

Stochastic fluctuations

of bosonic dark matter and statistical inference



G. Centers¹ The CASPER Collaboration

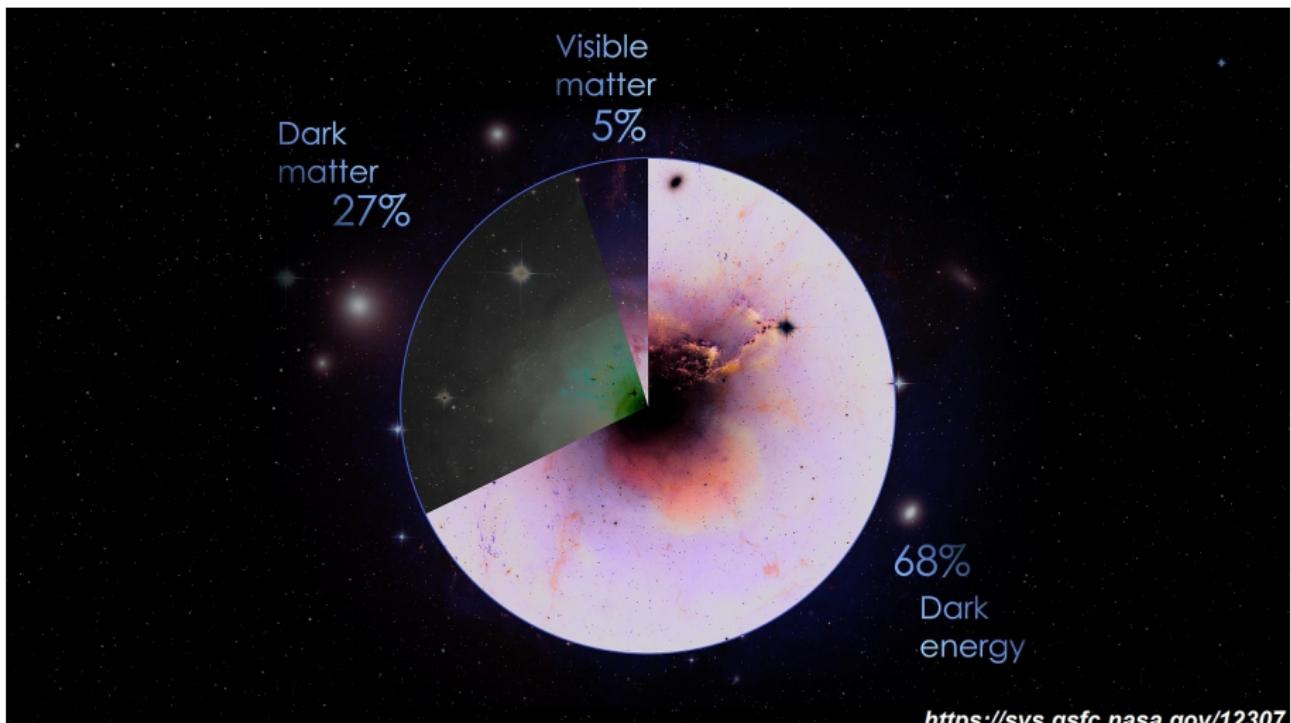
¹Helmholtz Institut Mainz, Johannes Gutenberg Universität, 55128 Mainz, Germany

