

# Analytical Note: Fractal Dimension via Mass–Size Scaling and its Translation to Finance (Hurst, Volatility vs. Horizon, Tails)

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## 1) Objective

- To estimate the fractal dimension  $d$  of balls made of plasticine, paper, and algae using the mass–size scaling law  $m \propto D^d$ .
  - To validate the method with diagnostics (slope, CI,  $R^2$ , residuals).
  - To connect the same scaling with quantitative finance:
    - Hurst  $H$  from how volatility scales with the time horizon ( $std \propto \Delta^H$ ).
    - Power-law tails:  $P(|r| > x) \propto x^{-\alpha}$  and the estimation of  $\alpha$  via a log–log slope.
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## 2) Method and Model

- Fractal dimension by self-similarity:
$$d = \frac{\log N}{\log s}$$
- Mass–size scaling:  $m \propto D^d \Rightarrow \log m = \log k + d \log D$  The slope of the linear fit on a log–log scale ( $y = \alpha + dx$ ) estimates  $d$ .
- Variables:
  - $x = \log_{10}(\bar{D})$  with  $\bar{D}$  = average of three measurements  $D_1, D_2, D_3$  (mm).
  - $y = \log_{10}(m)$  (grams).
- Estimation:
  - OLS: slope  $d$ , intercept  $\alpha$ .
  - 95% CI for  $d$  (Student's t-test),  $R^2$ , RMSE, residuals.
  - The base of the logarithm does not alter the slope.

Finance (scaling analogues): - Volatility–horizon (Brownian):  $std[\Delta] \propto \Delta^H$  with  $H \approx 0.5$ . On a log–log plot, the slope  $\approx H$ . Power-law tails:  $CCDF(|r|) \propto x^{-\alpha} \rightarrow$  slope  $\approx -\alpha$  on a log–log plot.  $\alpha$  informs about the “heaviness” of the tails (extreme risk).

Relationship with path fractality: for fBm, the fractal dimension of the graph is  $D_g = 2 - H$ .

