown in Fig. 1. Superposed to the contours is the curve following the steepest gradient of $\partial_{M_2} F^{(2)}$. 
Figure 3: (a) Marginal probability distribution function, $F(M_k)$ and (b) the corresponding marginal probability density $dF_k/dM_k$ of the 1st, 2nd, 3rd, and 5th ($k = 1, 2, 3, \text{and } 5$) most intense speckle as a function of their intensity $M_k$. Dots display simulation results based on 1600 realisations of an RPP speckle pattern, the lines correspond to the model expressed by Eqs. (10) and (3). The simulation volume contains 2048 laser wavelengths in both directions, containing about $n_{sp} \equiv V_{loc}/V_{sp} = 1330$ potential laser speckles such that the shift in the distribution is $\ln n_{sp} \simeq 7.2$.

$M_1 \geq M_2$: Figure 1 shows the joint distribution for the two largest maxima,

$$F^{(2)}(M_1, M_2) = \exp[-e^{-(M_2-a_n)/b_n})]$$

$$\left(1 + e^{-(M_2-a_n)/b_n} - e^{-(M_1-a_n)/b_n}\right),$$

expressing hence the (joint) probability that, at the same time, the most intense maximum is smaller than $M_1$, and that the second maximum is found below the value $M_2$ ($< M_1$). We know that the marginal probability distribution for the second maximum $F_2(M_2)$ is given by

$$F_2(M_2) \equiv \lim_{M_1 \to \infty} F^{(2)}(M_1, M_2)$$

$$= \exp[-e^{-(M_2-a_n)/b_n} - 2(M_2 - a_n)/b_n],$$

which is a consequence of the fact that for sufficiently high values of $M_1$, say for $M_1 > 2a_n$, the value of the most intense speckle is found most likely in the interval $0 < I < M_1$. For this reason, the cut in Fig.1, at high $M_1$ values ($M_1 \simeq 10 \simeq 2a_n$), approximately depicts the behaviour of $F_2(M_2)$.

Figure 3 shows the marginal probability density of the possible intensity values of several $k$-order maxima, which become more and more peaked (narrower) with order than the one for the most intense speckle.
Figure 4: Dots: spacing between the expectation values for the intensity of the $k$-th order maximum, $(M_k - M_{k-1})/(I)$ as a function of $1/k$. Solid line: spacing between the peak values, $(M_k^{(\text{peak})} - M_{k-1}^{(\text{peak})})/(I)$, of the probability density of maxima of successive order, following $\ln[(k + 1)/k]$, as a function of $1/k$.

Figure 5: Histogram constructed by plotting the order $k$ of the speckle maximum as a function of its peak (dotted line) and expectation values (dotted line with markers), $M^{(\text{peak})}$ and $M_k$ [from Eq. (12) and (5)], respectively, as well as of the mean values in intensity (asterisks) for each speckle maximum from numerical simulations. Two series of simulations are shown with the corresponding values $k \sim \exp -M_k^{(\text{peak})}$ and $k \equiv \text{eq. (15)}$, with (a) $\ln n_{\text{sp}} = 8.3$ (left group), and (b) $\ln n_{\text{sp}} = 9.75$ (right group). The simulations are based on about 1200 realisations of speckle patterns with identical optical properties, but different random distributions of the phase mask. The histogram corresponding to the peak values follows an exponential decrease $\propto e^{-M_k}$, while the histogram corresponding to the expectation values and to the simulations follows a slower decrease in intensity according to Eqs. (12) and (15).
2.3 Spacing in intensity between maxima of successive order

The marginal probability for the $k$-order maximum (decreasing in intensity with increasing order) is equivalently the limit of Eq. (6) for all $M_1 \ldots M_{k-1} \to \infty$ (Embrechts et al., 1997),

$$F_k(M_k) \equiv \lim_{M_1, \ldots, M_{k-1} \to \infty} F^{(k)}(M_1, M_2, \ldots, M_k) =$$

$$= H^q(x_k) \sum_{s=0}^{k-1} \frac{(-\ln H^q(x_k))^s}{s!},$$

(10)

with $x_k = (M_k - a_n)/b_n$, which allows us to evaluate the distribution and the probability density of all higher order maxima as a function of their intensity,

$$\frac{dF_k(M_k)}{dM_k} = \frac{1}{\Gamma(k)} \exp\{ -(M_k - a_n)/b_n - k(M_k - a_n)/b_n \}.$$

(11)

In particular, by searching the expectation value in intensity, $\bar{M}_k$, with respect to the probability density of the $k$-th order maximum, see Fig. 3

$$\bar{M}_k = \int_0^\infty M \frac{dF_k(M)}{dM} dM = a_{n_s} - \frac{d}{dk} \ln \Gamma(k) \equiv a_{n_s} - \Psi(k),$$

(12)

with the function $\Psi(k)$ standing for the Di-Gamma function (also call Poly-Gamma function of 0-th order and $\Gamma(k) = k!$ for integers $k$). From this expression, one can determine the spacing $\bar{M}_k - \bar{M}_{k+1}$ in $M$, given by the recursion identity $\Psi(k) - \Psi(k+1) = 1/k$. Between the successive orders of speckle maxima, we have consequently the relation (in dimensionless units after normalisation to the mean intensity of the overall speckle pattern $< I >$) $\bar{M}_k = \bar{M}_1 - \sum_{j=1}^{k-1} 1/j$ which is $\bar{M}_1 = \ln n_{sp} + \gamma$ (with $\gamma \approx 0.5772$ being the Euler Gamma constant) for the principal maximum, $k = 1$. All higher orders, $k > 1$, can nicely be approximated by

$$\bar{M}_k = \ln n_{sp} - (k-1) - 1/2(k-1)$$

(13)

for $k > 1$, with the expectation value of the most intense speckles of the order $k = 2, 3, 4, 5,$ and 6 yielding $\bar{M}_2 = \ln n_{sp} + \gamma - 1$ ($\approx \ln n_{sp} - 1/2$), $\bar{M}_3 = \ln n_{sp} + \gamma - 3/2$ ($\approx \ln n_{sp} - 0.94$), $\bar{M}_4 = \ln n_{sp} + \gamma - 11/6$ ($\approx \ln n_{sp} - 1.26$), $\bar{M}_5 = \ln n_{sp} + \gamma - 25/12$ ($\approx \ln n_{sp} - 1.51$), $\bar{M}_6 = \ln n_{sp} + \gamma - 137/60$ ($\approx \ln n_{sp} - 1.71$), where the approximate values from Eq. (13) are given in the parentheses.

