

An introduction to extreme value statistics (part 2/2)

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Dresden
October 06, 2015

Range of applications

Gallery of correlations

Illustrations of EVS in statistical physics

- DNA replication times in *Xenopus Laevis*

- Order parameter in percolation

- EVS in trees

- EVS of some Gaussian processes

- Integer partitions and the ideal Bose gas

Range of applications

Overview/statistical physics

J.-P. Bouchaud, M. Potters

Theory of Financial Risks

Cambridge University Press (2000)

Introduction

D. Sornette

Critical Phenomena in Natural Sciences

Springer (2006)

Introduction

M. Clusel, E. Bertin

Int. J. Mod. Phys. B **22**, 3311 (2008)

Review

S. N. Majumdar, A. Pal

arXiv:1406.6768v3 (2014)

Lecture notes

J.-Y. Fortin, M. Clusel

J. Phys. A: Math. Theor. **48**, 183001 (2015)

Review

Range of applications

Modelling

R. W. Katz, M. B. Parlange, P. Naveau
Adv. Water Resour. **25**, 1287 (2002)

Hydrology

A. J. McNeil, R. Frey
J. Empir. Financ. **7**, 271 (2000)

Finance

T. Antal *et al.*
Eur. Phys. Lett. **88**, 59001 (2009)

Astronomy

J. Bechhoefer, B. Marshall
Phys. Rev. Lett. **98**, 098105 (2007)

Cell biology

S. Hallerberg, J. Bröcker, H. Kantz
In Nonlinear Time Series Analysis in the Geosciences,
R. V. Donner, S. M. Barbosa (eds), Springer (2008)

Prediction/Forecasting

Range of applications

Disordered systems

J.-P. Bouchaud, M. Mézard

J. Phys. A: Math. Gen. **30**, 7997 (1997)

Random energy model

S. N. Majumdar, P. L. Krapivsky

Physica A **318**, 161 (2003)

Hierarchical correlations

G. Biroli, J.-P. Bouchaud, M. Potters

J. Stat. Mech. P07019 (2007)

Disordered systems

Y. V. Fyodorov

Physica A **389**, 4229 (2010)

Logarithmic correlations

S. N. Majumdar, G. Schehr

J. Stat. Mech. P01012 (2014)

Random matrices

Range of applications

Dynamical systems

C. Nicolis, V. Balakrishnan, G. Nicolis
Phys. Rev. Lett. 210602 (2006)

Extreme events in chaotic dynamics

M. Ghil *et al.*
Nonlin. Processes Geophys. **18**, 295 (2011)

Review

J. M. Freitas
Dynamical Systems **28**, 302 (2013)

EVS in chaotic dynamics

H. Aytaç, J. M. Freitas, S. Vaienti
Trans. Amer. Math. Soc. **367**, 8229 (2015)

EVS in deterministic and random
dynamical systems

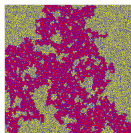
Gallery of correlations

Time series



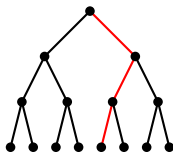
$$\langle x(t')x(t'+t) \rangle \sim t^{-\gamma}$$

Critical phenomena



$$g(r) \sim r^{-(d-2+\eta)}$$

Trees



$$g(i,j) \propto \text{length of shared path}$$

Random matrices

$$\begin{pmatrix} a_{11} & a_{12} & \cdots \\ a_{21} & a_{22} & \cdots \\ \vdots & \vdots & \ddots \end{pmatrix}$$

$$P[\lambda_i] \propto \exp \left[-\beta \left(\frac{N}{2} \sum_{i=1}^N \lambda_i^2 - \frac{1}{2} \sum_{i \neq j} \ln |\lambda_i - \lambda_j| \right) \right]$$

Outline

Range of applications

Gallery of correlations

Illustrations of EVS in statistical physics

DNA replication times in *Xenopus Laevis*

Order parameter in percolation

EVS in trees

EVS of some Gaussian processes

Integer partitions and the ideal Bose gas

DNA replication in *Xenopus Laevis*

Bechhoefer, Marshall 2007

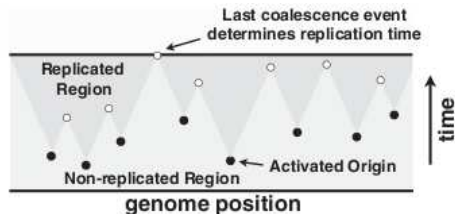


FIG. 1. Schematic of DNA replication model. Space-time diagram showing multiple origins (filled circles), each expanding symmetrically at constant velocity. Domains coalesce when they meet (open circles).

- $\sim 3 \times 10^9$ bases
- $\sim 10^5$ replication origins
- $\sim 1\text{kb}/\text{min}$ fork velocity
- ~ 20 min replication time
- $\sim 10^{-4}$ failure probability

DNA replication in *Xenopus Laevis*

Bechhoefer, Marshall 2007

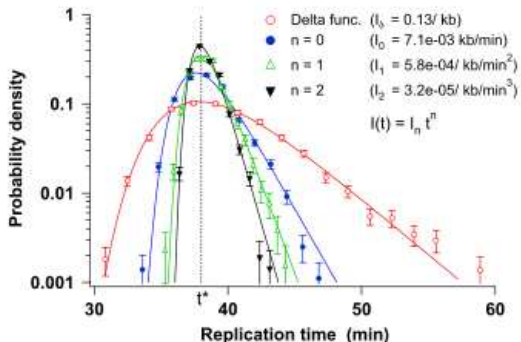


FIG. 2 (color online). Replication-time distribution function, fixing the mode to be $t^* = 38$ minutes. Markers are results from Monte Carlo simulations (3000 trials per simulation); solid lines are fits to the Gumbel distribution.