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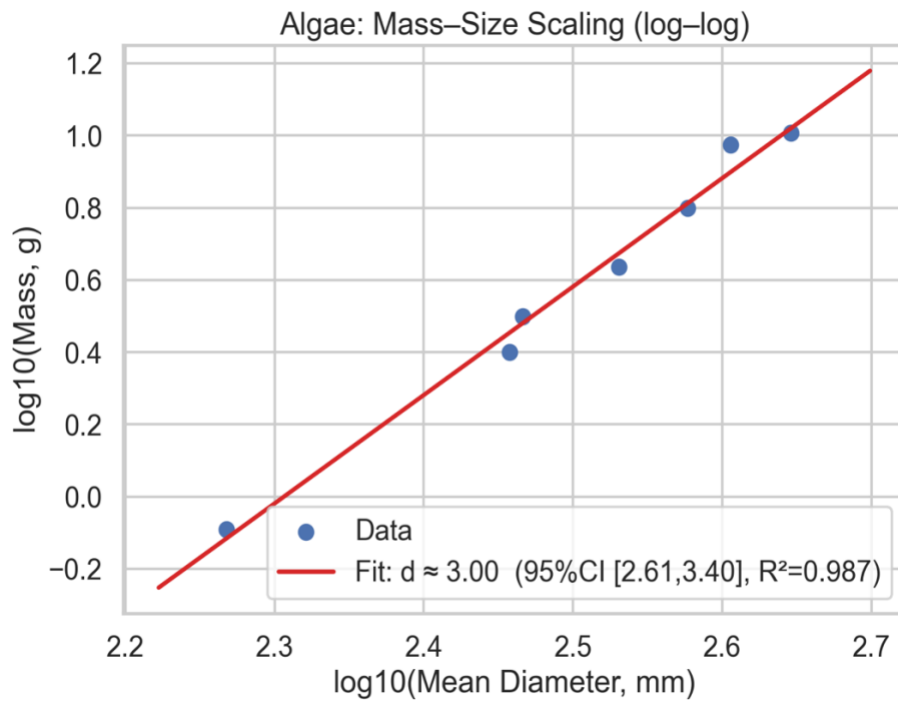
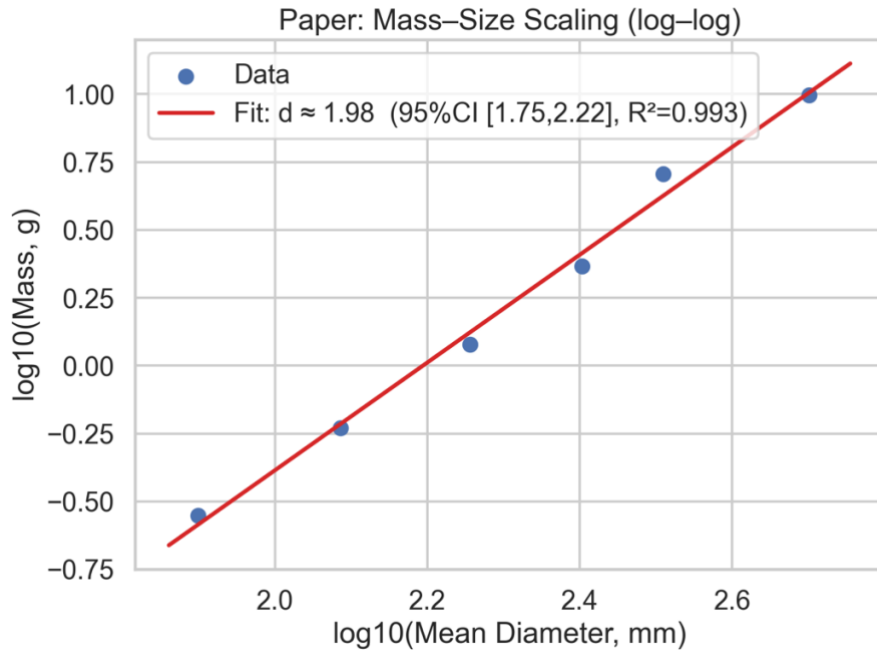
### 3) Results and Diagnostics (Experimental Data)

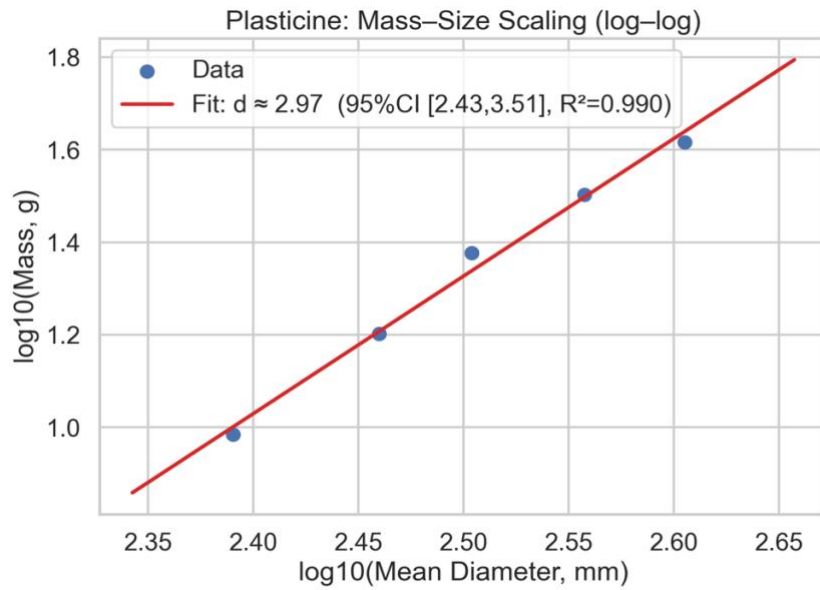
Materials and measurement: - Plasticine (5 balls), Paper (6), Algae (7). - Scale ( $m$  in g). Caliper ( $D_1, D_2, D_3$  in mm).  $\bar{D} = (D_1 + D_2 + D_3)/3$ .