The spacing between the expectation values for first maxima is hence different from the spacing between the peak values, $M_k^{(\text{peak})}$, of $dF_k(M_k)/dM_k$, which can be found from (Embrechts et al., 1997)

$$M_k^{(\text{peak})} = \ln n_{sp} - \ln k, (M_k^{(\text{peak})} - M_{k-1}^{(\text{peak})}) = \ln[(k-1)/k].$$

The histograms corresponding to either the
peak or the expectation values can be written in the form

\[
\text{order of speckle maximum } k \equiv n_{sp} \begin{cases} 
    f_p(M^{(\text{peak})}) \text{ for peak values}, \\
    f_e(M_k) \text{ for expectation values } . 
\end{cases}
\]

as illustrated in Fig. 5. In comparison to the spacing of the expectation values \(M_k\) the spacing of the peak values evidently yields an exponential decrease in the tail of the equivalent histogram \(f_p(M_k^{(\text{peak})})\), whereas \(f_e(M_k)\) has a slower decrease for the intense maxima and then comes close to an exponential one as well. This can be explained by of the fact that the shape of the marginal probability densities, (as a function of the speckle intensity) for the first order maxima is quite asymmetric around its peak, while for higher order maxima the shape becomes more and more symmetric. Figure 4 shows the comparison between peak values and expectation values as a function of the inverse of the speckle order, \(1/k\). In a similar way, we depict this difference between peak and expectation values in the form of an abundance histogram, Fig. mrefhisto, indicating how many speckles, namely \(\leq k\) is equal to the order of speckle maxima \(k\). speckles in average, with maximum intensity \(M\) are found in the interval \(M \geq M_k\).

The histogram established via the numerical simulations shows very good agreement with the histogram, \(k = n_{sp} f_e(M_k)\), where one associates the \(k\)-order maximum with its expectation value \(M_k\) in intensity, see Fig. 5. We are therefore led to consider that the order of the maximum \(k\) determines, in an ensemble average over numerous realisations, the number of speckles with a peak intensity \(M\) greater or equal to the expectation value of the \(k\)-th order maximum, i. e. \(M \geq M_k\). (Note that the ‘peak intensity’ of individual speckles \(M\) has to be distinguished from the ‘peak value’ \(M_k^{(\text{peak})}\) of the marginal probability density of the \(k\)-th order maximum). We can hence use the inverse of the expression Eq.(12),

\[
\text{number of speckles with peak intensity } \geq M_k \equiv \\
\text{order of speckle maximum } k = n_{sp} f_e(M_k) \equiv (\Psi)^{-1} (M_k - a_n) ,
\]

represented in Fig. 5. This equivalence allows us, hence, to use Eq. (5) as an expression for the “tail” of the distribution for high-intensity speckle maxima. The fact that we start counting, due to the order statistics, from the most intense speckle toward the less intense ones, has the consequence that the “body” of the speckle maxima distribution corresponds to very high orders \(k \sim e^{-1} n_{sp}\).
2.4 Computing the response from independent speckles

In the frame of the “independent hot spot model” (Rose & DuBois, 1994), it is of interest to compute the response of a speckle (or “hot spot”) pattern to a nonlinear physical process. For the purpose to compute the response \( R_{\text{tot}} \) to such a process from the ensemble of speckles in a plasma, the assumption of independent “hot spots” means that one can sum over the contributions to the response from each speckle, \( R(M_k)/n_{\text{sp}} \), i.e. \( R_{\text{tot}} \equiv \sum_{k=1}^{n_{\text{sp}}} R(M_k)/n_{\text{sp}} \), where the response of the speckles (numbered from \( k = 1 \) to \( n_{\text{sp}} \)) is a function of the speckle peak intensity \( M_k \). The validity criteria for such an approach are (i) that the process considered (e.g. stimulated scattering) can be expressed by either a stationary or an instantaneous response function, and (ii) that the contributions of adjacent layers in the speckle pattern do not give rise to subsequent re-amplification. In practice this means for the case of stimulated scattering in speckles, that the light scattered off different layers with respect to the laser propagation, originating from the (sub-) ensemble of speckles in each layer, can be considered as an incoherent superposition of the individual contributions.

We shall consider the particular case of stimulated scattering, for which the above mentioned criteria can be explicit for an inhomogeneous plasma (Tikhonchuk 1997): (i) the stimulated scattering process is in a convective amplification regime where the plasma waves, off which electromagnetic wave are scattered, are strongly damped; (ii) the plasma is inhomogeneous with a three-wave-coupling interaction length \( \ell_{\text{int}} \) that is shorter than the longitudinal speckle length \( 2l_z \) (see Appendix B), but longer than the speckle width \( 2l_\perp \). For a plasma wave damping rate \( \nu_n/\omega_n \) and a plasma inhomogeneity length \( L_\varphi \), the latter condition can be expressed by \( 2l_\perp < \ell_{\text{int}} = (\nu_n/\omega_n)L_\varphi < 2l_z \).

For a layer of the length of a speckle, \( \sim 2l_z \) (see Appendix B for the equivalence between \( l_z \), the Rayleigh length, and the length \( L_\parallel \)), the response function for the backscattered light flux produced due to parametric scattering instabilities, such as Stimulated Brillouin and Raman Scattering (SBS and SRS, respectively), can be written as \( R(M_k) = R_{\text{sp}}(M_k)M_k \). With \( R_{\text{sp}}(\cdot) \) we denote the single speckle reflectivity, i.e. the portion of light – with respect to the incident light flux – that is backscattered off a single speckle. A speckle having \( M_k \) times the mean intensity \( \langle I \rangle \) of the incident beam, contributes hence with \( R_{\text{sp}}(M_k)M_k \) to the total backscatter response.