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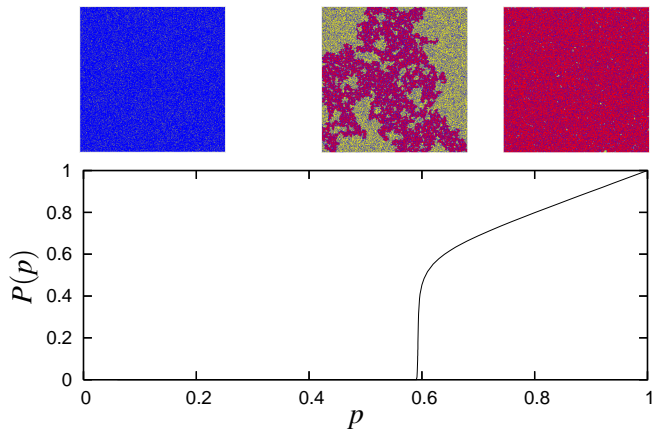
Order parameter in percolation

EVS in trees

EVS of some Gaussian processes

Integer partitions and the ideal Bose gas

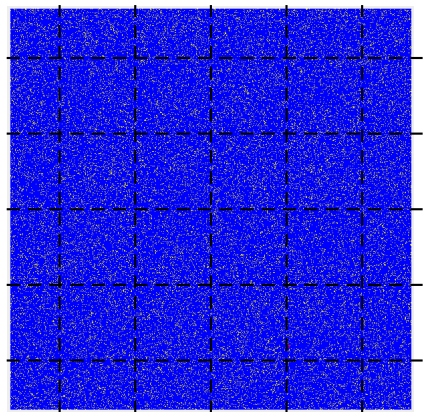
Percolation



Fraction occupied by the largest cluster is the **order parameter**.

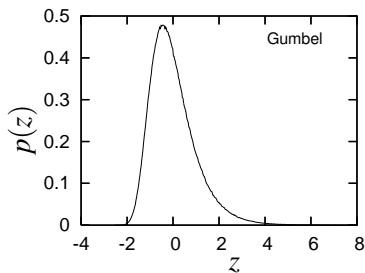
Subcritical percolation ($p \ll p_c$)

Duxbury, Leath 1987; Bazant 2000; Borgs et al. 2001; van der Hofstad, Redig 2006



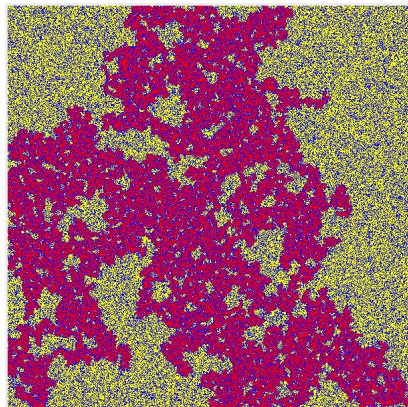
largest cluster $\sim \log L$

overall largest cluster is
the largest cluster from
uncorrelated blocks



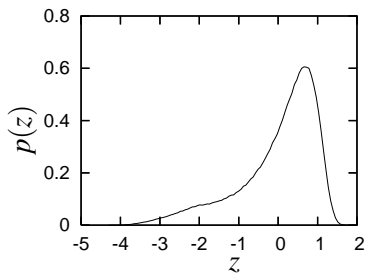
Critical percolation ($p = p_c$)

Hovi, Aharony 1996; Sen 2001; Borgs et al. 2001



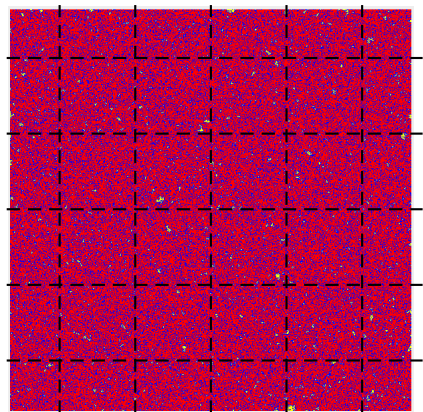
largest cluster $\sim L^D$ ($D = 91/48$)

correlations span the system!
non-trivial distribution emerges



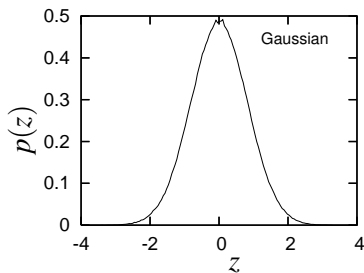
Supercritical percolation ($p \gg p_c$)

Borgs et al. 2001



largest cluster $\sim L^d$

largest cluster is extensive:
sum up contributions from
uncorrelated blocks



Percolation

P. M. Duxbury, P. L. Leath

J. Phys. A: Math. Gen. **20**, L411 (1987)

Largest cluster in subcritical percolation

J.-P. Hovi, A. Aharony

Phys. Rev. E **53**, 235 (1996)

Clusters in critical percolation

M. Z. Bazant

Phys. Rev. E **62**, 1660 (2000)

Largest cluster in subcritical percolation

P. Sen

J. Phys. A: Math. Gen. **34**, 8477 (2001)

Largest clusters in critical percolation

C. Borgs *et al.*

Commun. Math. Phys. **224**, 153 (2001)

Finite-size scaling of largest cluster

R. van der Hofstad, F. Redig

J. Stat. Phys. **122**, 671 (2006)

Largest clusters in non-critical percolation

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DNA replication times in *Xenopus Laevis*

Order parameter in percolation

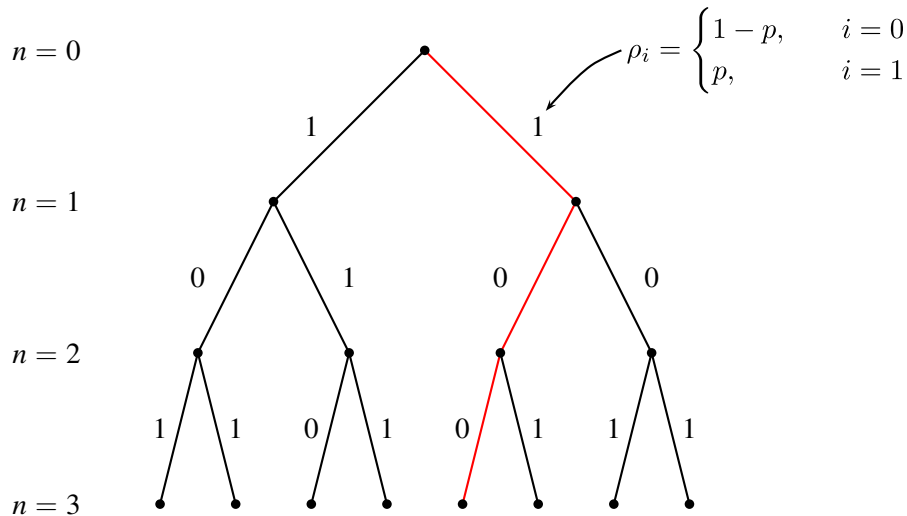
EVS in trees

EVS of some Gaussian processes

Integer partitions and the ideal Bose gas

EVS in trees

Majumdar, Krapivsky 2000



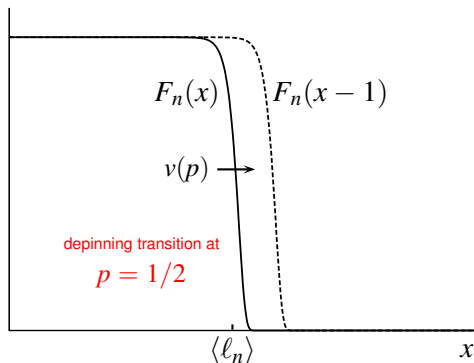
$\ell_3 = \text{shortest path (at depth 3)} = 1 + 0 + 0 = 1$