Fuzzy Dark Matter Workshop, 2020



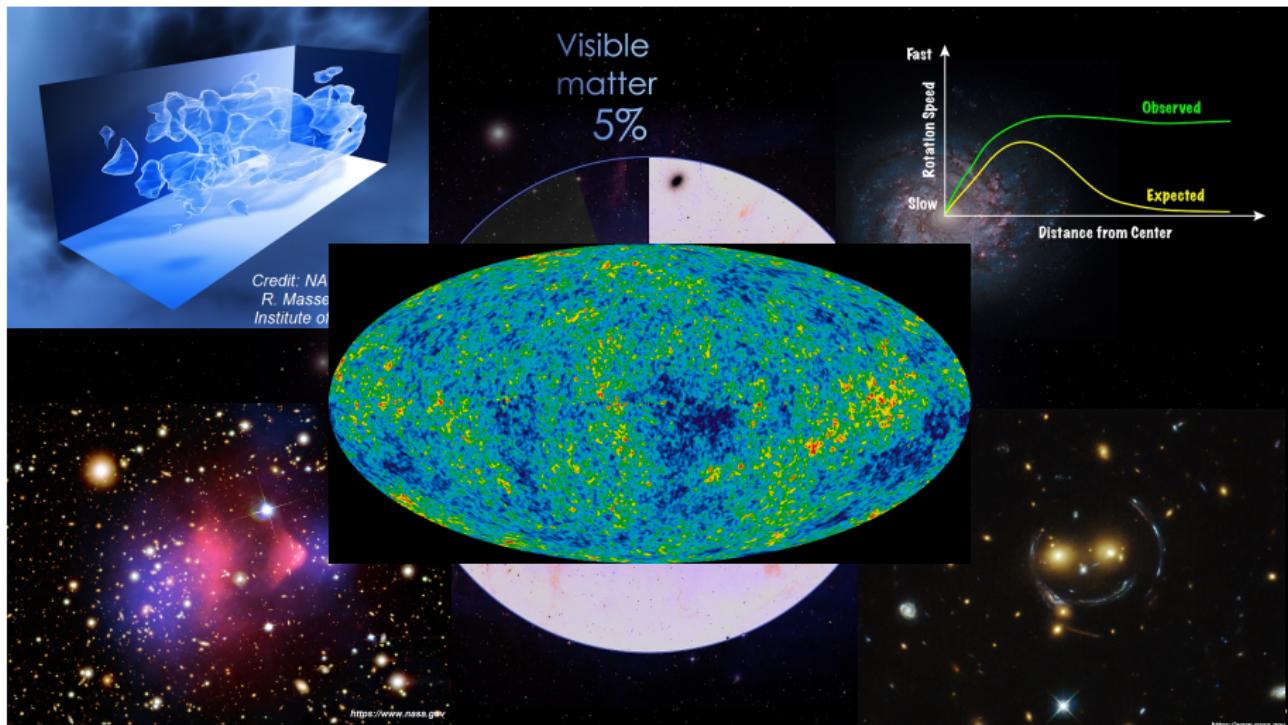
The Dark Matter

It's there, but what is it?



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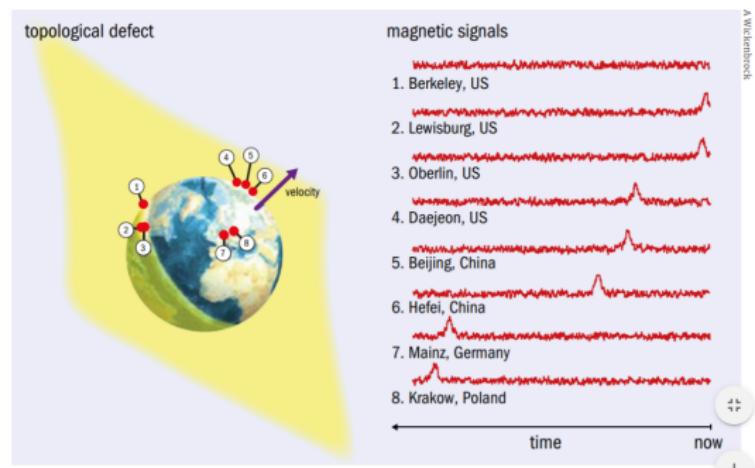
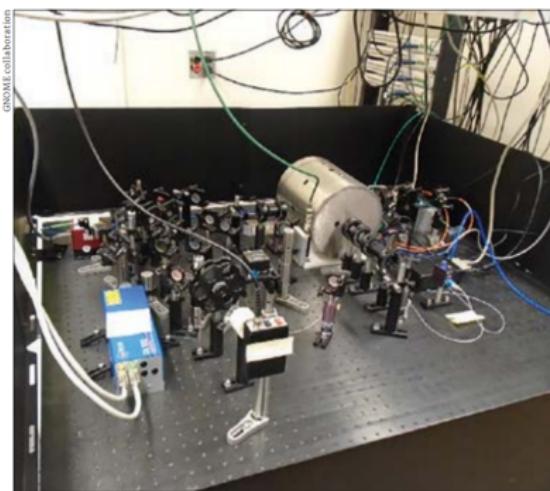
DM Constraints (see all publications at budker.uni-mainz.de)