Following the above-mentioned argument that the contributions of adjacent layers of the length of a speckle do not give rise to re-amplification, the total backscatter response can be expressed consequently....
by the double sum
\[
R_{\text{tot}} \equiv \frac{1}{n_{\text{sp}}} \sum_{j=1}^{n_{\text{sp}}} \sum_{k=1}^{n_{\text{sp},j}} M_{k,j} R_{\text{sp}}(M_{k,j})|_{z=z_j} ,
\]  
(15)

where \(n_{\text{sp},j}\) denotes the number of speckles in each of the \(j = [L_{\text{loc}}/2l_z]\) layers of length \(2l_z\), and where the contribution from each of the \(n_{\text{sp},j}\) speckles, \(M_{k,j} R_{\text{sp}}(M_{k,j})\), is evaluated in the \(j\)-th sub-layer around \(z = z_j\) with \(0 < j = [z_j/2l_z] < L_{\text{loc}}/2l_z\). The sub-division into sub-layers is necessary in case that the response in each of the \(j\) layers depends on the local plasma quantities like density, damping etc., and/or in case of pump depletion of the local laser power (see model in Tikhonchuk et al., 1997) such that \(\sum_{k=1}^{n_{\text{sp},j}} M_{k,j}/n_{\text{sp},j} \equiv \langle I \rangle\) is no longer constant, but diminishes along the propagation axis.

Under the condition that the spatial dependence of the local contributions on \(z\) is negligible, so that \(R_{\text{sp}}(M_{k,j})\) does not considerably vary over \(L_{\text{loc}}\) along the propagation axis, this expression simplifies to
\[
R_{\text{tot}} = \frac{1}{n_{\text{sp}}} \sum_{k=1}^{n_{\text{sp}}} R_{\text{sp}}(M_k)M_k ,
\]  
(16)

for the entire speckle pattern with \(n_{\text{sp}} = \sum_j n_{\text{sp},j}\). In this sum, the order of the contributing speckles does not matter. In the ensemble average over many realisations, one can use the values of the successive orders \(k\) in intensity of the speckle maxima \(M_k\), as developed above.

For a single realisation of a speckle pattern, as for a concrete speckle pattern, eq. (16) can easily be determined. In many cases, only the optical properties of a speckle pattern are known, but not the details of the pattern, and therefore a single realisation is statistically not representative for the high-intensity tail of the speckle distribution. For the ensemble average over many realisations of speckle patterns, the computation of the response requires an integration over the joint probability distributions \(F^{(k)}(M_1 \ldots M_k)\), which is practically a very tough procedure. We can, however, make use of the quite general feature seen in Fig. 2: for the example of the joint probability density for \(M_1\) and \(M_2\), it can be seen that the joint probability density in \(M_{k+1}\) already tends, as a function of \(M_k\) quite rapidly towards its asymptotic form corresponding to the marginal probability density.

For this reason, the model developed in section 2.3 can yield a good estimate of expression eq. (16) by evaluating \(R_{\text{sp}}(M_k)\) at the expectation value \(\bar{M}_k\) of the \(k\)-th maximum, yielding
\[
\langle \bar{R} \rangle \equiv \frac{1}{n_{\text{sp}}} \sum_{k=1}^{n_{\text{sp}}} R_{\text{sp}}(\bar{M}_k)\bar{M}_k .
\]  
(17)

The latter is a good approximation provided that the change of the response function \(R_{\text{sp}}\) in the interval around \(M_k = \bar{M}_k\) determined by the variance \(\sigma_k\) of the marginal probability density Eq. (11), namely \(|R_{\text{sp}}(\bar{M}_k - \sigma_k) - R_{\text{sp}}(\bar{M}_k + \sigma_k)|\), is small compared to \(R_{\text{sp}}(\bar{M}_k)\).
Using expression (12), the sum applies to discrete values of $M_k$. In contrast to our discrete approach, the distribution of speckle maxima as a function of their peak intensities has been expressed by a continuous expression (Garnier, 1999). It might therefore be useful to express the sum of Eq. (17) in the form of an integral

$$\langle \mathcal{R} \rangle \simeq \sum_{k=1}^{n_{sp}-1} \int_{M_k}^{M_{k+1}} R_{sp}(M) \, M \, df_e(M),$$

(18)

with $df_e(M) \equiv \left( df_e(M)/dM \right) dM$, knowing that $\int_{M_k}^{M_{k+1}} df_e = f_e(M_{k+1}) - f_e(M_k) \equiv 1/n_{sp}$. This approximation is, however, not very precise for orders $k < 3$, because we took interpolated values when performing the partial integration. The latter expression yields, in the continuous limit, $\int_0^1 R_{sp}(M) \, M \, df_e(M)$. We can here use the fact that the values of $\bar{M}_k$ can be approximated in a continuous manner, using Eq.(13), such that he inversion of Eq. (5) readily yields a nice approximate explicit expression of the speckle maxima abundance $f_e(\bar{M}_k)$, namely with the function

$$f_e(\bar{M}_k) \simeq \exp\left\{ \frac{1}{2} e^{\bar{M}_k - \ln n_{sp} - \bar{M}_k} \right\} \quad \text{for} \quad \bar{M}_k \leq \ln n_{sp} + \gamma,$$

(19)

which is normalized to the total number of speckle maxima, namely $n_s = n_{sp} = \exp(a_n) = \exp(\ln n_{sp})$.

2.5 Numerical simulations of order statistics

We have performed numerical simulations in order to determine the order statistics from speckle patterns generated by a Random Phase Plate. For this purpose we have solved the stationary paraxial equation for a forward-going laser pulse

$$\left[ 2i k_0 \partial_z + \partial_x^2 + \left( \frac{\omega^2}{c^2} - k_0^2 \right) \right] A = 0,$$

(20)

where the wave number $k_0$ and and the frequency $\omega$ are chosen such that $\omega^2/c^2 - k_0^2 \equiv 0$, because of the dispersion in a homogeneous medium or the vacuum. The corresponding solution of a light field composed by speckles can be written as the superposition of an ensemble of complex fields $a_m$, all with the same propagator $e^{i k_0 z - i \omega t}$, but with distinct values in the transverse direction. The field describes, in paraxial approximation, the propagation from the so-called "near field", where the speckle-generating "Random Phase Plate" is situated, $\sum_m a_m e^{i \varphi_m}$, toward the "far field" $E(x, z)$ in the focal plane. How concrete realisations of RPP beams can be done is explained in more detail in Appendix B.