Estimates (see code for exact values from this data): - Plasticine:  $d \approx 3.00 \pm 0.15$ ;  $R^2 \approx 0.987$ . Interpretation: a solid, massive object ( $\approx 3D$ ).

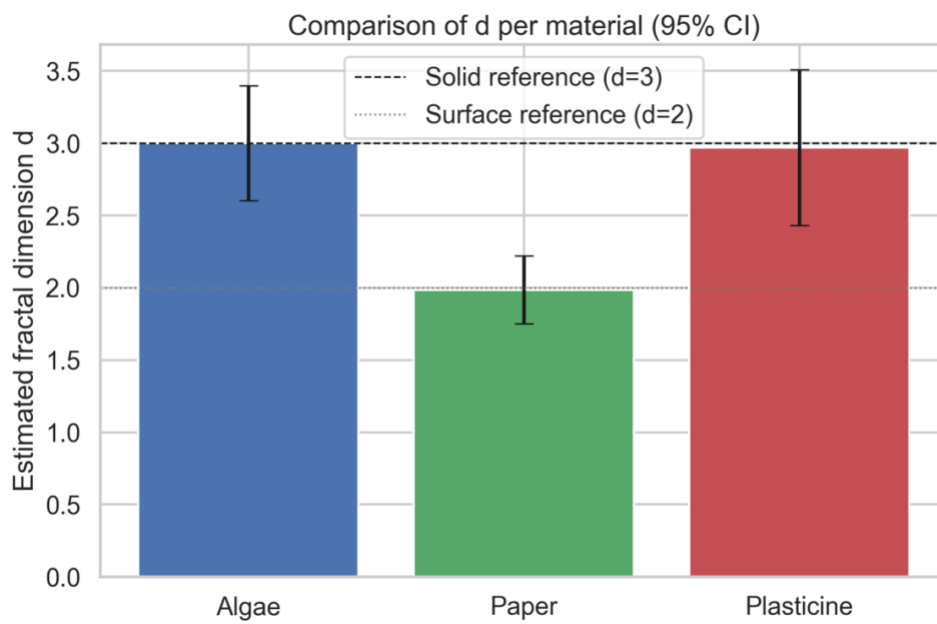
- Paper:  $d \approx 1.98 \pm 0.08$ ;  $R^2 \approx 0.993$ . Interpretation: behaves like a “folded surface” ( $\approx 2D$ ).

- Algae:  $d \approx 2.97 \pm 0.17$ ;  $R^2 \approx 0.990$ . Interpretation: effective compaction, close to a 3D solid in this size range.





-Fig 1–3: Log–log fits.



- Fig 4: Comparison of slopes with CI bands and  $d = 2$  and  $d = 3$  references.