EVS in trees

Analogy between 1-0 behaviour and wave fronts (c.f. $\partial_t \phi = \partial_{xx} \phi + \phi - \phi^2$)

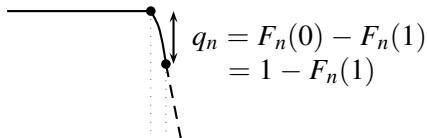
$$\mathbb{P}(\ell_{n+1} \geq x) = F_{n+1}(x) = [(1-p)F_n(x) + pF_n(x-1)]^2$$

$$\text{initial condition: } F_0(x) = \begin{cases} 1, & x \leq 0 \\ 0, & x > 0 \end{cases}$$



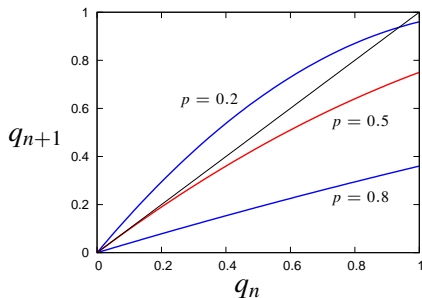
EVS in trees

Tracking the tip of the front



Substitute $q_n = 1 - F_n(1)$ into recursion relation for F_n :

$$q_{n+1} = 2(1-p)q_n - (1-p)^2 q_n^2$$



EVS in trees

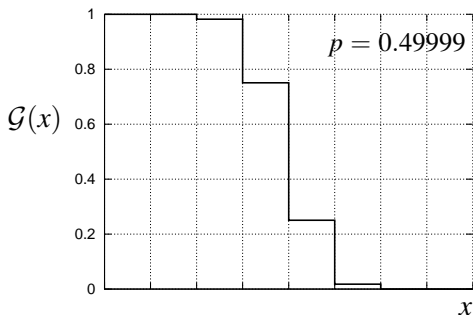
$p < 1/2$: wave front is 'pinned'

Wave front gets pinned at a finite $\langle \ell_n \rangle$ as $n \rightarrow \infty$.

$F_n(x)$ approaches a stationary distribution:

$$\mathcal{G}(x) = [(1 - p)\mathcal{G}(x) + p\mathcal{G}(x - 1)]^2$$

Iterate, starting with $\mathcal{G}(0) = 1$, to obtain $\mathcal{G}(1), \mathcal{G}(2), \dots$



EVS in trees

$p \geq 1/2$: wave front is 'depinned'

Wave front travels as a fixed shape \mathcal{F}^2 located at $\langle \ell_n \rangle \approx v(p)n$:

$$F_{n+1}(x) \rightarrow \mathcal{F}^2(x - v(n+1))$$

$$F_n(x) \rightarrow \mathcal{F}^2(x - vn)$$

$$F_n(x-1) \rightarrow \mathcal{F}^2(x - v - 1)$$

Substitute into recursion relation:

$$\mathcal{F}(z - v) = (1 - p)\mathcal{F}^2(z) + p\mathcal{F}^2(z - 1), \quad z = x - vn.$$

Behind the front, $1 - \mathcal{F}(z) \approx e^{\lambda z}$. Substitute and linearise:

$$v_\lambda(p) = -\frac{\log[2(1-p) + 2pe^{-\lambda}]}{\lambda}.$$

Front selects $v_{\lambda^*}(p)$.

EVS in trees

Bramson 1978; Brunet, Derrida 1997; Ebert, van Saarloos 1998; Majumdar, Krapivsky 2000

(Majumdar & Krapivsky 2000)

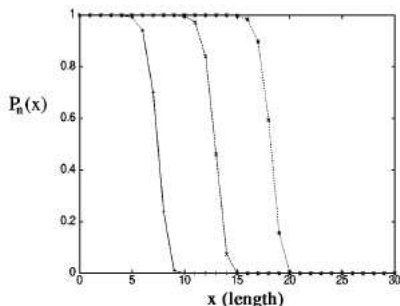


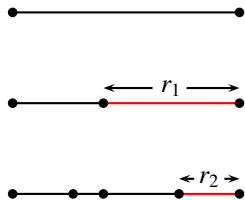
FIG. 1. The propagating front for the cumulative distribution $P_n(x)$ of the minimal length for the bimodal distribution with $p=0.8$.

$$\langle \ell_n \rangle = \begin{cases} c(p), & p < 1/2 \\ (\log 2)^{-1} \log \log n, & p = 1/2 \\ v(p)n, & p > 1/2. \end{cases}$$

EVS in trees

Hierarchical structure is rather generic...

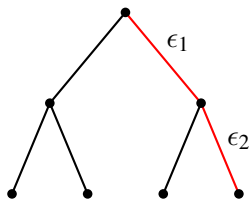
fragmentation process



$$\ell_k^{(n)} = \prod_{i=1}^n r_i$$

$$-\log r_i = \epsilon_i$$

polymer/tree



$$E_k^{(n)} = \sum_{i=1}^n \epsilon_i$$

$$\mathbb{P}(\min_k E_k^{(n+1)} > x) = F_{n+1}(x) = \left[\int_0^\infty d\epsilon F_n(x - \epsilon) \rho(\epsilon) \right]^2$$

- ▶ Write down recursion relation between F_n and F_{n+1} .
- ▶ Identify conditions for travelling wave solution.
- ▶ Linearise wave equation.
- ▶ Calculate wave speeds, etc.