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Global Network of Optical Magnetometers for Exotic physics searches (GNOME):

- Masia-Roig et al. [10.1016/j.dark.2020.100494](https://doi.org/10.1016/j.dark.2020.100494)
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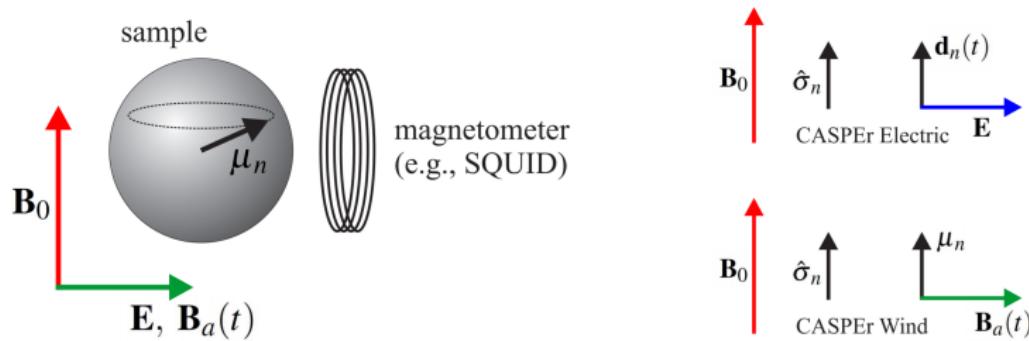
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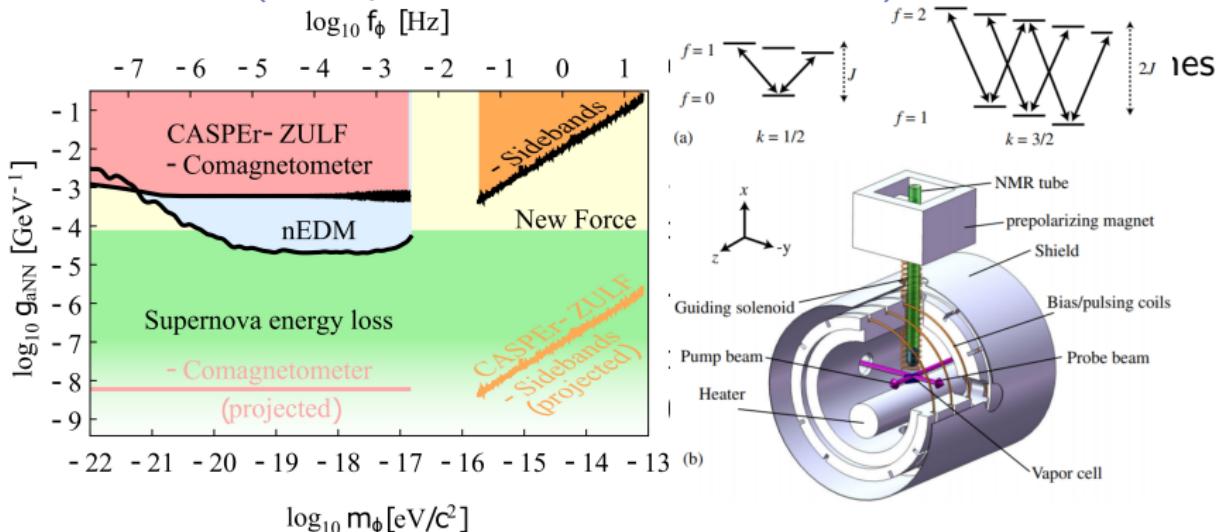
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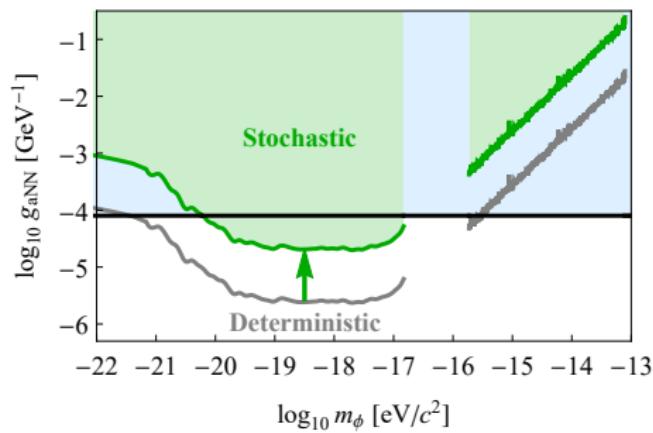
Bosonic Dark Matter