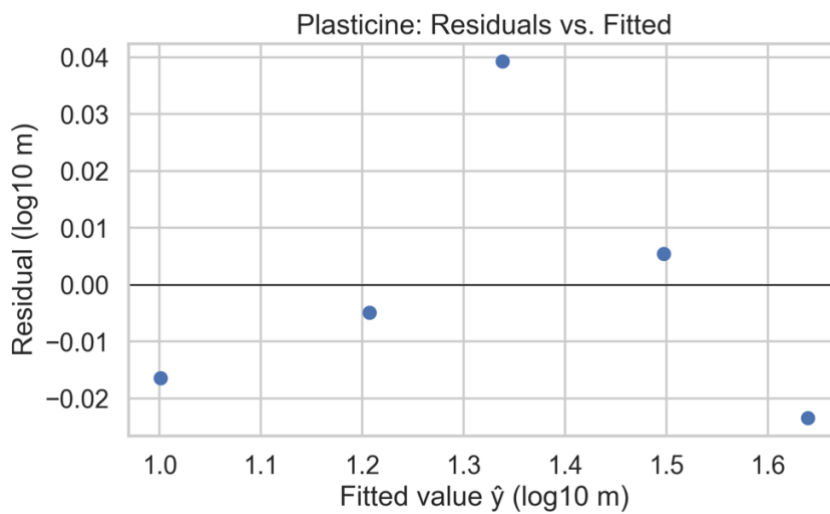
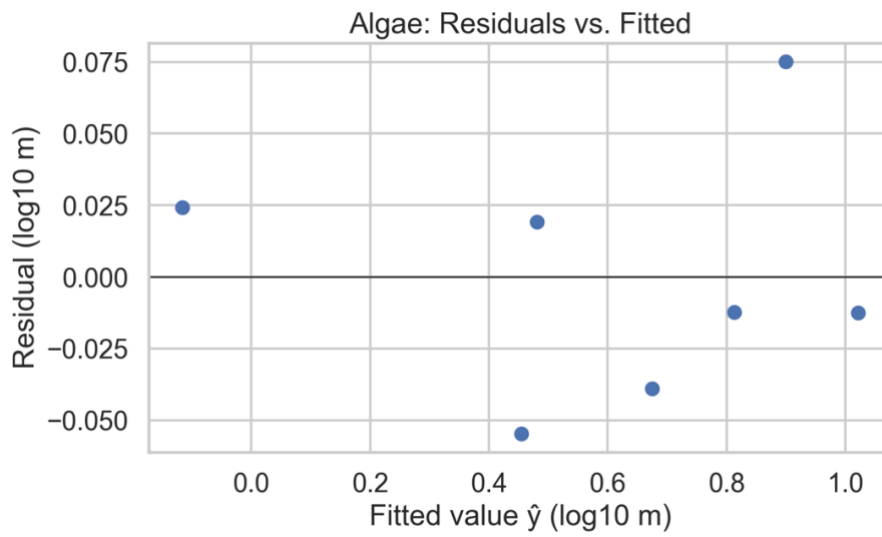
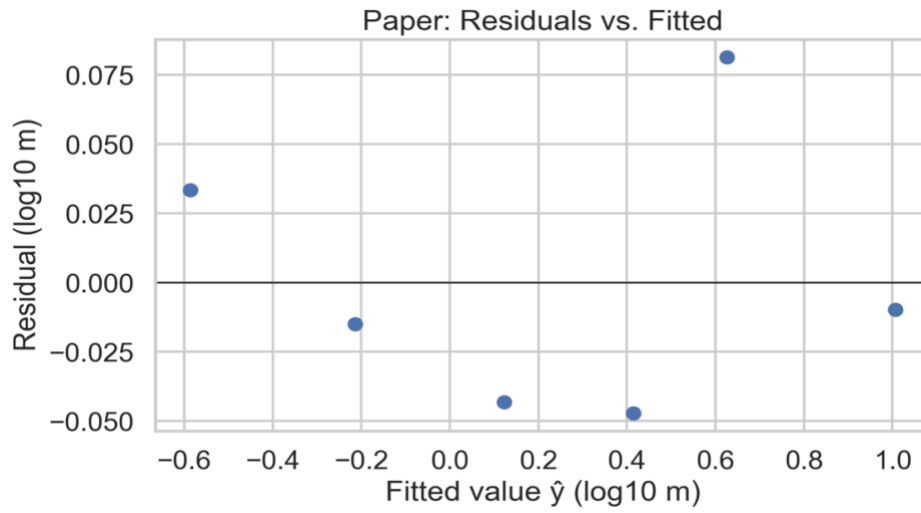
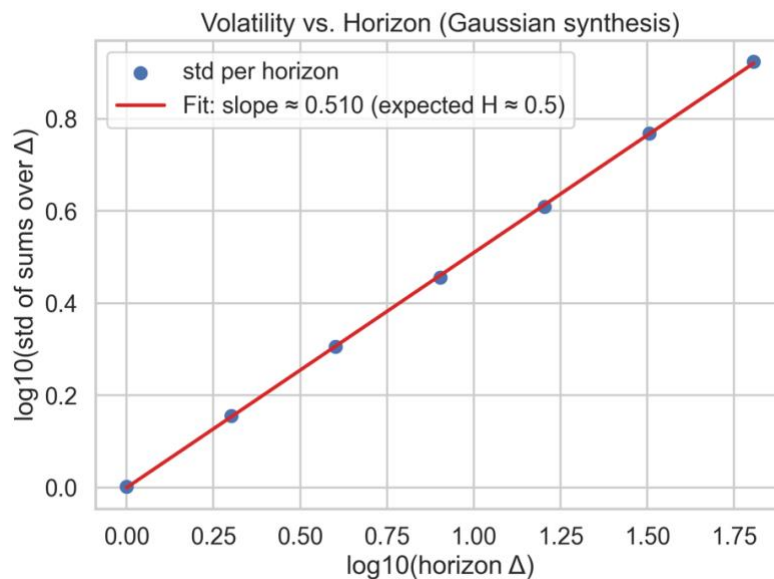


