

Cheating at coin tossing

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Randomness vs. determinism

Is coin tossing

- ▶ **random?** (definit probability?)
- ▶ or **deterministic?** (control?)
- ▶ Is this an exclusion?

Outline

Arguments for

- ▶ A "fair" coin has probability $1/2$.
- ▶ There is no physical probability attached to the coin, we can cheat on each toss (by sufficient control).

My aim:

- ▶ The coin toss is **fine-grained deterministic**, but **coarsgrained random**.
- ▶ This explains why certain coin tossings allow to cheat (control is fine-grained), certain won't (control is coars-grained).
- ▶ Apparent randomness depends on two parameters: the uncertainty in control and the quasi-chaotic dynamics of the coin.

Deterministic vs. Random

A system $(\Gamma, (\phi_t)_{t \in \mathbb{R}})$ is **deterministic**, if $\phi_t : \Gamma \rightarrow \Gamma$ is a function.

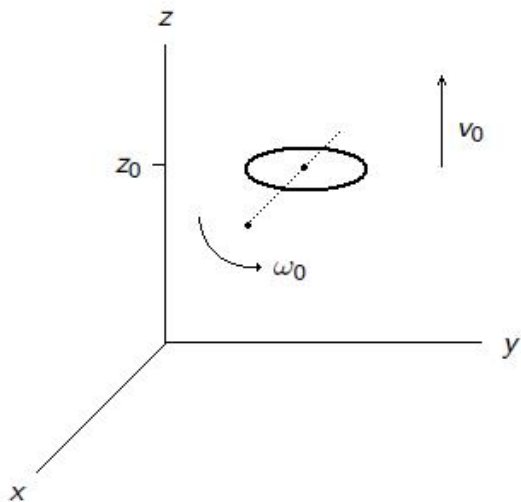
A system $\phi_{(t)} : \Gamma \rightarrow \Theta$, with Θ partitioned into two possible outcomes $\{A, \neg A\}$ is **fine grained deterministic**, if there exists a function

$$\chi_A : \Gamma \rightarrow \{0, 1\}$$

A system $\phi_{(t)} : \Gamma \rightarrow \Theta$ is **ϵ -coarse grained random**, if there exists $\epsilon > 0$ such that for every $p \in \Gamma$ and every open ball $B_\delta(p)$ ($\delta \geq \epsilon$), A has a non trivial probability in that ball, i.e.

$$P_{\delta,p}(A) := P(A|B_\delta(p)) \notin \{0, 1\}$$

Keller coin



The Keller coin is fine grained deterministic

A coin, tossed with initial velocity v_0 at height z_0 , will, at t , be at height

$$z(t) = z_0 + v_0 t - (g/2)t^2$$

Elapsed time until return to z_0 :

$$t^* = 2v_0/g$$

Flips per second

$$n_0 = \omega_0/\pi, \quad \omega_0 \text{ angular velocity}$$

Number of Flips

$$n = n_0 t^* = \frac{\omega_0}{\pi} \frac{2v_0}{g}$$

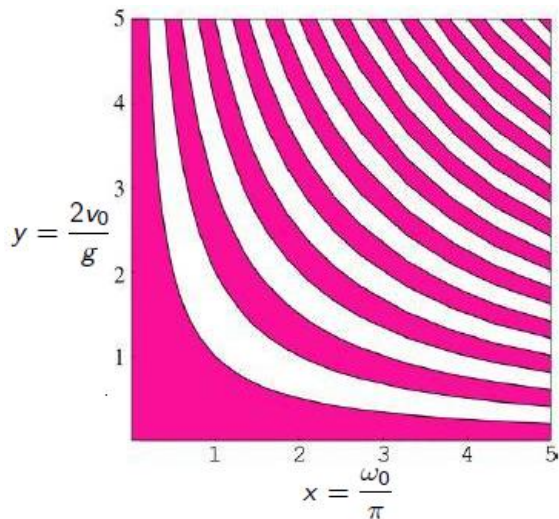
Coin lands

same side up if $0 \leq n \leq 1 \pmod{2}$

other side up if $1 \leq n \leq 2 \pmod{2}$

But also coars-grained random

Hyperbolas defined by $j = xy$, $j \in \mathbb{N}$



c.f. Keller 1986

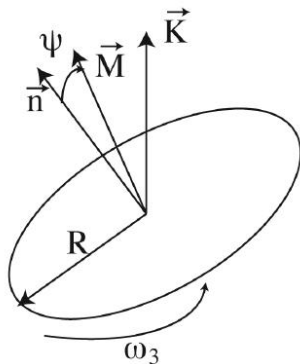
For physical probability

Assume a density on the (x, y) space. If it is approximately constant within a distance corresponding to two ribbons then