Stochastic fluctuations summary

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"Stochastic amplitude fluctuations and revised constraints on linear couplings" arxiv:1905.13650:

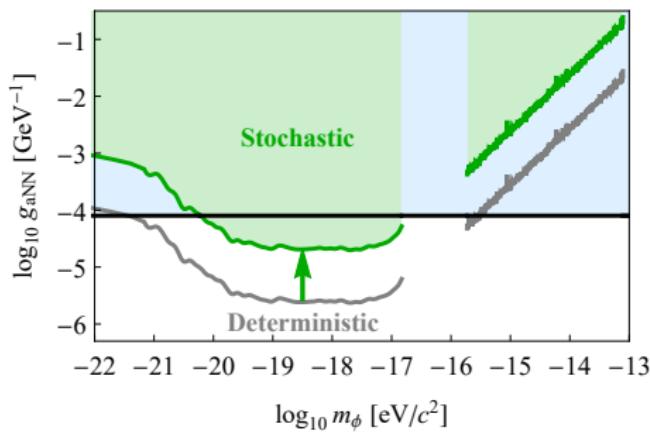


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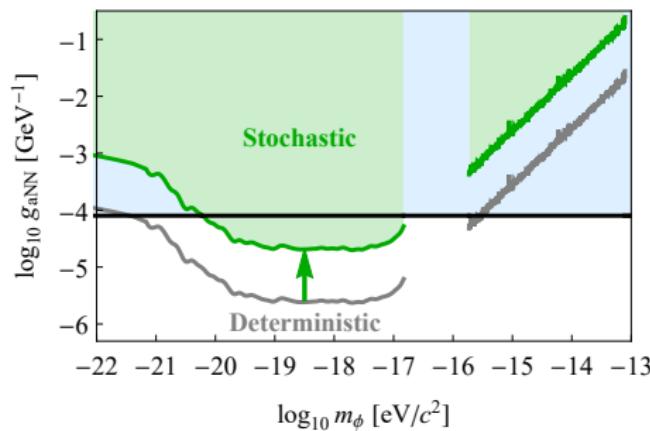


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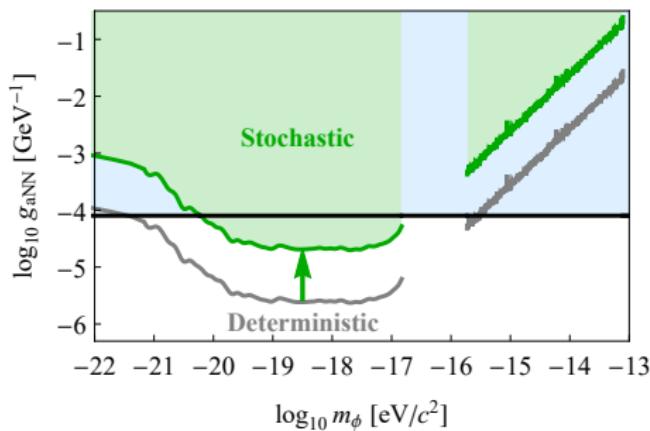


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Bosonic Dark Matter

Virialized UltraLight Fields (VULFs)

For $T \ll \tau_c$ where $\tau_c \equiv (f_c v_{\text{vir}}^2 / c^2)^{-1}$ with $f_c = m_\phi c^2 h^{-1}$:

Bosonic Dark Matter

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$$\text{scalar: } s(t) = \gamma \xi \phi(t) = \gamma \xi \Phi_0 \cos(2\pi f_\phi t + \theta)$$

$$\text{pseudoscalar: } s(t) = \gamma \xi \Phi_0 m_\phi \cos(2\pi f_\phi t + \theta) \vec{v} \cdot \vec{e}$$

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$$\boxed{\Phi_0 \xrightarrow{?} \Phi_{\text{DM}} = \hbar(m_\phi c)^{-1} \sqrt{2\rho_{\text{DM}}}}$$