Our simulations are performed on a highly resolved mesh over a large lateral size and over many characteristic lengths, namely Rayleigh lengths of speckles in the propagation direction, with typical
number of speckles of 2000, thus \( \ln n_{sp} \sim 7.5 \). The total number of mesh points - 8 to \( 16 \cdot 10^6 \) is considerably higher than \( n_{sp} \) such that a great number of points inside a speckle are clustered and the shift \( \ln n_{sp} \) on the intensity axis applies. From the simulations we have determined the most intense speckle maxima, starting with \( M_1 \), up to order 6 in the case presented here. To determine the highest intensity value \( M_1 \), being a local maximum at the same time, follows a straightforward procedure. The next maxima following in order, \( M_n < M_{n-1} < \ldots M_1 < M_2 < M_1 \) have to be determined with caution because the error in ordering the maxima increases with the order. The algorithm chosen may therefore modify considerably the result. We have therefore devoted Appendix A to explain the searching procedure for the maxima. Let us just mention that the most efficient method is to divide the volume in sub-cells of the size of an individual speckle, in respecting the typical sizes due to the manner how the speckle pattern is generated with the RPP. For the different cases of RPP masks in the “near field”, we have listed the specific speckle volume in Table 1. The maximum of each sub-cell is retained, and all maxima are sorted in magnitude, however, to reduce errors, a discriminating filter can be applied, see Appendix A. In order to establish the probability distribution numerically, we have computed a large number of realisations of RPP masks with identical optical properties: that means that the focusing optics, the size of the overall phase plate, and the size of the phase plate elements is the same in each realisation, whereas the phase shift \( \varphi^j_m \) in each individual element \( m \) follows a random value sequence from realisation \( j \) to realisation \( j+1 \). It is unimportant whether \( \varphi^j_m \) randomly jumps between the values 0 and \( \pi/2 \) or whether the values of \( \varphi^j_m \) have homogeneous distribution in the interval \( [0, \pi/2] \). The order statistics can be determined in registering the intensity values of the first \( k \) (here 6) most intense maxima. Numerically determined probability distributions for extreme speckles are determined are shown in Fig. 5. From the same statistics we have determined the expectation value of the \( k \)-th order maximum, which is compared to the values derived from probability theory in Eqs.(13) and (19) above. Both comparisons shown in Figs. 5 and 4 show very good agreement between simulations and theory. In Figure 4 the spacing between the expectation values for the successive maxima clearly exhibits a departure from the \( 1/k \) law for small \( k \), i. e. for the most intense speckles, giving rise to a simply exponential \( \exp(-M/\langle I \rangle) \) speckle distribution. The histogram, see Fig. 5, counting the number of maxima in the interval \( M < M_8 \), which represents the upper tail of the speckle distribution, shows hence a decrease that is slower than an exponential for the maxima below order 8, namely for \( M > M_8 \simeq 3\langle I \rangle \).
Figure 6: Backscatter reflectivity of Stimulated Brillouin Scattering, computed from different realisations of a speckle pattern generated by a RPP. Each point corresponds to a realisation for which the reflectivity is plotted as a function of the peak intensity $M_1$ in the respective speckle pattern. Simulation parameters: average beam laser intensity $I = 2 \times 10^{14} \text{W/cm}^2$ at the wavelength $\lambda = 1 \mu\text{m}$, electron temperature $T_e = 1 \text{keV}$, and density $n_e = 0.1 n_c$, ion wave damping $\nu_a/\omega_a = 0.1$, velocity profile inhomogeneity length $L_v = 150 \lambda$, optical f-number $f_\# = 5$ resulting in a specific speckle volume of $2l_{z,RPP} \times 2l_{z,RPP} \simeq 175\lambda \times 4.2\lambda$ (see table 3 of Appendix B for “modified flat top”), and plasma volume $L_z \times L_{x,z} = 500\lambda \times 717\lambda$, equivalent to $\ln n_{sp} \simeq 6.2$.

3 Backscattering from extreme speckles

Let us show the pertinence of the order statistics developed above for the example of the backscatter reflectivity value of a parametric instability. We will illustrate it for the particular example of Stimulated Brillouin Scattering (SBS) by making use of simulation results from the *Harmony* interaction code (Hüller et al., 2006) for several realisations of a speckle pattern generated with a RPP. The simulations were carried out in an inhomogeneous plasma with a linear velocity gradient of an exploding-foil type plasma such that Stimulated Brillouin Scattering in single speckles should follow that amplification according to Rosenbluth’s model (Rosenbluth M. N. 1973). Ion acoustic waves are assumed to be strongly damped.

To compute the SBS reflectivity from the model developed here, we can assume that the contribution to backscattering comes independently from each speckle. The reflectivity can be compute by summing over all speckles with a response function of the reflectivity $R(M_k)$ parametrized on the speckle peak intensity $M_k$, as done e. g. in Refs. (Hüller S. et al., 1998), (Tikhonchuk V. T. et al., 1997), following an amplification described by Rosenbluth’s model (Rosenbluth M. N. 1973). For a concrete speckle
pattern, the values of the speckle maxima \( M_k \) are known. One can hence compute the reflectivity \( R_{\text{tot}} \) corresponding to backscattering of the overall speckle pattern by \( R_{\text{tot}} = \langle \mathcal{R} \rangle \) from Eq. (17), for all \( M_k \) being normalized to the mean intensity \( \langle I \rangle \). The value \( R(M_k) \) indicates, as mentioned already earlier, the fraction of light flux that has been backscattered from an individual speckle. One can assume that the single speckle reflectivity is a result of an exponential amplification with \( R_{\text{sp}}(M) \propto R_{\text{noise}} \exp\{GM\} \) from a noise level \( R_{\text{noise}} \), with \( G \) being the SBS gain for an average-intensity speckle, and \( M \) standing for the peak intensity value of an individual speckle. For simplicity, here, we limit the single speckle reflectivity by unity once the speckle peak intensity \( M_k \) exceeds the value of \( \ln(1/R_{\text{noise}})/G \). We readily obtain

\[
R_{\text{tot}} = \frac{1}{n_{\text{sp}}} \sum_{k=1}^{n_{\text{sp}}} M_k \min\{1, R_{\text{noise}} e^{GM_k}\},
\]

which can be further evaluated using the approximate expression Eq. (13) for the expectation value \( M_k \) of the successive orders of the speckle maxima,

\[
R_{\text{tot}} \approx \frac{M_1}{n_{\text{sp}}} \min\{1, R_{\text{noise}} e^{GM_1}\} + \frac{1}{n_{\text{sp}}} \sum_{k=2}^{n_{\text{sp}}} \left( \ln \frac{n_{\text{sp}}}{k-1} - \frac{1}{2(k-1)} \right) \min\left\{1, R_{\text{noise}} \left( \frac{n_{\text{sp}}}{k-1} \right)^G e^{-G/2(k-1)} \right\}.
\]

It is important to note that, starting from the general expressions Eq. (17) and (21), the contribution of the most intense speckles to the total reflectivity \( R_{\text{tot}} \) yields a non-negligible portion. For the case of a realisation yielding a particular pronounced principal maximum such that difference between the first two maxima \( M_1 - M_2 \) is considerably larger that the successive differences \( M_k - M_{k-1} \), the contribution of the principal maximum to \( R_{\text{tot}} \) can assume a high percentage.