$$P(H) \approx P(T) \approx \frac{1}{2}$$

However, this becomes less true the smaller the control-ball.

Cheating at the coin



cf. Diaconis (2007).

Control $\psi \leq \pi/4$, to cheat.

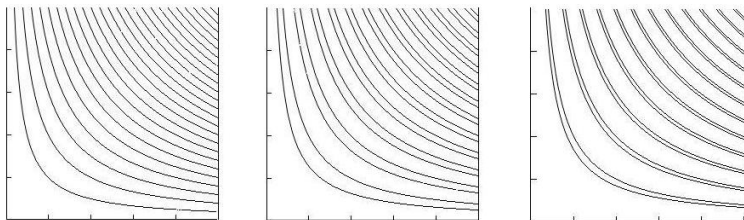
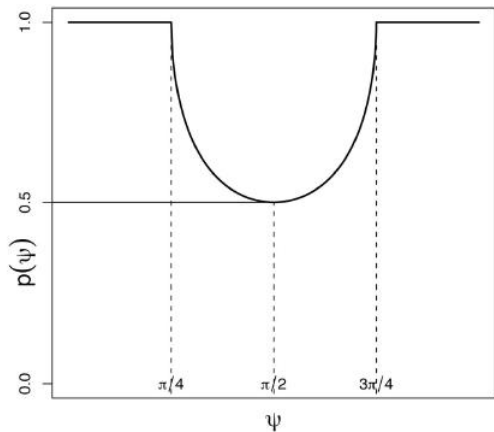


Figure: Coin with $\psi = \frac{\pi}{2}$, $\psi = \frac{5}{16}\pi$, $\psi = \frac{26}{100}\pi$.

Hyperbolas defined by

$$xy = \begin{cases} j & \text{if } j = 2n \\ j + \frac{2}{\pi} \sin^{-1} \cot^2 \psi & \text{else} \end{cases}$$

with $y = \frac{2v_0}{g}$, $x = \frac{\omega M}{\pi}$.



Diaconis (2007)

$$P_\psi = \begin{cases} \frac{1}{2} + \frac{\sin^{-1} \cot^2 \psi}{\pi} & \text{if } \frac{\pi}{4} \leq \psi \leq \frac{3\pi}{4} \\ \frac{1}{2} & \text{else} \end{cases}$$

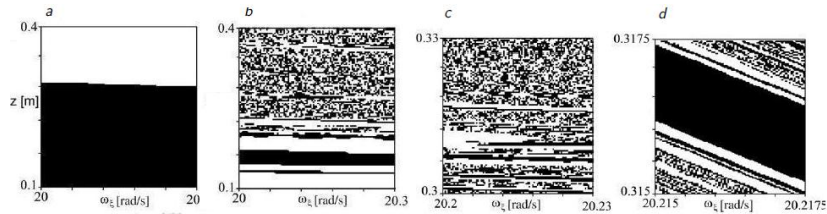
Against physical probability

There is no physical probability of the coin (cf. Jaynes 2003).
Everything depends on controlling the tossing.

- ▶ Are there systems where, tossing control might not be enough to cheat?
- ▶ This has to do with the shape of the basins of attractions which has to do with the sensitivity of the mechanism.

