Handbook on excursion sets

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1 Priliminary calculations

1.1 Differentiation under the integral sign

$$\frac{\mathrm{d}}{\mathrm{d}x} \left(\int_{a(x)}^{b(x)} f(x,t)dt \right) = f(x,b(x))b'(x) - f(x,a(x))a'(x) + \int_{a(x)}^{b(x)} \frac{\partial}{\partial x} f(x,t)dt$$
(1)

1.2 Functions

1.2.1 Gamma function: Γ

$$\Gamma: z \to \int_0^\infty t^{z-1} e^{-t} dt \tag{2}$$

1.2.2 Error function: erf

$$\operatorname{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt \tag{3}$$

1.3 Lipschitz-Killing curvatures (LKC)

1.3.1 For a cube in 1,2 and 3 dimensions

For
$$\mathcal{M}_N = \prod_{i=1}^N [0 \ a], \ \mathcal{L}_j(\mathcal{M}_N) = \begin{pmatrix} N \\ j \end{pmatrix} a^j$$
 (4)

| N | LKC | Value | Meaning |
|---|------------------------------|--------|----------------------------|
| 1 | $\mathscr{L}_0(\mathcal{M})$ | 1 | Euler Characteristic |
| 1 | $\mathscr{L}_1(\mathcal{M})$ | a | Length of the segment |
| 2 | $\mathscr{L}_0(\mathcal{M})$ | 1 | Euler Characteristic |
| 2 | $\mathscr{L}_1(\mathcal{M})$ | 2a | Half the boundary length |
| 2 | $\mathscr{L}_2(\mathcal{M})$ | a^2 | Surface area |
| 3 | $\mathscr{L}_0(\mathcal{M})$ | 1 | Euler Characteristic |
| 3 | $\mathscr{L}_1(\mathcal{M})$ | 3a | Twice the caliper diameter |
| 3 | $\mathscr{L}_2(\mathcal{M})$ | $3a^2$ | Half the surface area |
| 3 | $\mathscr{L}_3(\mathcal{M})$ | a^3 | Volume |

1.4 Volume of the unit ball

1.5 Probabilistic Hermite polynomials

$$\mathcal{H}_{n} = (-1)^{n} e^{x^{2}/2} \frac{d^{n}}{dx^{n}} e^{-x^{2}/2}$$
 (7)
$$\begin{cases} \mathcal{H}_{0}(x) = 1 \\ \mathcal{H}_{1}(x) = x \\ \mathcal{H}_{2}(x) = x^{2} - 1 \\ \mathcal{H}_{3}(x) = x^{3} - 3 \\ \dots \end{cases}$$
 (8)

1.6 Flag coefficients

$$\begin{bmatrix} n & j & \text{nag} \\ \hline 0 & 0 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 1 \\ 2 & 0 & 1 \\ 2 & 1 & \pi/2 \\ 2 & 2 & 1 \\ 3 & 0 & 1 \\ 3 & 1 & 2 \\ 3 & 2 & 2 \\ 3 & 3 & 1 \end{bmatrix}$$
 (10)

1.7 Second spectral moment

$$\lambda_2 = \mathbb{V}\left\{\frac{\partial g(\boldsymbol{x})}{\partial x_i}\right\} = \frac{\mathrm{d}^2 \mathcal{C}(h)}{\mathrm{d}h^2}$$
 (11)

$$\begin{array}{c|cccc} Covariance & Parameter & \lambda_2 \\ \hline Gaussian & - & 2\sigma^2/L_c^2 \end{array} \tag{12}$$

2 Gaussian Minkowsky functionals

2.1 Preliminary

2.1.1 Gaussian measure: $\gamma^k(\mathcal{H})$

$$\gamma^{k}(\mathcal{H}) = P\{\boldsymbol{X} \in \mathcal{H}\} = \frac{1}{\sigma^{k}(2\pi)^{k/2}} \int_{\mathcal{H}} e^{-\|\boldsymbol{x} - \boldsymbol{\mu}\|^{2}/2\sigma^{2}} d\boldsymbol{x}$$
(13)

2.1.2 Tube Taylor expansion

$$\gamma^{k}(\text{tube}(\mathcal{H}, \rho)) = \sum_{j=0}^{\infty} \frac{\rho^{j}}{j!} \mathscr{M}_{j}^{\gamma^{k}}(\mathcal{H})$$
 (14)