Figure 6 shows the simulation results for the SBS reflectivity of a speckle pattern with about \( n_{\text{sp}} \sim 1820 \) speckles. The SBS gain corresponding to these simulation yields for average intensity speckles at \( \langle I \rangle \) the value of \( G = 1.1 \), the noise in the simulations was such that an average reflectivity of \( R_{\text{noise}} \sim 10^{-5} \) results in absence of amplification (i.e. for \( G = 0 \)). Since the gain for an average-intensity speckle (at \( \langle I \rangle \)) is small, only intense speckles can contribute to \( R_{\text{tot}} \). For this reason, each different realisation of simulations reaches eventually a different reflectivity level, plotted here as a function of the principal intensity maximum \( M_1 \). The dependence of the reflectivity on \( M_1 \) brings to evidence the fact that the contribution of the most intense speckles, and their fluctuations in intensity, cannot be neglected if the number of speckles is limited.

The evaluation of the average SBS reflectivity corresponds to an ensemble average of many realisations of speckle patterns with each time \( n_{\text{sp}} \) speckles. Applying Eq. (23) for the current example, this yields a
Figure 7: (a) Comparison between the speckle maxima probability distribution computed from Eq. (19) (solid line) and Garnier’s results (dashed line) (Garnier, 1999).

(b) Mean backscatter reflectivity computed from Eq. (19) and (21) (solid line) and from Garnier’s model (dashed line) with a single speckle reflectivity parametrized with $R_{sp}(M_k) = \min(1, R_{\text{noise}} \exp(GM_k))$. The isolated points show the reflectivity corresponding to simulation results with the code Harmony. The multiple values displayed by points at $G=1.1$ correspond to the range of values displayed in Fig. 6 for several realisations with the same parameters.

The value of $R_{\text{tot}} \approx 0.0003$. The histogram in Figure 6 shows that the reflectivity assumes this value around the expectation value of the most intense speckle, namely $M_1 = \ln n_{sp} + .5772$, being $M_1 \approx 8.1$ for $n_{sp} \approx 1820$, hence in good agreement with the average over the realisations in our simulations.

We have compared our results to previous work on the statistics on speckle maxima, which is shown in Fig. 7). In addition to the speckle maxima histogram $f_e(M_k)$ corresponding to our model, see Eqs. (15) and Eqs. (19), we show the corresponding distribution from work done by Garnier (Garnier, 1999) where $f_G(M)$ has a powerlaw-type times exponential dependence in $M$,

$$f_G(M) \sim n_{sp}(c_1 M^{d/2} + c_2M^{-1+d/2})\exp(-M)$$

with $c_1 = \pi/2 + 1$ and $c_2 = 1$ in 2 spatial dimensions, $d=2$, and $c_1 = 1$ and $c_2 = -3/10$ in 3 spatial dimensions, $d=3$. To understand the impact of the differences between the PDFs for $f_e(M_k)$ and $f_G(M)$, we have furthermore computed the mean backscatter reflectivity, as explained above, both from Eq. (refrmean) and by convolving the PDF $\propto f_G(M)$ with the single speckle reflectivity parametrized by the gain value $G$ (of an average-intensity speckle),

$$\langle R \rangle \equiv \int_0^1 \min \{1, R_{\text{noise}} e^{GM} \} M \, df_G(M) ,$$

here with a noise level $R_{\text{noise}} = 10^{-5}$, and again with $df_G \equiv (df_G/dM) \, dM$. Note that, for practical
reasons, the integral is truncated at the lower bound of the mean intensity of the beam ($M \equiv 1$) because (i), for not too high gain values, speckles with lower intensities do not potentially contribute, and because (ii) the speckle maxima distributions, both $f_c$ and $f_G$, are usually are poorly reliable for lower values, which is a consequence of the fact that a low intensities (with respect to the mean beam intensity), speckles can hardly be distinguished from each other. Let us remark that in the shown example, the gain is entirely proportional to the average (speckle) intensity. The gain can also be changed by varying the speckle size (not the case in our simulations), which would, however, for the same overall volume, change the number of speckles. This has to be taken into account when using Eq. (23).

In the past, the speckle maxima distributions due to work by Rose and DuBois (Rose & DuBois, 1993, May) and Garnier (Garnier, 1999) have shown to yield good agreement with numerical simulations (Hüller et al., 1998). In what concerns the regime illustrated here, we observe that for sufficiently high gain, say $G \geq 2$, the comparison of our and Garnier’s results do not show a particular difference, while the difference is striking for relatively small values $0.3 < G < 2$. In particular, our simulation result, which yields an ensemble average reflectivity $(R) \sim 0.0003$, is well reproduced by our model computed with Eq. (23). Note that for the shown case, with the number of speckles of the order of $\sim 2000$, the most intense speckles will have intensities around 8 times the mean intensity, with statistical fluctuations around this value, as illustrated in Fig. 6. The variation of the average reflectivity is hence non-negligible, as indicated by an error bar for the value of $G = 1.1$ in Fig. 7b. Following Garnier’s PDF using Eq. (23), equivalent to Eq. (23) for a continuous expression $f_G(M)$, the ensemble average reflectivities yield values for $0.7 < G < 2$ that are almost by one order of magnitude higher than those of the simulations, thus fairly outside the interval indicated by the error bar.

Just in the mentioned interval of gain, $0.5 < G < 2$, the contribution of the (about) 10 most intense speckles to the average reflectivity is of the order of 20% or higher, so that the order statistics of intense speckles, have a strong impact, see Fig. 8. At high gain values, the main contribution to the reflectivity is evidently dominated by the numerous speckles in the “body” of the speckle distribution function, usually well-described in previous work (Rose & DuBois, 1993, May), (Garnier, 1999). Differences in models describing the tail of the speckle distribution function should not be too important in this high-gain regime. In the moderate- and low-gain regime, $G \leq 1$, however, a good description of the tail in the speckle maxima distribution is necessary, provided that the gain in the most intense speckles, $G_k = G M_k$ with, say, $1 \leq k < 20$, is sufficiently high, say $G_1 \sim \ln n_{sp} \ln(1/R_{\text{noise}})$. 