EVS in trees

S. N. Majumdar, P. L. Krapivksy

Phys. Rev. E **62**, 7735 (2000)

Extremal paths on random Cayley trees

E. Ben-Naim, P. L. Krapivksy, S. N. Majumdar

Phys. Rev. E **64**, 035101(R) (2001)

Extremal paths on random Cayley trees

D. S. Dean, S. N. Majumdar

Phys. Rev. E **64**, 046121 (2001)

Polymers on random binary trees

S. N. Majumdar, P. L. Krapivksy

Phys. Rev. E **65**, 036127 (2002)

Polymers, binary search trees

D. S. Dean, S. N. Majumdar

J. Phys. A: Math. Gen. **35**, L501 (2002)

Fragmentation, m -ary search

S. N. Majumdar, P. L. Krapivksy

Physica A **318**, 161 (2003)

Overview article

Outline

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DNA replication times in *Xenopus Laevis*

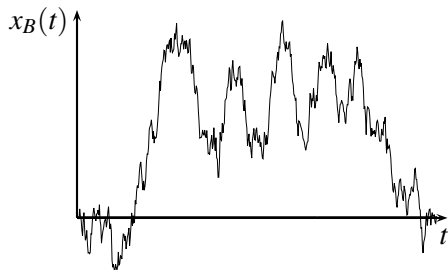
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Maximum of a Brownian bridge



$$\langle X(s)X(t) \rangle = \min(s, t) - st$$

(Berman's condition fails)

$$\mathbb{P}(\max_t x_B(t) \leq m) = \frac{W_m(x_t = 0, t | x_0 = 0, 0)}{W(x_t = 0, t | x_0 = 0, 0)},$$

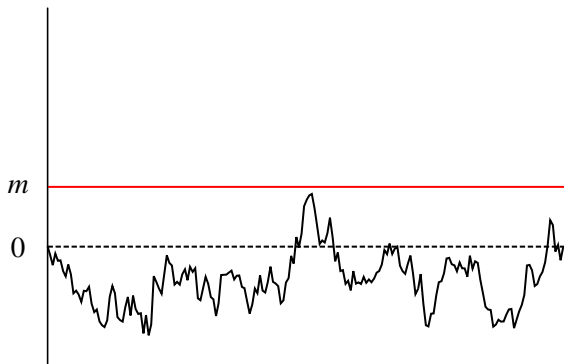
where transition probability $W_m(x_t, t | x_0 = 0, 0)$ satisfies

$$\left(\frac{\partial}{\partial t} - \frac{1}{2} \frac{\partial^2}{\partial x^2} + V_m(x) \right) W_m = 0$$

$$V_m(x) = \begin{cases} 0, & x < m, \\ \infty, & x \geq m. \end{cases}$$

Maximum of a Brownian bridge

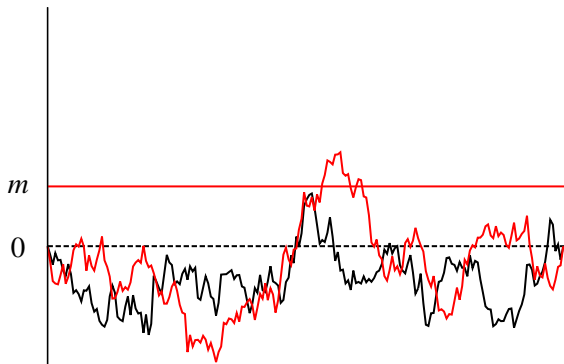
Reflection principle



$$\begin{aligned} W(x_t = 0, t | x_0 = 0, 0) &= \frac{1}{\sqrt{2\pi t}} \exp(-x_t^2/2t) \Big|_{x_t=0} \\ &= \frac{1}{\sqrt{2\pi t}} \end{aligned}$$

Maximum of a Brownian bridge

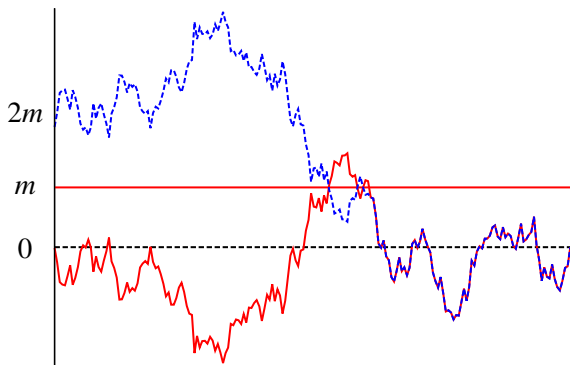
Reflection principle



$$W_m(0, t | 0, 0) = W(0, t | 0, 0) - \text{overcounting term}$$

Maximum of a Brownian bridge

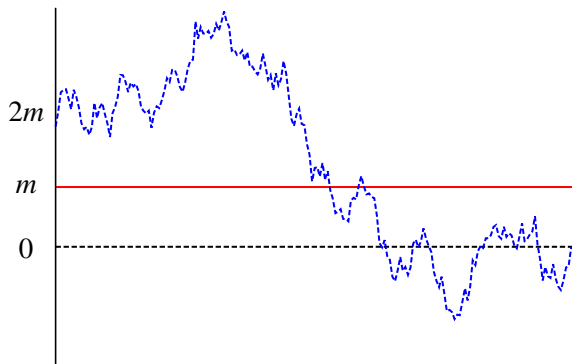
Reflection principle



$$W_m(0, t | 0, 0) = W(0, t | 0, 0) - \text{overcounting term}$$

Maximum of a Brownian bridge

Reflection principle



$$\begin{aligned}W_m(0, t | 0, 0) &= W(0, t | 0, 0) - W(2m, t | 0, 0) \\ &= \frac{1}{\sqrt{2\pi t}} \left(1 - e^{-2m^2/t}\right)\end{aligned}$$

$$\mathbb{P}(\max_t x_B(t) \leq m) = 1 - e^{-2m^2/t}$$

Mean-subtracted Brownian bridge

Watson 1961

Define a new Gaussian process by subtracting the bridge's mean:

$$x_W(t) := x_B(t) - \int_0^t dt' x_B(t').$$

'Watson' bridge has correlations

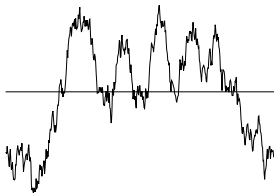
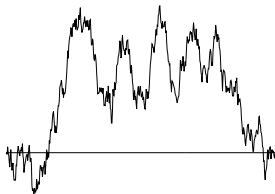
$$\langle X(s)X(t) \rangle = \min(s, t) - \frac{1}{2}(s + t) + \frac{1}{2}(s - t)^2 + \frac{1}{12}.$$