2.2 GMFs for Gaussian distribution

2.2.1 Application to one-dimensional tail hitting set: k = 1 and $\mathcal{H} = [\kappa \infty)$

Tail probability \bar{F} :

$$\gamma^{1}([\kappa \infty)) = P\{X \ge \kappa\} = \frac{1}{\sigma\sqrt{2\pi}} \int_{\kappa}^{\infty} e^{-(x-\mu)^{2}/2\sigma^{2}} dx = \bar{F}(\kappa)$$
 (15)

Error function erf:

$$\gamma^{1}([\kappa \infty)) = \bar{F}(\kappa) = \frac{1}{\sigma\sqrt{2\pi}} \int_{\frac{\kappa-\mu}{\sigma\sqrt{2}}}^{\infty} e^{-t^{2}} \sqrt{2}\sigma dt = \frac{1}{\sqrt{\pi}} \int_{\frac{\kappa-\mu}{\sigma\sqrt{2}}}^{\infty} e^{-t^{2}} dt$$

$$= \frac{1}{\sqrt{\pi}} \left[\int_{0}^{\infty} e^{-t^{2}} dt - \int_{0}^{\frac{\kappa-\mu}{\sigma\sqrt{2}}} e^{-t^{2}} dt \right]$$

$$= \frac{1}{\sqrt{\pi}} \left[\frac{\sqrt{\pi}}{2} - \frac{\sqrt{\pi}}{2} \operatorname{erf}\left(\frac{\kappa-\mu}{\sigma\sqrt{2}}\right) \right]$$

$$= \frac{1}{2} \left(1 - \operatorname{erf}\left(\frac{\kappa-\mu}{\sigma\sqrt{2}}\right) \right)$$

Tube expansion:

$$\gamma^{1}(\text{tube}([\kappa \infty), \rho)) = \gamma^{1}([\kappa - \rho \infty)) = \bar{F}(\kappa - \rho) = \sum_{j=0}^{\infty} \frac{(-\rho)^{j}}{j!} \frac{\mathrm{d}^{j} \bar{F}(\kappa)}{\mathrm{d}\kappa^{j}} \quad (17)$$

Identification:

• Notations:

Tail hitting set:
$$\bar{\mathcal{H}}^{\text{\tiny ID}} = [\kappa \infty)$$
 (18a)

Variable substitution:
$$\tilde{\kappa} \leftarrow \frac{\kappa - \mu}{\sigma}$$
 (18b)

• For j = 0:

$$\mathscr{M}_0^{\gamma^1}(\bar{\mathcal{H}}^{1D}) = \bar{F}(\kappa) \tag{19}$$

• For j > 0:

$$\mathscr{M}_{j}^{\gamma^{1}}(\bar{\mathcal{H}}^{1D}) = (-1)^{j} \frac{\mathrm{d}^{j} \bar{F}(\kappa)}{\mathrm{d}\kappa^{j}} = \frac{e^{-(\kappa-\mu)^{2}/2\sigma^{2}}}{\sigma^{j}\sqrt{2\pi}} \mathscr{H}_{j-1}((\kappa-\mu)/\sigma)$$
(20)

• First values:

$$\mathcal{M}_{0}^{\gamma^{1}}(\bar{\mathcal{H}}^{\text{1D}}) = \frac{1}{2} \left(1 - \operatorname{erf} \left(\frac{\tilde{\kappa}}{\sqrt{2}} \right) \right)
\mathcal{M}_{1}^{\gamma^{1}}(\bar{\mathcal{H}}^{\text{1D}}) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\tilde{\kappa}^{2}/2}
\mathcal{M}_{2}^{\gamma^{1}}(\bar{\mathcal{H}}^{\text{1D}}) = \frac{1}{\sigma^{2}\sqrt{2\pi}} \tilde{\kappa} e^{-\tilde{\kappa}^{2}/2}
\mathcal{M}_{3}^{\gamma^{1}}(\bar{\mathcal{H}}^{\text{1D}}) = \frac{1}{\sigma^{3}\sqrt{2\pi}} (\tilde{\kappa}^{2} - 1) e^{-\tilde{\kappa}^{2}/2}
\mathcal{M}_{4}^{\gamma^{1}}(\bar{\mathcal{H}}^{\text{1D}}) = \frac{1}{\sigma^{4}\sqrt{2\pi}} (\tilde{\kappa}^{3} - 3\tilde{\kappa}) e^{-\tilde{\kappa}^{2}/2}
\dots$$
(21)