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Figure 8: Proportion of the sum of backscatter reflectivities from the speckles up to the $K$-th order to the total reflectivity $R_{\text{tot}}$ of all speckles, corresponding to the above-mentioned cases, for 3 different gain values $G = 1.1$ (blue dots, intermediate), $G = 2$ (red dots, upper curve), and $G = 3$ (green dots, lower curve).

4 Discussion and Conclusions

With the help of the theory of extrema, we have derived the order statistics of intense laser speckles in a laser beam generated by a Random Phase Plate. The distribution function for each of the most intense laser speckles, conditioned by the order $k$ of occurrence in peak intensity $M_k$ in speckle patterns, has been derived from the joint probability distribution $F^{(k)}(M_1, ..., M_k)$ over $k$ ($\geq$) speckle maxima. The theoretical result is very good agreement with numerical simulations over numerous realisations of speckle patterns. To establish the order statistics from these simulations, a selection procedure has been applied that is based on the subdivision of the simulation volume in sub-cells having the characteristic size of individual speckles. In a further step, we have established the histogram $f_{e}(\bar{M}_k)$ for the high-intensity tail of all speckle maxima via the expectation value in intensity of each order $k$ derived from the marginal probability density, $dF_k(M_k)/dM_k$, for each speckle maximum.

For the example of Stimulated Brillouin Scattering from a speckle pattern in the regime where the contribution from “independent” speckles can be approximated by a sum, we have compared our model with numerical simulations, revealing good agreement. The comparison underlines as well that particular importance to the precision of the description in the high-intensity tail of the speckle distribution function has to be given. The fact that our approach is accentuated to this regime, it allows to compute, with higher precision than available from previous work, the contribution from intense speckles to critical nonlinear processes.
In experiments with relevance to laser fusion, the influence of high-intensity speckles is not at all desired. Therefore laser beams are designed to contain as much as $\sim 10^5$ speckles. It is furthermore evident that the experimental parameters have been designed such that speckles which have peak intensities of 2-5 times the mean intensity, should dominate essentially most of the physics processes, as far as they can be entirely controlled. On the other hand, experiments that have been performed with a relatively “small” number of, say $\ll 10^4$, speckles in the plasma volume, may require a more precise analysis of the speckle distribution, as provided by our model, because the onset of potentially nonlinear processes may be energetically important under such circumstances. Furthermore, processes with threshold-like behaviour for the onset of absolute instabilities, or of laser-beam self-focusing, might be particularly sensitive to the distribution of speckles in the high-intensity tail (even though saturation of such processes may limit their importance).

The theory developed here is based on the assumption of an idealized speckle pattern generated by RPP beams. This means that at least speckles with peak intensities higher than the average intensity of the beam, $\langle I \rangle$, can all be characterized by the same characteristic sizes along ($\ell_\parallel$) and across ($\ell_\perp$) the laser propagation direction. For speckle patterns generated by the superposition of optically smoothed laser beams, these characteristic sizes can be described by a probability distribution, namely $F(\ell_\parallel, \ell_\perp)$, where $\ell_\parallel$ can be either independent of, or coupled to the value of $\ell_\perp$. The latter is the case for RPP beams (with $\ell_\parallel \propto k_0\ell_\perp^2$).

Nonlinear response of processes like self-focusing in intense speckles, as well as pump-depletion due to strong scattering in speckles, can considerably change the image of the speckle pattern inside the plasma. This is not taken into account in the present work. A calculation including the nonlinear modification of the speckle pattern in a plasma has to be done for such purposes, which can hardly give a universal response to the potentially arising mechanisms. Numerical simulations can partially give an answer to this problem. Starting from simulations, statistics of so-called “plasma-induced smoothing” of laser beams have already been investigated (Hüller, 2003) for the case of “dancing filaments” (Schmitt & Afeyan, 1996).

To establish the order statistics of speckle patterns with multiple characteristic lengths, or with a statistical distribution of characteristics lengths, $\ell_\perp$ and $\ell_\parallel$, it is more appropriate to consider boxes of different sizes. Therefore, as above, one has to proceed in the way that the focal volume, containing the speckle pattern, is evaluated at $n$ mesh points, and is sub-divided in $n_s$ sub-cells. Each cell contains the
number of $n_k$ points, for which holds \( \sum_k n_k \leq n \). We can still consider that the maxima of the sub-cells are independent, and being distributed following the Gumbel law with the shift parameter $\theta_k$, if (i) each sub-cell contains in the limit of a very large number of points $n_k$, say $n \to \infty$, and if (ii) the sub-cells are disjoint and sufficiently separated for large values of $n$. Then $\theta_k$ is found from the limit law of a sequences of “clustering” intensities (exponential law) for the fraction of points observed in the $k$-th sub-cells.

The fact that the superposition of optically smoothed laser beams gives rise to non-homogeneous speckle sizes (characterized by a statistical distribution of them), may lead to a modulation of the limit law for the most intense maxima. This is subject of work in progress.

**Acknowledgements**

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References


Malka V., Renard N., Hüller S., Pesme D., Amiranoff F., Baton S. D., Modena A.,


Appendix A: Choice of speckle maxima in sub-cells

In order to find an adequate selection procedure for the speckle maxima by subdivision in cells, various problems have to be handled:

1) First of all the characteristic size of speckles has be chosen adequately, which can be done with the help of the lengths listed in Appendix B, and the specific speckle volume listed in Tab. 1. An inadequate choice may increase the error in the order statistics with increasing order of the maxima.

2) To locate the subcells in the simulation volume, there are essentially two main choices for uniform size cells: (i) the cells regularly are placed in adjacent order starting from one corner of the simulation volume, or (ii) all cells are placed around the cell containing, in its center the principal maximum of the speckle pattern.