This depends only on time differences:

$$\langle X(s)X(s + t) \rangle = \frac{1}{12} [1 - 6|t|(1 - |t|)].$$

$x_B(t)$ $x_W(t)$

realisation



local variance

 $t(1-t)$ $1/12$

Fourier decomposition

 $a_k, b_k \stackrel{d}{\sim} N(0, 1)$

$$\frac{\sqrt{2}}{\pi} \sum_{k=1}^{\infty} \frac{a_k}{k} \sin \pi k t$$

$$\frac{1}{\sqrt{2}\pi} \sum_{k=1}^{\infty} \frac{1}{k} (a_k \sin 2\pi k t + b_k \cos 2\pi k t)$$

Periodic Gaussian $1/f^\alpha$ noise

Signal construction

$$h(t) \propto \sum_{k=1}^n \frac{1}{k^{\alpha/2}} \left[a_k \cos\left(\frac{2\pi k}{T}t\right) + b_k \sin\left(\frac{2\pi k}{T}t\right) \right], \quad a_k, b_k \stackrel{d}{\sim} N(0, 1)$$

white
noise

$1/f$
noise

Edwards
Wilkinson

Mullins
Herring

single
mode



$\alpha = 0$

1

2

4

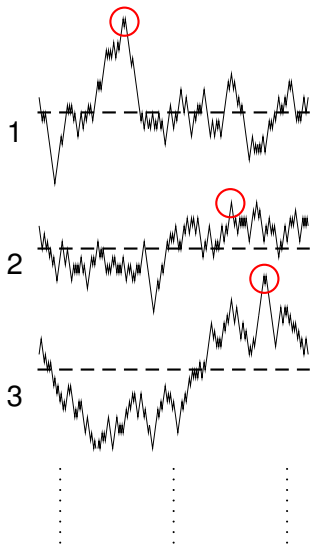
$\alpha \rightarrow \infty$

weaker
← correlations

stronger
→ correlations

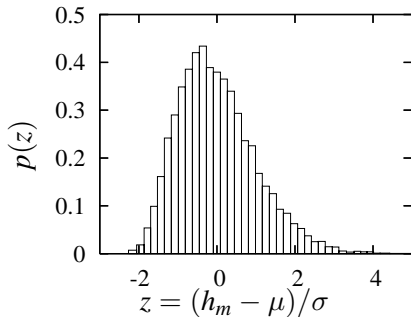
Periodic Gaussian $1/f^\alpha$ noise

Maximum relative height: Raychaudhuri *et al.* 2001



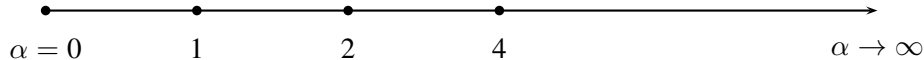
build
→
histogram

$$h_m := \max_t h(t) - \overline{h(t)}$$



Periodic Gaussian $1/f^\alpha$ noise

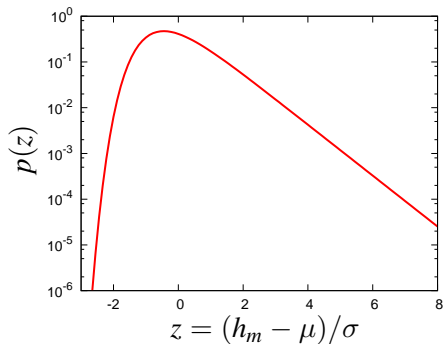
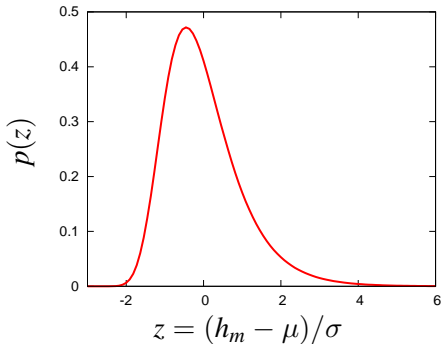
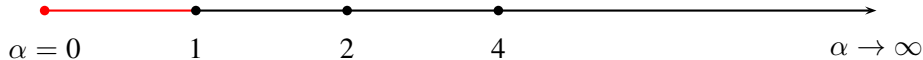
Maximum relative height



Periodic Gaussian $1/f^\alpha$ noise

Maximum relative height

Berman 1964



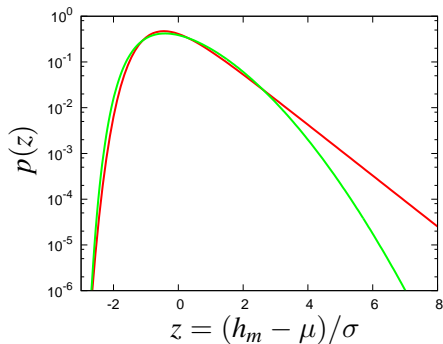
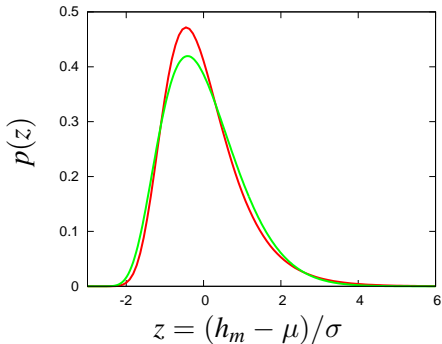
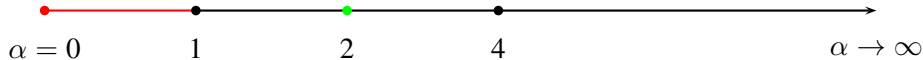
Periodic Gaussian $1/f^\alpha$ noise