• Derivatives:

$$\forall j, \quad \frac{\mathrm{d}\mathscr{M}_{j}^{\gamma^{1}}}{\mathrm{d}\kappa}(\bar{\mathcal{H}}^{\mathrm{1D}}) = (-1)^{j} \frac{\mathrm{d}^{j+1}\bar{F}(\kappa)}{\mathrm{d}\kappa^{j+1}} = -\mathscr{M}_{j+1}^{\gamma^{1}}(\bar{\mathcal{H}}^{\mathrm{1D}}) \tag{22}$$

2.2.2 Application to one-dimensional cumulative hitting set: k = 1 and $\mathcal{H} = (-\infty \ \kappa]$

Cumulative probability F:

$$\gamma^{1}((-\infty \kappa]) = P\{X \le \kappa\} = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\kappa} e^{-(x-\mu)^{2}/2\sigma^{2}} dx = F(\kappa)$$
 (23)

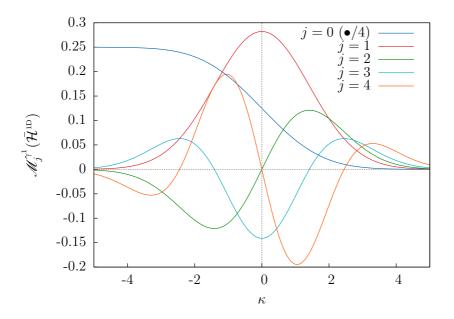


Figure 1: First 4 GMF for $\sigma^2=2$ and $\mathcal{H}=\bar{\mathcal{H}}^{\text{\tiny 1D}}=[\kappa \ \infty)$

Error function erf:

$$\gamma^{1}((-\infty \kappa]) = F(\kappa) = \frac{1}{2} \left(1 + \operatorname{erf}\left(\frac{\kappa - \mu}{\sigma\sqrt{2}}\right) \right)$$
 (24)

Tube expension:

$$\gamma^{1}(\text{tube}((-\infty \kappa], \rho)) = \gamma^{1}((-\infty \kappa + \rho]) = F(\kappa + \rho) = \sum_{j=0}^{\infty} \frac{\rho^{j}}{j!} \frac{d^{j} F(\kappa)}{d\kappa^{j}} \quad (25)$$

Identification:

• Notations:

Tail hitting set:
$$\mathcal{H}^{\text{1D}} = (-\infty \ \kappa]$$
 (26a)

Variable substitution:
$$\tilde{\kappa} \leftarrow \frac{\kappa - \mu}{\sigma}$$
 (26b)

• For j = 0:

$$\mathscr{M}_0^{\gamma^1}(\mathcal{H}^{\text{\tiny 1D}}) = F(\kappa) \tag{27}$$

• For j > 0:

$$\mathcal{M}_{j}^{\gamma^{1}}(\mathcal{H}^{1D}) = \frac{\mathrm{d}^{j} F(\kappa)}{\mathrm{d}\kappa^{j}} = (-1)^{j+1} \frac{e^{-(\kappa-\mu)^{2}/2\sigma^{2}}}{\sigma^{j} \sqrt{2\pi}} \mathcal{H}_{j-1}((\kappa-\mu)/\sigma) \quad (28)$$

• First values:

$$\mathcal{M}_{0}^{\gamma^{1}}(\mathcal{H}^{1D}) = \frac{1}{2} \left(1 - \operatorname{erf} \left(\frac{\tilde{\kappa}}{\sqrt{2}} \right) \right)
\mathcal{M}_{1}^{\gamma^{1}}(\mathcal{H}^{1D}) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\tilde{\kappa}^{2}/2}
\mathcal{M}_{2}^{\gamma^{1}}(\mathcal{H}^{1D}) = \frac{-1}{\sigma^{2}\sqrt{2\pi}} \tilde{\kappa} e^{-\tilde{\kappa}^{2}/2}
\mathcal{M}_{3}^{\gamma^{1}}(\mathcal{H}^{1D}) = \frac{1}{\sigma^{3}\sqrt{2\pi}} (\tilde{\kappa}^{2} - 1) e^{-\tilde{\kappa}^{2}/2}
\mathcal{M}_{4}^{\gamma^{1}}(\mathcal{H}^{1D}) = \frac{-1}{\sigma^{4}\sqrt{2\pi}} (\tilde{\kappa}^{3} - 3\tilde{\kappa}) e^{-\tilde{\kappa}^{2}/2}$$