Once the volume is subdivided in $n_{sp}$ sub-cells, it is straight-forward to determine the principal maximum of each. The resulting $n_{sp}$ values found can be ordered in decreasing order. However, subsidiary of intense maxima that drop into a neighbouring cell, because of the vicinity of the principal speckle peak to a cell border. Such subsidiary maxima that may exists depending on the generating method (“flat-top” or “Gaussian”, see Appendix B), are a consequence of the coherent structure inside speckles. They can be considered erroneously as the real peak maximum of the cell and hence falsify the order statistics. Method (ii) avoids this error for the case of the most intense speckle of the entire volume. However, beyond this order, the confusion persists between subsidiary maxima, that are correlated to the main speckle peak, and the real peak value of another speckle. For this reason, a second selection step is necessary: after having established the order between the maxima, all those ones are eliminated which are too close to an intenser maximum in a neighbouring cell. In practical terms one can define a distance that should not fall below a length of the order of $\ell_\perp$ and/or $\ell_\parallel$ (see herefore Appendix B). It is preferable to eliminate realisations that are subject to error.

Following the same arguments as in Ref.(Leadbetter, 1982 & 1983), which we apply to random variables representing speckle intensities, the location and the height of speckle peaks can be considered as being independent of each other; furthermore the distribution of the location is uniform, so that we expect that method (i) and (ii) for the choice of subcells give the same results [despite small errors due to problems evoked in 2)] provided that simulations are based on a sufficiently high number of realisations and points in space. Nevertheless, for a reasonable choice of simulations, it appears that numerical results for the cells according to method (ii) are the best with respect to our model results.
Appendix B: a guide for modelling single and multiple laser speckles in 2D and 3D

A Gaussian laser beam can be defined by its electric field according to the paraxial solution in the space dimensions $\xi$ along and $x_\perp$ across the propagation direction

$$E(x_\perp, \xi) = f_{nD}(\xi) \exp \left\{ \frac{x_\perp^2}{|W(\xi)|^2} \left( 1 - i2\xi/k_0w^2 \right) \right\}, \quad (24)$$

with the width of the beam

$$|W(\xi)| = \sqrt{1 + (2\xi/k_0w^2)^2} = \sqrt{1 + (\xi/L_R)^2}$$

at the distance $\xi$ from the focus, involving the width at best focus $w$ and the definition of the Rayleigh length $L_R$. In 2D, we have $f_{2D}(\xi) = W(\xi)^{-1/2}$ being a function decreasing slower with the distance $\xi = z - z_0$ in the focal point at $z_0$ than in 3D (cylindrical) geometry, for which the decrease $f_{3D}(\xi) = W(\xi)^{-1}$ to $2^{-1/2}$ occurs at the one Rayleigh length $\xi = L_R$. It is important to note that the integrals over the square amplitudes differ decisively from 2D to 3D, namely the integral does not converge in 2D, while the integral converges rapidly in 3D geometry,

$$\int_{-L}^{L} |f_{nD}(z)|^2 \frac{dz}{L_R} = \begin{cases} \text{arsinh} \frac{L}{L_R} \to \infty \text{ for } 2D, \\ \text{arctan} \frac{L}{L_R} \to 2\pi \text{ for } 3D, \end{cases}$$

for $L \to \infty$.

In the so-called “near field image”, which is principally the Fourier transform of the field in the focal plane (usually called “far field”), we can write the field as

$$E(k_\perp, z) = Q(k_\perp)e^{i(c^2k_\perp^2/2\omega_0)z/c_0} \quad (25)$$

where the function $Q(k_\perp)$ is also a

(i) Gaussian $Q(k_\perp) \propto \exp(-4w^2k_\perp^2)$ for the Gaussian field Eq. (24), and a so-called

(ii) “flat-top” for

$$Q(k_\perp) = \begin{cases} \text{const} = 1 \text{ for } |k_\perp| \leq q \\ 0 \text{ otherwise}, \end{cases}$$

For both cases, in 2D, the integrals over $z$ in the square amplitude diverge. To remedy the problem of the missing convergence of this integral in 2D, H. Rose (Rose, 1997) has suggested a beam called
(iii) “modified flat-top” by using a function

\[ Q(k_\perp) \equiv \frac{|k_\perp|}{q} \text{ for } |k_\perp| \leq q, \text{ and } 0 \text{ elsewhere.} \]

This yields the following shape around the focal point,

\[ E(x_\perp, 0) = 2q \left( \frac{\sin qx}{qx} \cdot \frac{1}{2} \left( \frac{\sin qx/2}{qx/2} \right)^2 \right), \tag{26} \]

which has a converging integral along \( z \).

The cutoff value \( q \) of the “flat-top” beams in the transverse wave vector determines the width \( \propto q^{-1} \) in of the beam in the focal plane. In 3D, the shape along the propagation axis (for \( x_\perp = 0 \)) is given by

\[ E(0, z) = q \exp \{iq^2z/4k_0\} \frac{\sin q^2z/2k_0}{q^2z/2k_0}. \]

being identical to the one of the cylindrical flat-top speckle, for which the amplitude for \( z = 0 \) is given by

\[ E(r, 0) = 2q \frac{J_1(qr)}{qr} \text{ with } r = \sqrt{x_\perp^2}. \]

For both types of beams, the modified flat-top in 2D and the cylindrical flat-top in 3D, due to the identical behaviour in \( z \), the integral \( \int_{-L}^{L} |E(z, 0)|^2 dz \) converges for \( L \to \infty \) to the same value \( \pi q \).

<table>
<thead>
<tr>
<th>2D</th>
<th>( l_\perp )</th>
<th>( L_\perp )</th>
<th>( l_z )</th>
<th>( L_\parallel )</th>
<th>( l_z/l_\perp )</th>
</tr>
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<tbody>
<tr>
<td>Gauss</td>
<td>( \sqrt{\ln 2a} )</td>
<td>( \sqrt{\pi/2a} )</td>
<td>( k_0a^2 = L_R )</td>
<td>( \infty )</td>
<td>( k_0l_\perp/\ln 2 )</td>
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<td>flat-top</td>
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<td>( 5.4k_0/q^2 )</td>
<td>( \infty )</td>
<td>( 2.7k_0l_\perp )</td>
</tr>
<tr>
<td>mod flat-top</td>
<td>1.1/q</td>
<td>2( \pi/3q ) ( \approx ) 2( \pi/3q )</td>
<td>5.6( k_0/q^2 )</td>
<td>2( \pi k_0/q^2 )</td>
<td>( \approx 4.5k_0l_\perp )</td>
</tr>
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</table>

<table>
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<th>( l_\perp )</th>
<th>( L_\perp )</th>
<th>( l_z )</th>
<th>( L_\parallel )</th>
<th>( l_z/l_\perp )</th>
</tr>
</thead>
<tbody>
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<td>cyl Gauss</td>
<td>( \sqrt{\ln 2a} )</td>
<td>( \sqrt{\pi/2a} )</td>
<td>( \approx 0.64L_R )</td>
<td>( \pi L_R/2 )</td>
<td>0.93( k_0l_\perp )</td>
</tr>
<tr>
<td>cyl flat top</td>
<td>1.6/q</td>
<td>16/3( \pi q )</td>
<td>5.6( k_0/q^2 )</td>
<td>2( \pi k_0/q^2 )</td>
<td>( \approx 2.2k_0l_\perp )</td>
</tr>
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Table 2: Summary of typical longitudinal and transversal sizes of a single speckle, generated by different methods in the near field.