Maximum relative height

Berman 1964



Majumdar 2004
Comtet



Periodic Gaussian $1/f^\alpha$ noise

Maximum relative height

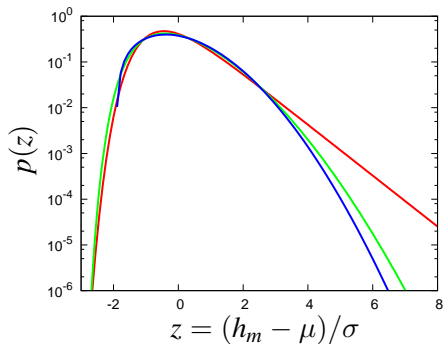
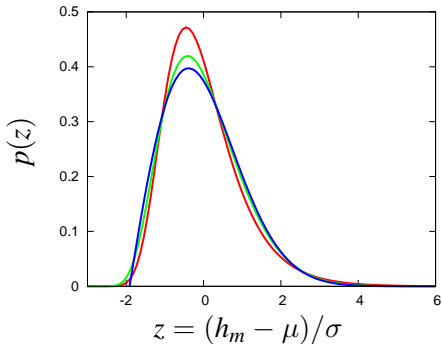
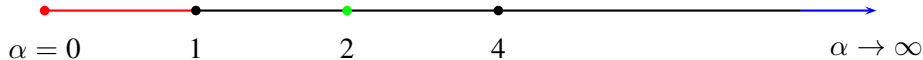
Berman 1964



Majumdar 2004
Comtet



single
mode



Periodic Gaussian $1/f^\alpha$ noise

Maximum relative height

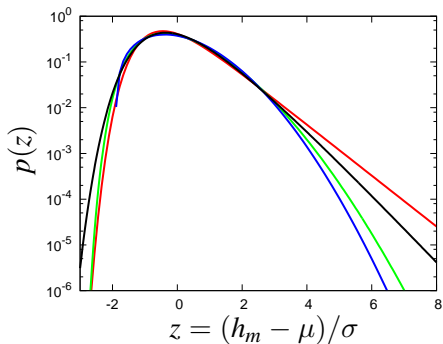
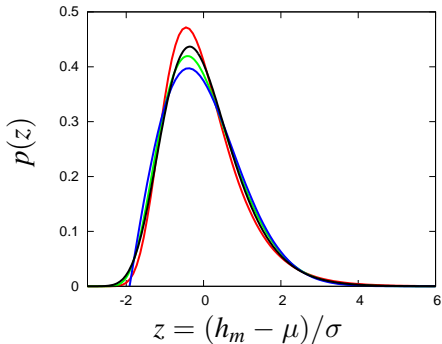
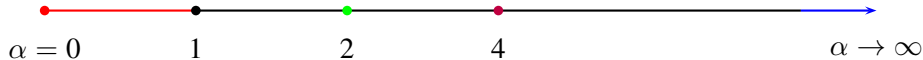
Berman 1964

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single
mode



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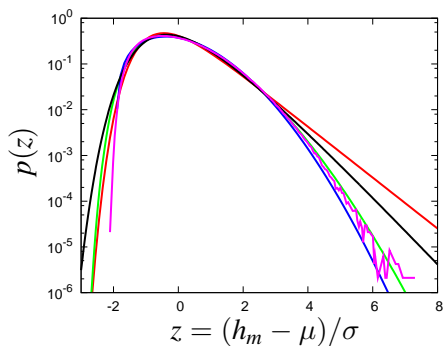
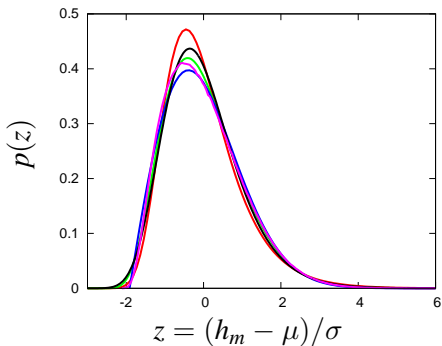
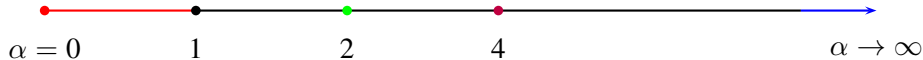
Berman 1964

Fyodorov 2008
Bouchaud

Majumdar 2004
Comtet

numerics

single
mode



Extremes of Gaussian processes

S. M. Berman

Ann. Math. Stat. **33**, 502 (1964)

Berman's condition

P. Biane, J. Pitman, M. Yor

Bull. Amer. Math. Soc **38**, 435 (2001)

Review

S. N. Majumdar, A. Comtet

J. Stat. Phys. **119**, 777 (2005)

MRH for Edwards-Wilkinson ($1/f^2$)

G. Györgyi *et al.*

Phys. Rev. E **75**, 021123 (2007)

MRH for $1/f^\alpha$ noise

Y. V. Fyodorov, J.-P. Bouchaud

J. Phys. A: Math. Theor. **41**, 372001 (2008)

MRH for $1/f$ noise

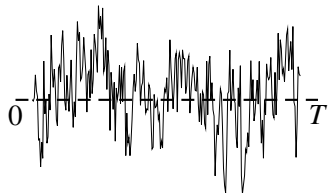
H. J. Hilhorst, P. Calka, G. Schehr

J. Stat. Mech. P10010 (2008)

Random acceleration process ($1/f^4$)

Roughness of $1/f$ noise

Antal *et al.* 2001



$$w^2 := \frac{1}{T} \int_0^T dt [h(t) - \overline{h(t)}]^2 \propto \sum_{k=1}^n \frac{1}{k} (a_k^2 + b_k^2)$$
$$\propto \sum_{k=1}^n \frac{\varepsilon_k}{k},$$

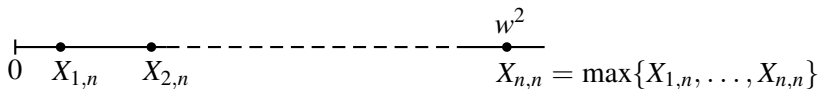
where ε_k are standard exponential variables.