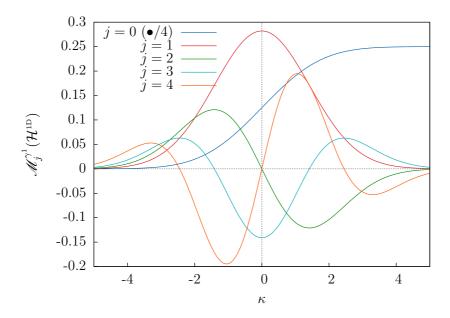


Figure 2: First 4 GMF for $\sigma^2=2$ and $\mathcal{H}=\mathcal{H}^{\text{\tiny 1D}}=(-\infty~\kappa]$

• Derivatives:

$$\forall j, \quad \frac{\mathrm{d}\mathscr{M}_{j}^{\gamma^{1}}}{\mathrm{d}\kappa}(\mathcal{H}^{\text{\tiny 1D}}) = \frac{\mathrm{d}^{j+1}F(\kappa)}{\mathrm{d}\kappa^{j+1}} = \mathscr{M}_{j+1}^{\gamma^{1}}(\mathcal{H}^{\text{\tiny 1D}})$$
(30)

2.3 GMFs for Gaussian related Random Fields

$$g_r: \Omega \times \mathbf{R}^N \xrightarrow{\mathbf{g}} \mathbf{R}^k \xrightarrow{S} \mathbf{R}$$
 (31)

$$P\{g_r \in \mathcal{H}\} = P\{S(g) \in \mathcal{H}\} = P\{g \in S^{-1}(\mathcal{H})\} = \gamma^k(S^{-1}(\mathcal{H}))$$
 (32)

2.3.1 Log-normal distribution

$$k = 1 , S = \exp \text{ and } S^{-1} = \ln$$
 (33)

$$\begin{array}{rcl}
\mu & = & \ln(\mu_{\rm ln}) - \frac{1}{2} \ln(1 + \sigma_{\rm ln}^2 / \mu_{\rm ln}^2) \\
\sigma^2 & = & \ln(1 + \sigma_{\rm ln}^2 / \mu_{\rm ln}^2)
\end{array} \tag{34}$$

• Notations:

Log tail hitting set:
$$\bar{\mathcal{H}}_{ln}^{1D} = [ln(\kappa) \infty)$$
 (35a)

Log cumulative hitting set:
$$\mathcal{H}_{\ln}^{\text{\tiny 1D}} = (-\infty \ln(\kappa)]$$
 (35b)

Variable substitution:
$$\tilde{\kappa}_{ln} \leftarrow \frac{\ln(\kappa) - \mu}{\sigma}$$
 (35c)

• For
$$\mathcal{H} = \bar{\mathcal{H}}^{\text{\tiny ID}} = [\kappa \infty) \rightarrow S^{-1}(\mathcal{H}) = \bar{\mathcal{H}}^{\text{\tiny ID}}_{\text{ln}} = [\ln(\kappa) \infty)$$

$$\mathcal{M}_{0}^{\gamma^{1}}(\bar{\mathcal{H}}_{\ln}^{\text{1D}}) = \frac{1}{2} \left(1 - \operatorname{erf} \left(\frac{\tilde{\kappa}_{\ln}}{\sqrt{2}} \right) \right)$$

$$\mathcal{M}_{1}^{\gamma^{1}}(\bar{\mathcal{H}}_{\ln}^{\text{1D}}) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\tilde{\kappa}_{\ln}^{2}/2}$$

$$\mathcal{M}_{2}^{\gamma^{1}}(\bar{\mathcal{H}}_{\ln}^{\text{1D}}) = \frac{1}{\sigma^{2}\sqrt{2\pi}} \tilde{\kappa}_{\ln} e^{-\tilde{\kappa}_{\ln}^{2}/2}$$

$$\mathcal{M}_{3}^{\gamma^{1}}(\bar{\mathcal{H}}_{\ln}^{\text{1D}}) = \frac{1}{\sigma^{3}\sqrt{2\pi}} (\tilde{\kappa}_{\ln}^{2} - 1) e^{-\tilde{\kappa}_{\ln}^{2}/2}$$

$$\mathcal{M}_{4}^{\gamma^{1}}(\bar{\mathcal{H}}_{\ln}^{\text{1D}}) = \frac{1}{\sigma^{4}\sqrt{2\pi}} (\tilde{\kappa}_{\ln}^{3} - 3\tilde{\kappa}_{\ln}) e^{-\tilde{\kappa}_{\ln}^{2}/2}$$

$$\dots$$
(36)