Bouncing (after free fall) with precession

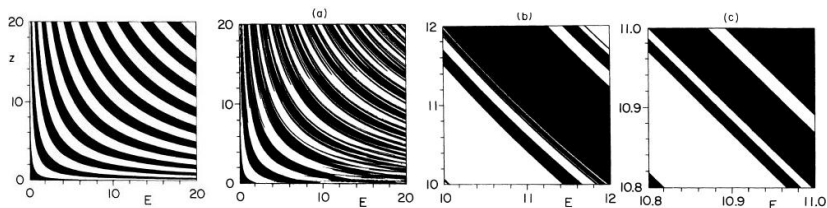
ω_ξ rotation around parallel to x axis.



Strzalko (2008)

Figure: (a) Keller coin (b) bouncing coin with successive enlargements (c, d) (no air resistance)

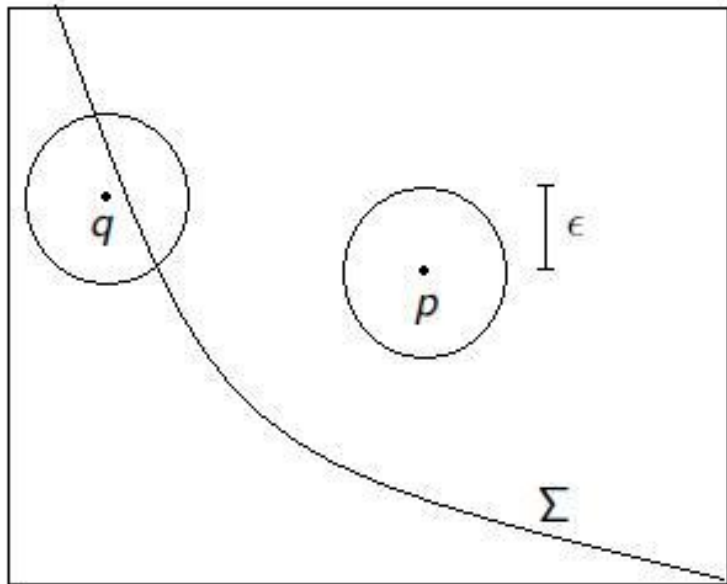
... without Precession

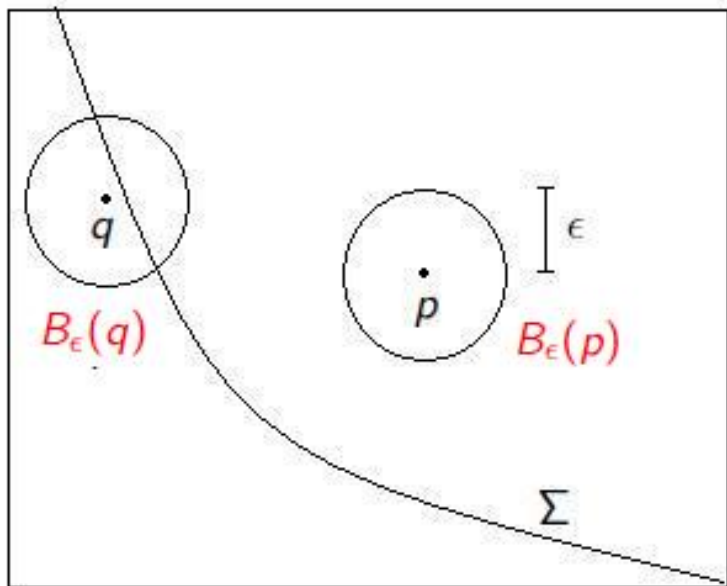


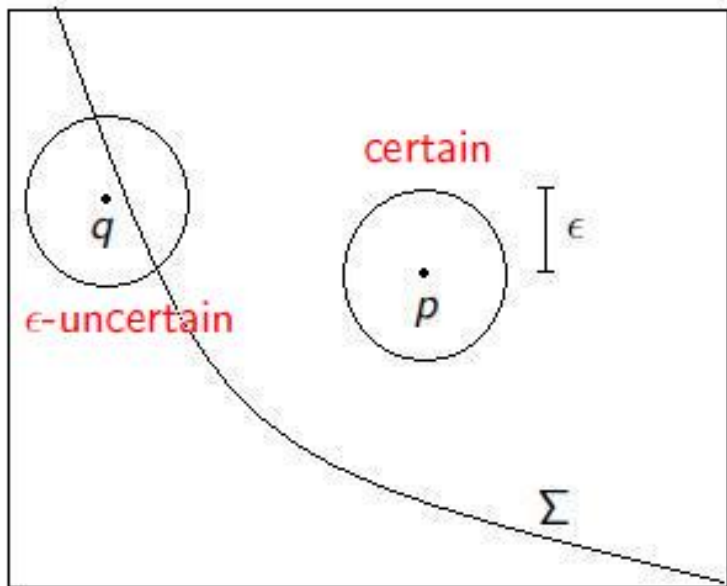
Vulovic, Prange (1986)