Antal *et al.* observed that

$$w^2 \stackrel{d}{\sim} \text{Gumbel!}$$

Roughness of $1/f$ noise

Rényi 1953; Bertin 2005



Interpret the partial sums

$$X_{m,n} := \sum_{k=1}^m \frac{\varepsilon_k}{n - k + 1}, \quad m \leq n$$

as marking the positions of a collection of ordered points $X_{m,n}$.
In particular

$$w^2 = X_{n,n} = \max\{X_{1,n}, \dots, X_{n,n}\}.$$

Roughness of $1/f$ noise

Rényi 1953; Bertin 2005

Starting from a collection of iid standard exponentials

$$f_{\varepsilon_1, \dots, \varepsilon_n}(y_1, \dots, y_n) = \prod_{k=1}^n \exp(-y_k) = \exp\left(-\sum_{k=1}^n y_k\right),$$

change variables with

$$x_m = \sum_{k=1}^m \frac{y_k}{n-k+1}, \quad \left| \frac{\partial y_i}{\partial x_j} \right| = n!$$

to obtain

$$f_{X_1, \dots, X_n}(x_1, \dots, x_n) = n! \exp\left(-\sum_{k=1}^n x_k\right) \mathbf{1}_{\{x_1 < \dots < x_n\}}$$

$$f_{X_1, \dots, X_n}(x_1, \dots, x_n) = \exp\left(-\sum_{k=1}^n x_k\right) = \prod_{k=1}^n \exp(-x_k).$$

Roughness of $1/f$ noise

Rényi 1953; Bertin 2005

Rényi's representation

$$\{X_{m,n}, m = 1, \dots, n\} \stackrel{d}{\sim} \left\{ \sum_{k=1}^m \frac{\varepsilon_k}{n - k + 1}, m = 1, \dots, n \right\}$$

implies roughness of $1/f$ noise

$$w^2 = \max\{X_1, \dots, X_n\} = X_{n,n} \stackrel{d}{\sim} \text{Gumbel},$$

where X_k, ε_k are standard exponentials.

Roughness of $1/f$ noise

A. Rényi

Acta Mathematica Scient. Hungar. **IV**, 191 (1953)

Order statistics

T. Antal *et al.*

Phys. Rev. Lett. **87**, 240601 (2001)

Roughness of $1/f$ noise

E. Bertin

Phys. Rev. Lett. **95**, 170601 (2005)

Order statistics, global fluctuations

E. Bertin, M. Clusel

J. Phys. A: Math. Gen. **39**, 7607 (2006)

Order statistics, global fluctuations

Outline

Range of applications

Gallery of correlations

Illustrations of EVS in statistical physics

DNA replication times in *Xenopus Laevis*

Order parameter in percolation

EVS in trees

EVS of some Gaussian processes

Integer partitions and the ideal Bose gas

Integer partitions and the ideal Bose gas

$\Omega(E)$ = number of ways of partitioning an integer E into a sum of (indistinguishable) positive integers.

E.g. $\Omega(5) = 7$:

5

4 + 1

3 + 2

3 + 1 + 1

2 + 2 + 1

2 + 1 + 1 + 1

1 + 1 + 1 + 1 + 1

Integer partitions and the ideal Bose gas

$\Omega(E)$ = number of ways of partitioning an integer E into a sum of (indistinguishable) positive integers.

E.g. $\Omega(5) = 7$:

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4 + 1

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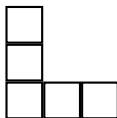
3 + 1 + 1

2 + 2 + 1

2 + 1 + 1 + 1

1 + 1 + 1 + 1 + 1

summands
expansion



Young
diagram

| ϵ_k | | n_k |
|--------------|--------------|-------|
| ϵ_5 | | 0 |
| ϵ_4 | | 0 |
| ϵ_3 | ● | 1 |
| ϵ_2 | | 0 |
| ϵ_1 | ● ● | 2 |
| ϵ_0 | ground state | n_0 |

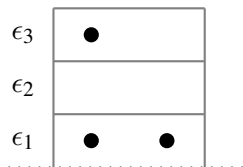
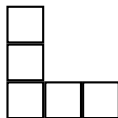
Bose gas

Integer partitions and the ideal Bose gas

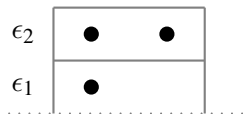
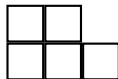
$\Omega(E, N)$ = number of ways of partitioning E with N integers.

E.g. $\Omega(E = 5, N = 3) = 2$:

$$3 + 1 + 1$$



$$2 + 2 + 1$$



summands
expansion

Young
diagram

Bose gas

Integer partitions and the ideal Bose gas

Hardy, Ramanujan 1918; van Lier, Uhlenbeck 1937; Erdős, Lehner 1941; Auluck, Kothari 1946

- ▶ Number of partitions grows rapidly
(e.g. $\Omega(E = 1000) \approx 2.4 \times 10^{31}$).

$$\Omega(E) \sim \frac{1}{4\sqrt{3}E} \exp\left(\pi\sqrt{\frac{2}{3}}E^{1/2}\right).$$

- ▶ Probability of partitioning E with N integers converges ($E, N \gg 1$) to **Gumbel distribution!**

$$\frac{\sum_{N'=1}^N \Omega(E, N')}{\Omega(E)} \sim \exp\left[-\exp\left(\frac{N - b_E}{a_E}\right)\right]$$

$$a_E = \frac{\sqrt{6}}{\pi} E^{1/2}$$

$$b_E = \frac{1}{\pi} \sqrt{\frac{3}{2}} E^{1/2} \log E.$$

Integer partitions and the ideal Bose gas

Auluck, Kothari 1946; Comtet, Leboeuf, Majumdar 2007

$$\Omega(E) = \sum_{\{n_k\}} \delta \left(E - \sum_{k=1}^{\infty} n_k \epsilon_k \right)$$
$$\mathcal{Z}(\beta) = \sum_E \Omega(E) e^{-\beta E} = \prod_{k=1}^{\infty} \frac{1}{1 - e^{-\beta \epsilon_k}}.$$

Saddle point approximation:

$$\Omega(E) \approx e^{S(\beta_0, E)},$$

where β_0 maximises

$$S(\beta, E) = \log \mathcal{Z}(\beta) + \beta E.$$

Using

$$E = \sum_{k=1}^{\infty} \frac{\epsilon_k}{e^{\beta_0 \epsilon_k} - 1} \approx \int_0^{\infty} d\epsilon \frac{\epsilon}{e^{\beta_0 \epsilon} - 1} = \frac{\pi^2}{6\beta_0^2},$$

one recovers Hardy-Ramanujan:

$$\Omega(E) \approx \exp \left(\pi \sqrt{\frac{2}{3}} E^{1/2} \right).$$

Integer partitions and the ideal Bose gas

Auluck, Kothari 1946; Comtet, Leboeuf, Majumdar 2007

$$\sum_{N'=1}^N \Omega(E, N') = \tilde{\Omega}(E, N) = \sum_{\{n_k\}} \delta \left(E - \sum_{k=1}^{\infty} n_k \epsilon_k \right) \theta \left(N - \sum_{k=1}^{\infty} n_k \right)$$
$$\mathcal{Z}(\beta, z) = \sum_{E, N} \tilde{\Omega}(E, N) e^{-\beta E} z^N = \prod_{k=1}^{\infty} \frac{1}{1 - z e^{-\beta \epsilon_k}},$$