Derivatives:

Help:
$$\left(e^{-(\ln(x) - \mu)^2 / 2\sigma^2} \right)' = -\frac{1}{\sigma^2} \frac{\ln(x) - \mu}{x} e^{-(\ln(x) - \mu)^2 / 2\sigma^2}$$

$$\frac{d\mathscr{M}_0^{-1}(\bar{\mathcal{H}}_{\ln}^{1D})}{d\kappa} = \frac{-1}{\sigma^2 \sqrt{2\pi}} \frac{1}{\kappa} e^{-(\ln(\kappa) - \mu)^2 / 2\sigma^2}$$

$$\frac{d\mathscr{M}_1^{-1}(\bar{\mathcal{H}}_{\ln}^{1D})}{d\kappa} = \frac{-1}{\sigma^2 \sqrt{2\pi}} \frac{1}{\kappa} \frac{\ln(\kappa) - \mu}{\sigma} e^{-(\ln(\kappa) - \mu)^2 / 2\sigma^2}$$

$$\frac{d\mathscr{M}_2^{-1}(\bar{\mathcal{H}}_{\ln}^{1D})}{d\kappa} = \frac{-1}{\sigma^3 \sqrt{2\pi}} \frac{1}{\kappa} \left[\frac{(\ln(\kappa) - \mu)^2}{\sigma^2} - 1 \right] e^{-(\ln(\kappa) - \mu)^2 / 2\sigma^2}$$

$$\frac{d\mathscr{M}_3^{-1}(\bar{\mathcal{H}}_{\ln}^{1D})}{d\kappa} = \frac{-1}{\sigma^4 \sqrt{2\pi}} \frac{1}{\kappa} \left[\frac{(\ln(\kappa) - \mu)^3}{\sigma^3} - 3 \frac{(\ln(\kappa) - \mu)}{\sigma} \right] e^{-(\ln(\kappa) - \mu)^2 / 2\sigma^2}$$

$$\cdots$$
Guess would be:
$$\forall j \geq 0,$$

$$\frac{d\mathscr{M}_j^{-1}(\bar{\mathcal{H}}_{\ln}^{1D})}{d\kappa} = \frac{-1}{\sigma^{j+1} \sqrt{2\pi}} \frac{1}{\kappa} \mathscr{M}_j(\tilde{\kappa}_{\ln}) e^{-\tilde{\kappa}_{\ln}^2/2}$$

$$\frac{\mathrm{d}\mathscr{M}_{j}^{\gamma^{1}}(\bar{\mathcal{H}}_{\ln}^{\mathrm{1D}})}{\mathrm{d}\kappa} = \frac{-1}{\sigma^{j+1}\sqrt{2\pi}} \frac{1}{\kappa} \mathscr{H}_{j}(\tilde{\kappa}_{\ln}) \ e^{-\tilde{\kappa}_{\ln}^{2}/2}$$

• For
$$\mathcal{H} = [0 \ \kappa] \rightarrow S^{-1}(\mathcal{H}) = \mathcal{H}_{\ln}^{1D} = (-\infty \ \ln(\kappa)]$$

$$\mathcal{M}_{0}^{\gamma^{1}}(\mathcal{H}_{\ln}^{1D}) = \frac{1}{2} \left(1 + \operatorname{erf} \left(\frac{\tilde{\kappa}_{\ln}}{\sqrt{2}} \right) \right)$$

$$\mathcal{M}_{1}^{\gamma^{1}}(\mathcal{H}_{\ln}^{1D}) = \frac{1}{\sigma^{1}\sqrt{2\pi}} e^{-\tilde{\kappa}_{\ln}^{2}/2}$$

$$\mathcal{M}_{2}^{\gamma^{1}}(\mathcal{H}_{\ln}^{1D}) = \frac{-1}{\sigma^{2}\sqrt{2\pi}} \tilde{\kappa}_{\ln} e^{-\tilde{\kappa}_{\ln}^{2}/2}$$