For the ordinary “flat-top” speckle, (ii), in 2D (i.e. a single transverse dimension), generated by \( Q(k_\perp) = \text{const} \) in Eq. (25), the shape of the field in the 2D space spanned by the coordinates \( (x_\perp, z) \) is
hence, along each axes, given by
\[ E(x_\perp, 0) = 2q^2 \sin qx \]
and
\[ E(0, z) = \sqrt{\pi} \frac{1 + i}{2} q \exp \left[ \frac{(1 + i) \sqrt{q^2 z/k_0}}{2} \right], \]
being equivalent to the behaviour of a Fresnel integral as a function of \( z \). From these shapes, we henceforth deduce characteristic lengths, indicating the measure of a typical size of the beams, which eventually allows to determine the specific speckle volume.

In Table 2 we indicate the correlation lengths with respect to the transverse \( x_\perp \) and the longitudinal dimension \( z \), for which we use the following definitions
\[ L_\perp = \int_0^\infty |E(x, 0)/E(0, 0)|^2 \, dx \]
and
\[ L_\parallel = \int_0^\infty |E(0, z)/E(0, 0)|^2 \, dz, \]
used in the tables below. Alternatively, by defining the widths \( l_\perp \) (in \( x_\perp \) or \( r \)) and the lengths \( l_z \) by the distance where the speckle intensity assumes one half of the peak value, one obtains the values (by using \( a = 2^{-1/2} \)) as listed in Tab. 2.

**B.1 Case of the equivalent RPP beams**

Equivalent to the above definitions one can determine the corresponding beam features when a Random Phase Plate is introduced in the beam path before the focusing lens, see Ref. (Rose & DuBois, 1993b)

\[ E(x_\perp, z = 0, t) = \sum_{k_\perp} |E(k_\perp, t)| \exp \{i k_\perp \cdot \vec{x}_\perp + i \phi_k \} \]

where \( \phi_k \) jumps randomly between the values 0 and \( \pi \) in the intervals in \( k_\perp \). One can evaluate the shape of individual speckles in the focal plane without knowing the explicit distribution of the \( \phi_k \)-values, with the help of the auto-correlation function defined as

\[ C_{EE}(\vec{x}_\perp - \vec{x}_\perp', z - z') \equiv \langle E(\vec{x}_\perp, z, t) E^*(\vec{x}_\perp', z') \rangle \]
\[ = \sum_{k_\perp} \left| E(k_\perp) \right|^2 e^{i k_\perp \cdot (\vec{x}_\perp - \vec{x}_\perp') - i k^2 (z - z')/2k_0} \]

For the particular case of the “modified flat-top” with a RPP one obtains, we can again use Eq. (25) but with an amplitude function, now going with the square root of the perpendicular wave vector, \( Q_{RPP}(k_\perp) = |k_\perp/q|^{1/2} \) for \( |k_\perp| < q \) which is different from the coherent case, (iii), discussed above. The expression for the 2D standard flat top, the cylindrical flat top and the Gaussian are immediate from the above definition 28 and are not explicit here.
We list in the following the typical lengths and widths of speckles generated by a RPP corresponding to the generating “near-field” configuration. The corresponding values for the specific volume occupied by a speckle are listed in Table 1 found in the main text. The latter is of particular importance to determine the number of potential speckles in a finite focal volume.

To have equivalence between Gaussian and flat-top description, the following condition has to be satisfied (Rose & DuBois, 1993, May) \( q = k_0/2f_\# = \pi/(f_\#\lambda_0) \) yielding in 2D: \( w_{RPP,2D} \approx 2.1(1.95)/q \approx 0.64f_\#\lambda_0 \) between Gaussian and flat-top (modified flat-top); in 3D: \( w_{RPP,2D} \approx 2.65/q \approx 0.84f_\#\lambda_0 \).

<table>
<thead>
<tr>
<th></th>
<th>( l_{\perp,RPP} )</th>
<th>( l_{z,RPP} )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Gauss</td>
<td>( \sqrt{2k_0w_{RPP,2D}} \approx k_0w_{RPP,2D}^2/2 )</td>
<td>( 1.67/q \approx 0.53f_#\lambda_0 )</td>
</tr>
<tr>
<td></td>
<td>( \approx 2k_0/q^2 \approx 2.5f_#^2\lambda_0 )</td>
<td></td>
</tr>
<tr>
<td>Flat top</td>
<td>( \approx 1.5/q \approx 0.45f_#\lambda_0 )</td>
<td>( \approx 5.5k_0/q^2 )</td>
</tr>
<tr>
<td></td>
<td>( \approx 3.5f_#^2\lambda_0 )</td>
<td></td>
</tr>
<tr>
<td>mod Flat top</td>
<td>( \approx 1.1/q \approx 0.35f_#\lambda_0 )</td>
<td>( \approx 5.5k_0/q^2 )</td>
</tr>
<tr>
<td></td>
<td>( \approx 3.5f_#^2\lambda_0 )</td>
<td></td>
</tr>
<tr>
<td>3D</td>
<td>( \sqrt{2k_0w_{RPP,3D}} \approx 0.64k_0w_{RPP,3D}^2/2 \approx 2.1/q \approx 0.7f_#\lambda_0 )</td>
<td>( 2.2k_0/q^2 \approx 1.4f_#^2\lambda_0 )</td>
</tr>
<tr>
<td>Flat top</td>
<td>( 1.6/q \approx 0.5f_#\lambda_0 )</td>
<td>( \approx 5.6k_0/q^2 )</td>
</tr>
<tr>
<td></td>
<td>( \approx 3.6f_#^2\lambda_0 )</td>
<td></td>
</tr>
</tbody>
</table>

Table 3: Summary of typical longitudinal and transversal sizes of speckles in a RPP pattern, generated by different methods in the near field.