Saddle point approximation:

$$\tilde{\Omega}(E, N) \approx e^{S(\beta_0, z_0, E, N)},$$

where β_0, z_0 maximise

$$S(\beta, z, E, N) = \log \mathcal{Z}(\beta) + \beta E - N \log(z).$$

Saddle points fix

$$E = \sum_{k=1}^{\infty} \frac{\epsilon_k}{z_0^{-1} e^{\beta_0 \epsilon_k} - 1} \approx \int_0^{\infty} d\epsilon \frac{\epsilon}{z_0^{-1} e^{\beta_0 \epsilon} - 1} = \frac{\text{Li}_2(z_0)}{\beta_0^2}$$
$$N = \sum_{k=1}^{\infty} \frac{1}{z_0^{-1} e^{\beta_0 \epsilon_k} - 1} \approx \int_0^{\infty} d\epsilon \frac{1}{z_0^{-1} e^{\beta_0 \epsilon} - 1} = -\frac{\log(1 - z_0)}{\beta_0}.$$

Integer partitions and the ideal Bose gas

Auluck, Kothari 1946; Comtet, Leboeuf, Majumdar 2007

Take large N limit ($z \uparrow 1$) to recover Erdős-Lehner:

$$\frac{\tilde{\Omega}(E, N)}{\Omega(E)} \sim \exp \left[- \exp \left(\frac{N - b_E}{a_E} \right) \right]$$

$$a_E = \frac{\sqrt{6}}{\pi} E^{1/2}$$

$$b_E = \frac{1}{\pi} \sqrt{\frac{3}{2}} E^{1/2} \log E.$$

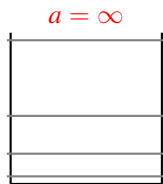
This calculations holds for equally spaced energy levels

$\epsilon_1 = 1, \epsilon_2 = 2, \dots$

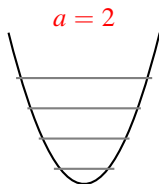
What about for $\rho(\epsilon) = \nu \epsilon^{\nu-1}$?

Integer partitions and the ideal Bose gas

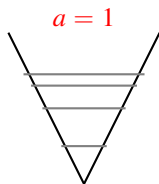
Density of states $\rho(\epsilon) = \nu \epsilon^{\nu-1}$ for different 1d potentials



$$\rho(\epsilon) \sim \epsilon^{-1/2}$$
$$(\nu = 1/2)$$



$$\rho(\epsilon) \sim \text{const.}$$
$$(\nu = 1)$$



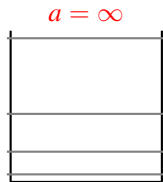
$$\rho(\epsilon) \sim \epsilon^{1/2}$$
$$(\nu = 3/2)$$

$$V(x) \sim |x|^a$$

$$\rho(\epsilon) \sim \epsilon^{(2-a)/2a}$$

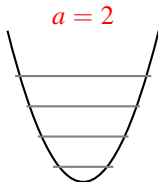
Integer partitions and the ideal Bose gas

Density of states $\rho(\epsilon) = \nu \epsilon^{\nu-1}$ for different $1d$ potentials



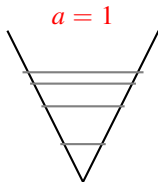
$$\rho(\epsilon) \sim \epsilon^{-1/2}$$

$$(\nu = 1/2)$$



$$\rho(\epsilon) \sim \text{const.}$$

$$(\nu = 1)$$



$$\rho(\epsilon) \sim \epsilon^{1/2}$$

$$(\nu = 3/2)$$

$$V(x) \sim |x|^a$$

$$\rho(\epsilon) \sim \epsilon^{(2-a)/2a}$$

| | | |
|---------------|-----------|-----------|
| $0 < \nu < 1$ | $\nu = 1$ | $\nu > 1$ |
|---------------|-----------|-----------|

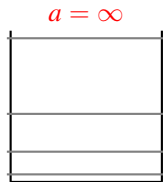
$$\frac{\tilde{\Omega}(E, N)}{\Omega(E)} \rightarrow \text{Fréchet}$$

Gumbel

Weibull

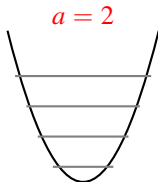
Integer partitions and the ideal Bose gas

Density of states $\rho(\epsilon) = \nu \epsilon^{\nu-1}$ for different $1d$ potentials



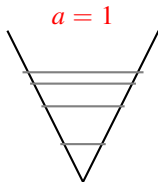
$$\rho(\epsilon) \sim \epsilon^{-1/2}$$

$$(\nu = 1/2)$$



$$\rho(\epsilon) \sim \text{const.}$$

$$(\nu = 1)$$



$$\rho(\epsilon) \sim \epsilon^{1/2}$$

$$(\nu = 3/2)$$

$$V(x) \sim |x|^a$$

$$\rho(\epsilon) \sim \epsilon^{(2-a)/2a}$$

| | | |
|---------------|-----------|-----------|
| $0 < \nu < 1$ | $\nu = 1$ | $\nu > 1$ |
|---------------|-----------|-----------|

$$\frac{\tilde{\Omega}(E, N)}{\Omega(E)} \rightarrow \text{Fréchet}$$

Gumbel

Weibull

e.g. $5 = 2^2 + 1^2$

$5 = 3 + 2$

$5 = 8^{2/3} + 1^{2/3}$

partition into powers > 1

partition into integers

partition into powers < 1

Integer partitions and the ideal Bose gas

G. H. Hardy, S. Ramanujan

Proc. London Math. Soc. **17**, 75 (1918)

Asymptotics of integer partitions

C. van Lier, G. E. Uhlenbeck

Physica, **4**, 531 (1937)

Ideal quantum gas, saddle point methods

P. Erdős, J. Lehner

Duke Math. J. **8**, 335 (1941)

Asymptotics of integer partitions, Gumbel

F. C. Auluck, D. S. Kothari

Proc. Cambridge Philos. Soc. **42**, 272 (1946)

Equally-spaced energy levels, Gumbel

A. Comtet, P. Leboeuf, S. N. Majumdar

Phys. Rev. Lett. **98**, 070404 (2007)

Power law-spaced energy levels, EVS