## 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  $\phi_t : \Gamma \to \Gamma$  is a function.

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

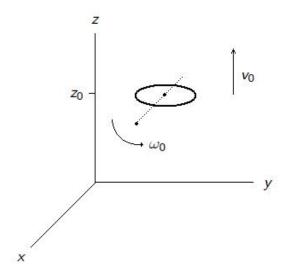
$$\chi_A:\Gamma \to \{0,1\}$$

A system  $\phi_{(t)}: \Gamma \to \Theta$  is  $\epsilon$ -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 hight  $z_0$ , will, at t, be at hight

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

**Elapsed time** until return to  $z_0$ :

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

Flips per second

$$n_0 = \omega_0/\pi$$
,  $\omega_0$  angular velocity

Number of Flips

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

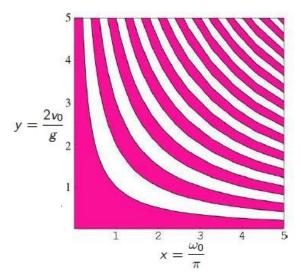
Coin lands

same side up if  $0 \le n \le 1 \mod 2$  other side up if  $1 \le n \le 2 \mod 2$ 



## But also coars-grained random

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



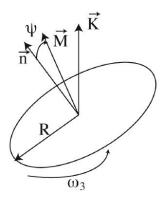
## For physical probability

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.

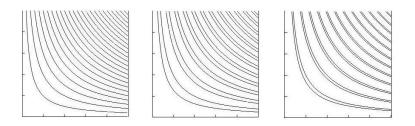
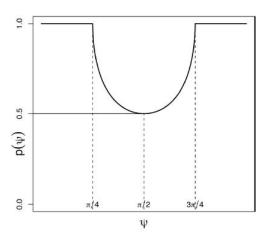


Figure: Coin with  $\psi = \frac{\pi}{2}$ ,  $\psi = \frac{5}{16}\pi$ ,  $\psi = \frac{26}{100}\pi$ .

#### Hyperbolas defined by

$$xy = \left\{ \begin{array}{ll} j & \text{if} & j = 2n \\ j + \frac{2}{\pi} \sin^- \cot^2 \psi & \text{else} \end{array} \right.$$

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



Diaconis (2007)

$$P_{\psi} = \left\{ egin{array}{ll} rac{1}{2} + rac{\sin^-\cot^2\psi}{\pi} & ext{if} & rac{\pi}{4} \leq \psi \leq rac{3\pi}{4} \ rac{1}{2} & ext{else} \end{array} 
ight.$$



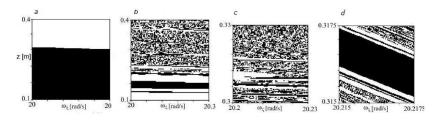
## Against physical probability

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

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

## Bouncing (after free fall) with precession

 $\omega_{\xi}$  rotation around parallel to x axis.

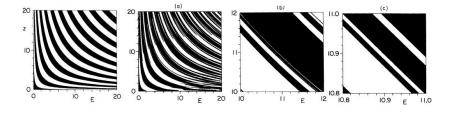


Strzalko (2008)

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



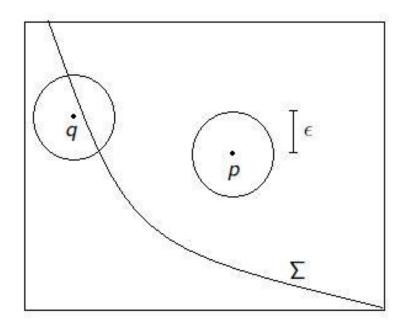
## ... without Precession

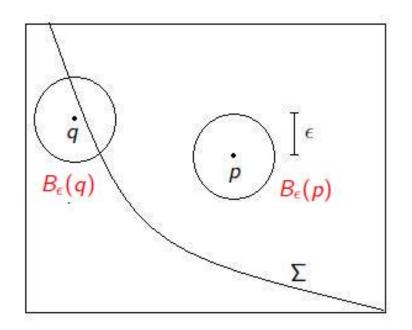


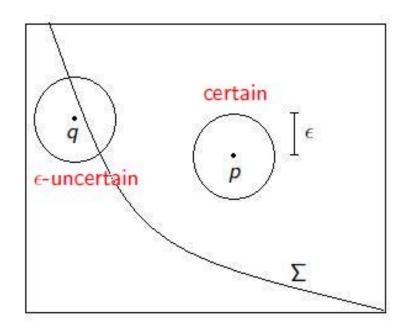
Vulovic, Prange (1986)

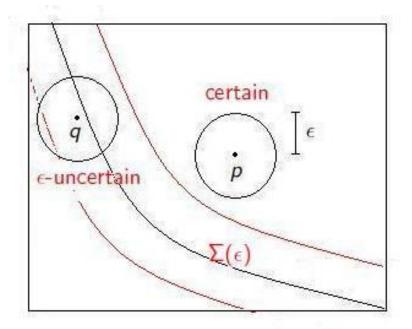
Figure: Keller coin and (a) bouncing coin with successive enlargements (b,c).  $E=0.51\omega^2$ .

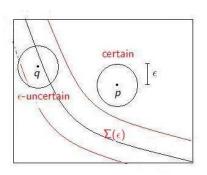
# Probability of error











#### 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  $\epsilon$  in control).

#### If $\Sigma$ is **non fractal** then

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

If  $\Sigma$  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  $\epsilon$  by a factor  $10^{10}$ . Improvent in prediction by improving accuracy in IC becomes harder as  $\alpha \to 0$ .

## Variation of $f_{\epsilon}$

 $f_{\epsilon}$  might vary in phase space (with location p)

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

Although  $\Sigma$  is not fractal,  $\Sigma_{\epsilon}(p)$  can "appraoch fractality" as  $p \to \infty$ .

Then

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

for  $p \to \infty$ .

#### 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 definit 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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