$$\mathcal{M}_{3}^{\gamma^{1}}(\mathcal{H}_{\ln}^{1D}) = \frac{1}{\sigma^{3}\sqrt{2\pi}} (\tilde{\kappa}_{\ln}^{2} - 1) e^{-\tilde{\kappa}_{\ln}^{2}/2}$$

$$\mathcal{M}_{4}^{\gamma^{1}}(\mathcal{H}_{\ln}^{1D}) = \frac{-1}{\sigma^{4}\sqrt{2\pi}} (\tilde{\kappa}_{\ln}^{3} - 3\tilde{\kappa}_{\ln}) e^{-\tilde{\kappa}_{\ln}^{2}/2}$$

$$\mathcal{M}_{4}^{\gamma^{1}}(\mathcal{H}_{\ln}^{1D}) = \frac{-1}{\sigma^{4}\sqrt{2\pi}} (\tilde{\kappa}_{\ln}^{3} - 3\tilde{\kappa}_{\ln}) e^{-\tilde{\kappa}_{\ln}^{2}/2}$$

Derivatives:

Guess would be:
$$\forall j \geq 0$$
,

$$\frac{d\mathscr{M}_{j}^{\gamma^{1}}(\mathcal{H}_{\ln}^{1D})}{d\kappa} = \frac{(-1)^{j}}{\sigma^{j+1}\sqrt{2\pi}} \frac{1}{\kappa} \mathscr{H}_{j}(\tilde{\kappa}_{\ln}) e^{-\tilde{\kappa}_{\ln}^{2}/2}$$
(39)

3 Expectation Formula

3.1 Gaussian distribution

3.1.1 General case

$$\mathbb{E}\left\{ \mathcal{L}_{j}(\mathcal{E}) \right\} = \sum_{i=0}^{N-j} \left[\begin{array}{c} i+j \\ i \end{array} \right] \left(\frac{\lambda_{2}}{2\pi} \right)^{i/2} \mathcal{L}_{i+j}(\mathcal{M}) \mathcal{M}_{i}^{\gamma^{k}}(S^{-1}(\mathcal{H}))$$
 (40)

Notations:

LKC of
$$\mathcal{M}$$
: $\mathcal{L}_j(\mathcal{M}) = \mathcal{L}_j^{\mathcal{M}}$ (41a)

LKC of
$$\mathcal{E}$$
: $\mathcal{L}_j(\mathcal{E}) = \mathcal{L}_j^{\mathcal{E}}$ (41b)

GMF of
$$S^{-1}(\mathcal{H})$$
: $\mathcal{M}_j^{\gamma^k}(S^{-1}(\mathcal{H})) = \mathcal{M}_j^{\gamma^k}$ (41c)

3.1.2 One dimension

$$\mathbb{E}\left\{\mathcal{L}_{0}^{\mathcal{E}}\right\} = \left(\frac{\lambda_{2}}{2\pi}\right)^{1/2} \mathcal{L}_{1}^{\mathcal{M}} \mathcal{M}_{1}^{\gamma^{k}} + \mathcal{L}_{0}^{\mathcal{M}} \mathcal{M}_{0}^{\gamma^{k}}$$
(42a)

$$\mathbb{E}\left\{\mathcal{L}_{1}^{\mathcal{E}}\right\} = \mathcal{L}_{1}^{\mathcal{M}} \mathcal{M}_{0}^{\gamma^{k}} \tag{42b}$$

3.1.3 Two dimensions

$$\mathbb{E}\left\{\mathcal{L}_{0}^{\mathcal{E}}\right\} = \frac{\lambda_{2}}{2\pi}\mathcal{L}_{2}^{\mathcal{M}}\mathcal{M}_{2}^{\gamma^{k}} + \sqrt{\frac{\lambda_{2}}{2\pi}}\mathcal{L}_{1}^{\mathcal{M}}\mathcal{M}_{1}^{\gamma^{k}} + \mathcal{L}_{0}^{\mathcal{M}}\mathcal{M}_{0}^{\gamma^{k}}$$
(43a)

$$\mathbb{E}\left\{\mathcal{L}_{1}^{\mathcal{E}}\right\} = \sqrt{\frac{\lambda_{2}\pi}{8}} \mathcal{L}_{2}^{\mathcal{M}} \mathcal{M}_{1}^{\gamma^{k}} + \mathcal{L}_{1}^{\mathcal{M}} \mathcal{M}_{0}^{\gamma^{k}}$$
(43b)

$$\mathbb{E}\left\{\mathcal{L}_{2}^{\mathcal{E}}\right\} = \mathcal{L}_{2}^{\mathcal{M}} \mathcal{M}_{0}^{\gamma^{k}} \tag{43c}$$