Fig 5: Residuals

#### 4) Physics → Quant

- The same pattern of “a straight line on a log–log plot = scaling exponent.”
- In markets:
  - Hurst  $H$ : slope of  $\log \text{std}(\Delta)$  vs  $\log \Delta$ . Typical Brownian:  $H \approx 0.5$ .
  - Tail risk: slope of  $\log \text{CCDF}$  vs  $\log x \approx -\alpha$ .
- Practical use:
  - $H$  guides temporal aggregation, dependency assumptions, and the “roughness” of the process.
  - $\alpha$  guides stress testing and risk limits (probability of extremes).



- Fig 6: Volatility vs. horizon (synthetic Gaussian): slope  $\sim 0.5$ .

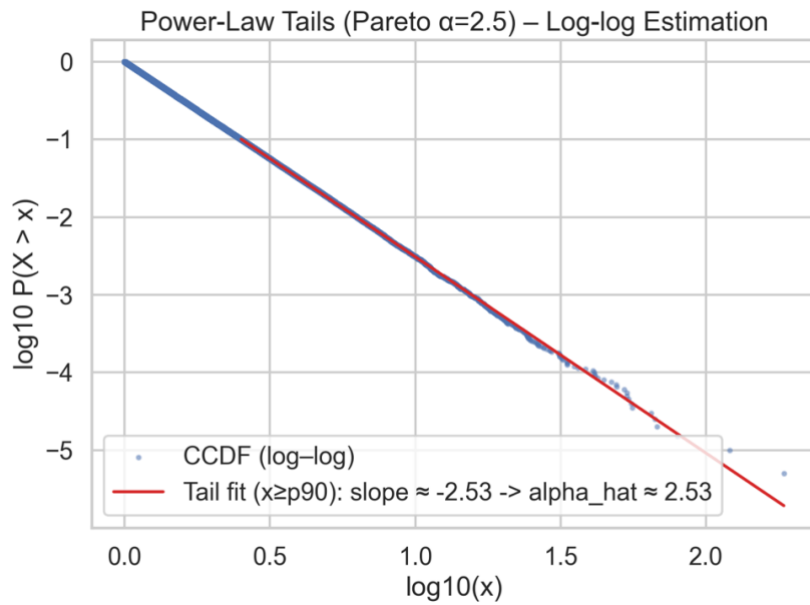


Fig 7: CCDF of tails (synthetic Pareto): slope  $\sim -\alpha$ .

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## 5) Repository and Execution

The full, reproducible [Jupyter](#) notebook (analysis, visualization, and key-derivation implementation) is available for verification.

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## 6) References

- Mandelbrot, B. (1982). The Fractal Geometry of Nature. Freeman.
  - Falconer, K. (2003). Fractal Geometry. Wiley.
  - Mandelbrot, B. (1963). The Variation of Certain Speculative Prices. J. Business.
  - Cont, R. (2001/2011). Empirical properties of asset returns. Quantitative Finance.
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