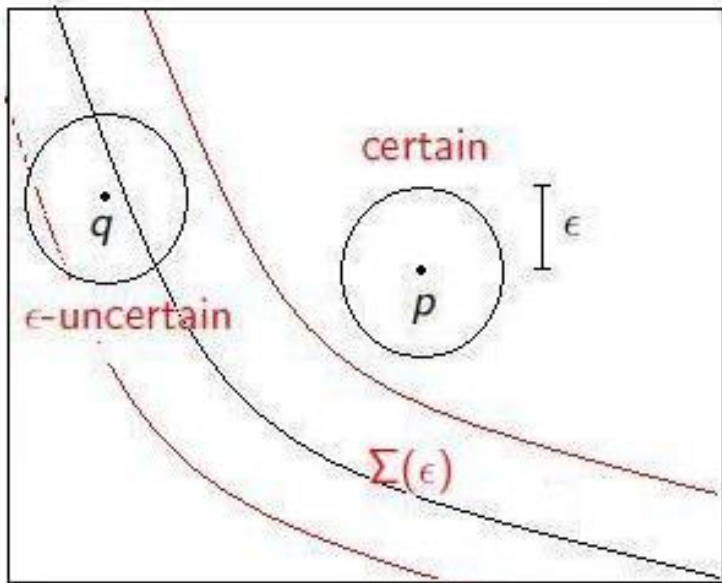
Figure: Keller coin and (a) bouncing coin with successive enlargements (b,c). $E = 0.51\omega^2$.

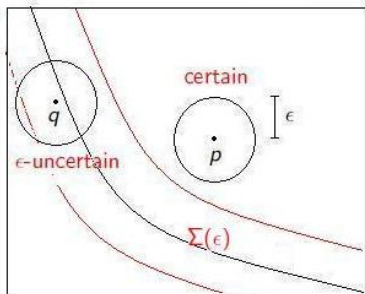
Probability of error











The probability of error

$$f(\epsilon) = \frac{\mu(\Sigma_\epsilon)}{\mu(S)}$$

can be interpreted as a **measure of how probable it is that cheating fails** (given the error ϵ in control).

If Σ is **non fractal** then

$$f(\epsilon) \sim \epsilon$$

If Σ is **fractal** then

$$f(\epsilon) \sim \epsilon^\alpha$$

$$\lim_{\epsilon} \frac{\ln f(\epsilon)}{\ln \epsilon} = \alpha$$

$$\alpha = N - D_0, \quad \alpha < 1$$

Eg. $\alpha = 0.1$ To reduce $f(\epsilon)$ by a factor 10 we need to reduce ϵ by a factor 10^{10} .
Improvement in prediction by improving accuracy in IC becomes harder as $\alpha \rightarrow 0$.

Variation of f_ϵ

f_ϵ might vary in phase space (with location p)

$$f_\epsilon(p) = \frac{\mu(\Sigma_\epsilon(p))}{\mu(B_\epsilon(p))}$$

Although Σ is not fractal, $\Sigma_\epsilon(p)$ can "approach fractality" as $p \rightarrow \infty$.

Then








$$f_\epsilon(p) \sim \epsilon^\alpha$$

for $p \rightarrow \infty$.

Conclusion

If a system is fine-grained deterministic and coarse grained random, then

- ▶ The probability of error depends not only on our general ability to control, but also on the regions of (high/low) sensitivity of the system.
- ▶ One may argue for a certain definite probability of an outcome in a system, if across different regions the probability of error is high and the color pattern is sufficiently regular.

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