4 Size effect

4.1 One dimensional case

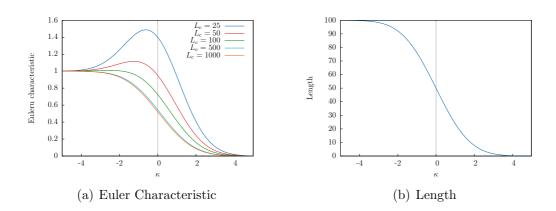


Figure 3: Expectation of LKC for $a=100,\,\sigma^2=2$ and $\mathcal{H}=[\kappa\,\infty)$ in 1D.

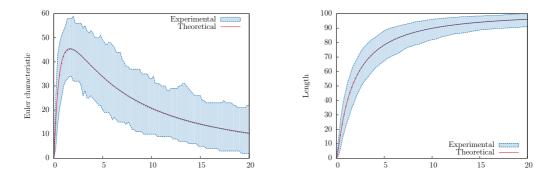


Figure 4: Expectation of LKC for $a=100, \mu=0.5, \sigma^2=2, L_c=0.5,$ and $\mathcal{H}=[0 \ \kappa]$ (log-normal distribution with cumulative hitting set) in 1D. Experimental results are calculated over 1000 realizations.

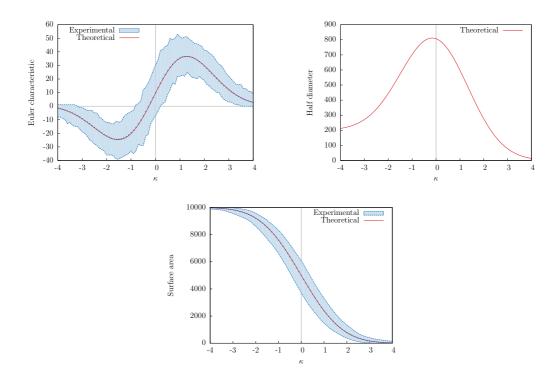


Figure 5: Expectation of LKC for $a=100, \mu=0.0, \sigma^2=2, L_c=5$, and $\mathcal{H}=[\kappa \infty)$ (gaussian distribution with tail hitting set) in 2D. Experimental results are calculated over 1000 realizations.

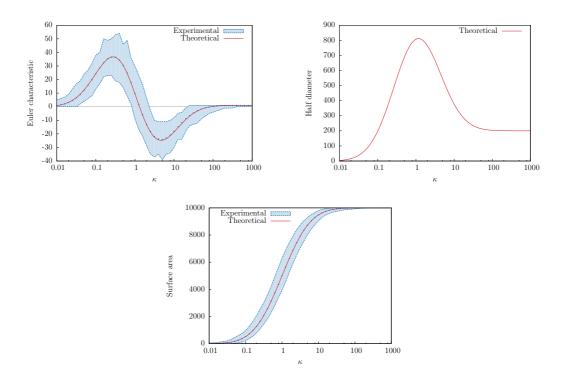


Figure 6: Expectation of LKC for $a=100, \mu=0.0, \sigma^2=2, L_c=5,$ and $\mathcal{H}=[0 \ \kappa]$ (lognormal distribution with cumulative hitting set) in 2D. Experimental results are calculated over 1000 realizations.

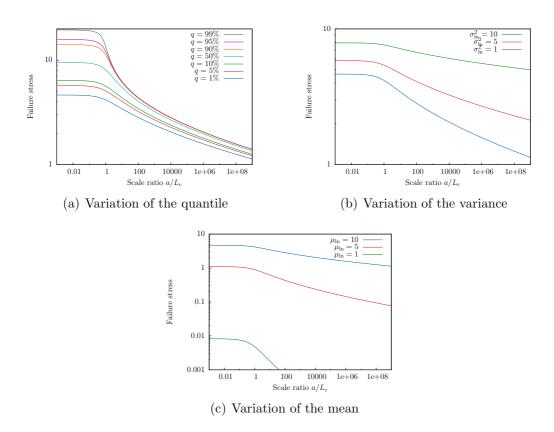


Figure 7: Size Effect: variations around $\mu_{\rm ln}=10,\,\sigma_{\rm ln}^2=10$ and q=1% in 1D.