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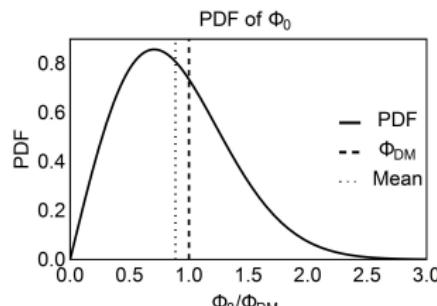
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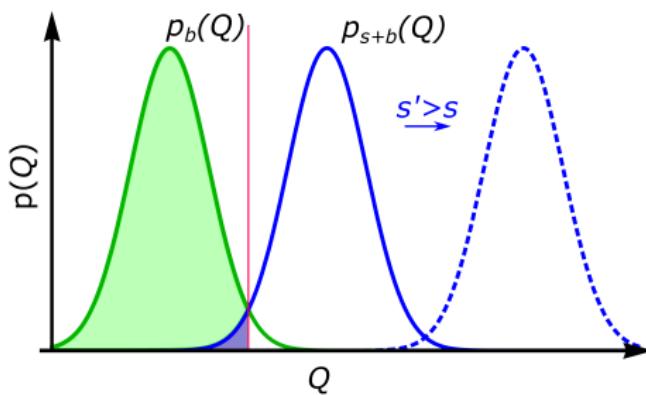
$$\Phi_0 \rightarrow \Phi_{\text{DM}} = \hbar(m_\phi c)^{-1} \sqrt{2\rho_{\text{DM}}}$$

$$\rightarrow p(\Phi_0 | \Phi_{\text{DM}}) = \begin{cases} \delta(\Phi_0 - \Phi_{\text{DM}}), & \text{deterministic} \\ \frac{2\Phi_0}{\Phi_{\text{DM}}^2} \exp\left(-\frac{\Phi_0^2}{\Phi_{\text{DM}}^2}\right), & \text{stochastic} \end{cases}$$



Comparing one sides limits

General/frequentist hypothesis testing



- Red line is “detection threshold,” defined by choice of α the Type-I error (false-positive rate under null hypothesis)
- Excluded parameter space for $s' > s$ when Type-II error $\beta < \alpha$ (false-negative rate given alternative hypothesis)

Comparing one sides limits

Lindley's paradox

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Approach	$\gamma_{95\%}^{\text{stoch}} / \gamma_{95\%}^{\text{det}}$
Frequentist	≈ 3
Bayesian	≈ 10

- Expected? Bayesian and frequentist inference try to answer different questions - Lindley's paradox

Bayesian inference

Basic probability

Conditional probability, Bayes Theorem:

$$p(B|A) = p(A|B)p(B)/p(A)$$

Bayesian inference

Basic probability

Conditional probability, Bayes Theorem:

$$p(\gamma|D) = \mathcal{L}(D|\gamma)p(\gamma)/p(D)$$

- $\mathcal{L}(D|\gamma)$ is the likelihood
- $p(\gamma)$ is the prior probability distribution
- $p(\gamma|D)$ is the posterior distribution for γ
- $p(D)$ can be treated as a normalization constant

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- $\mathcal{L}(D|\gamma)$ is the likelihood
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- $p(\gamma|D)$ is the posterior distribution for γ
- $p(D)$ can be treated as a normalization constant
- Bayesian vs. frequentist inference

Bayesian inference

Nuisance parameters

How to deal with them?

$$p(\gamma|D) = \mathcal{L}(D|\gamma)p(\gamma)/p(D)$$

Bayesian inference

Nuisance parameters

How to deal with them?

$$p(\gamma, \Phi_0 | D) = \mathcal{L}(D | \gamma, \Phi_0) p(\gamma, \Phi_0) / p(D)$$

$$p(\gamma | D) = \int d\Phi_0 p(\gamma, \Phi_0 | D)$$

- Straightforward approach within Bayesian framework
- There are several frequentist approaches: profile likelihood, marginalized likelihood, MC based approaches, etc.

The uniform prior

Something curious

- Results depend on choice of variable (Data in power vs. amplitude e.g.)
- Posterior is improper (divergent normalization) when working in power + nuisance variable marginalization
- Objective priors are one potential solution! See:
 - ▶ Kass and Wasserman, Journal of the American Statistical Association 91, 1343 (1996)
 - ▶ Berger and Bernardo (1992)

Choice of prior is critical!

Resolution of Lindley's paradox

Approach	$\gamma_{95\%}^{\text{stoch}} / \gamma_{95\%}^{\text{det}}$
Frequentist	≈ 3
Bayesian ($p(\gamma) = 1$)	≈ 10
Bayesian ($p(\gamma)$ = objective)	≈ 3

- Objective priors yield favorable properties of posterior distributions
- Using the Berger-Bernardo reference prior (similar to Jeffreys' prior but more general) Bayes/freq. agree
- For pseudoscalar coupling correction factor is up to 8.4

Acknowledgments



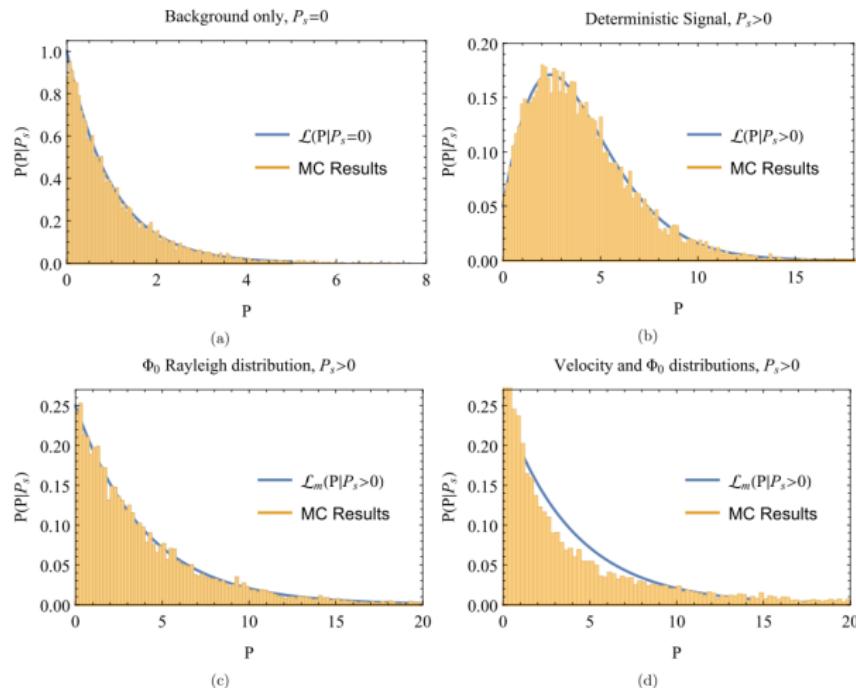
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Pseudoscalar case

Brute-force Monte Carlo

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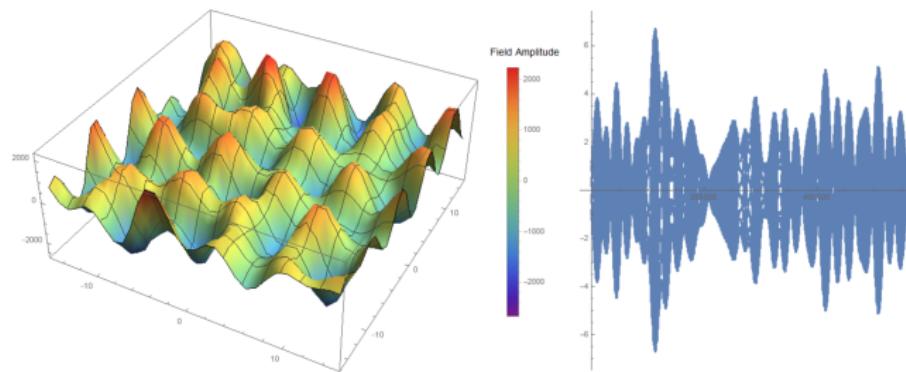


Correction factor up to 8.4

Chaotic light analogy

Rayleigh distribution and random phase

- Add N plane waves in a box: $\sum_{j=0}^N \exp -i(\omega_a t + \vec{k}_j \cdot \vec{x}_j + \phi)$
- Resulting field amplitude is just a random walk in the complex plane
- Quick simulation: $\omega_a = 2\pi$, $N = 1000$, $|v|$ is MB dist. $v_{avg} = 10^{-3}$



See Foster/Safdi 2018