

# Probability Theory

GeoComput & ML

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# ReCap

$\text{tot err} = \text{comput.} + \text{dat.}$

# ReCap

comput error

- machine precision :  $\epsilon_{\text{mach}} = \beta^{1-p} / 2$
- representation
- operation

# ReCap

question : well posed

solution : well conditioned

condition number :  $\left| \frac{\Delta y / y}{\Delta x / x} \right|$

# ReCap

two cases

- linear system :  $\text{cond} = \| \mathbf{A} \| \| \mathbf{A}^{-1} \|$
- least square :  $\text{cond} = \| \mathbf{A} \| \| \mathbf{A}^+ \|$
- projection
- residual

# Guided Reading

# Why a Hydrology Paper

- very broad : geoscience
- interdisciplinary : nature of geoscience
- math link : eqn 1-3
- personal experience
- class promise

# Clouds

## Physics

- outliers → discovery
- Thomas Kuhn : science revolution



# Clouds

Hydrology

- scale
- uncertainty

# Clouds

# Direction

Theory-Guided data science

# Direction

## Traditional Science

- iteration between data and hypotheses
- knowledge discovery
- knowledge buildup

# Direction

## Data Science

- actionable models
- data under/misrepresentation
- interpretation

# Scale

SDM : relating field obs to its environment  
what's the scale for environment

# Scale

Motivating example

coordinate :  $x$

abundance :  $N(x)$

$\lambda(\cdot)$  as a function of its env  
 $z(x)$

$$s(z(x), \sigma) = \sum z(x_j) w(x_i, x_j, \sigma)$$

$$\log(\lambda(x)) = \sum_{i=1}^p \beta_i s_i(z_i(x_i), \sigma_i)^{i-1}$$

# Probability Theory



# Basic Concepts

- sample space ( $S$ ) : the collection of all the outcomes from a random experiment
- event ( $A$ )  $\subseteq S$
- Prob. function ( $P$ ) :  $A \rightarrow \#$

# Axioms

- $P(A) \in [0, 1]$
- $P(S) = 0$
- $P(\cup A) = \sum P(A)$

# Propositions

$$P(A^c) = 1 - P(A)$$

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

# Propositions

Notation

$$[A^c] = 1 - [A]$$

$$[A + B] = [A] + [B] - [A, B]$$

# Cond. Prob.

$$[A|B] = \frac{[A, B]}{[B]}$$

# Cond. Prob.

Example :

$$[\text{rain}, \text{Sat}] = [\text{rain}, \text{Sun}] = 0.5$$

$$[\text{rain}, \text{two conted days}] = 0.6$$

$$[\text{rain}, \text{weekend}] ?$$

# Cond. Prob.

$$[\text{Sat} + \text{Sun}] = [\text{Sat}] + [\text{Sun}] - [\text{Sat}, \text{Sun}]$$

$$[\text{Sat}, \text{Sun}] = [\text{Sun} | \text{Sat}] [\text{Sat}] = 0.3$$

$$[\text{Sat} + \text{Sun}] = 1 - [\text{Sat}, \text{Sun}] = 0.7$$

# Independence

$$[A, B] = [A][B]$$



# Independence

Example :

$$[\text{rain}, \text{Sat}] = [\text{rain}, \text{Sun}] = 0.5$$

Sat  $\perp\!\!\!\perp$  Sun

[rain, weekend] ?

# Independence

$$[\text{Sat}, \text{Sun}] = [\text{Sat}] + [\text{Sun}] = 0.25$$

$$[\text{Sat} + \text{Sun}] = 0.75$$

# Law of Total Probability

$$[B] = \sum [B|A][A]$$

# Law of Total Probability

## Kidney Stone Treatment

	A	B
S	81/87=0.93	234/270=0.87
L	192/263=0.73	55/80=0.69
	273/350=0.78	289/350=0.83

$$[E | A] = [E | A, S][S | A] + [E | A, L][L | A]$$

# Law of Total Probability

Bonus

$$\begin{aligned}[E|A] &= [[S + L, E]|A] \\ &= [[S, E] + [L, E]|A] \\ &= [[S, E]|A] + [[L, E]|A] \\ &= \frac{[S, E, A] + [L, E, A]}{[A]} \\ &= [E|S, A][S, A] + [E|L, A][L, A]\end{aligned}$$

# Bayes Theorem

$$[B_j|A] = \frac{[A|B_j][B_j]}{\sum [A|B_j][B_j]}$$

$$[B_j|A] = \frac{[A|B_j][B_j]}{[A]} = \frac{[A, B_j]}{[A]}$$

# Bayes Theorem

Example :

- 1% pop have cancer :  $[Y] = 0.01$
- 80% test + if cancer :  $[+ | Y] = 0.8$
- 9.6% test + if no cancer :  $[+ | N] = 0.096$

$$[Y | +] = ?$$

$$[Y | +] = \frac{[+ | Y][Y]}{[+ | Y][Y] + [+ | N][N]} = 0.48$$

# Random Variable

RV : real valued function mapped onto the sample space



# Random Variable

## Example

flip a coin twice, denote  $X$  as the # of heads

$$X(TT)=0, X(TH)=X(HT)=1, X(HH)=2$$

$$[X=0]=1/4, [X=1]=1/2, [X=2]=1/4$$

Prob. Distr. of  $X$

# Random Variable

pmf

$$[x_k] = [X=x_k], k=1,2,3...$$

# Expectation Value

$$E(X) = \sum x_k [x_k]$$

# Expectation Value

coin game

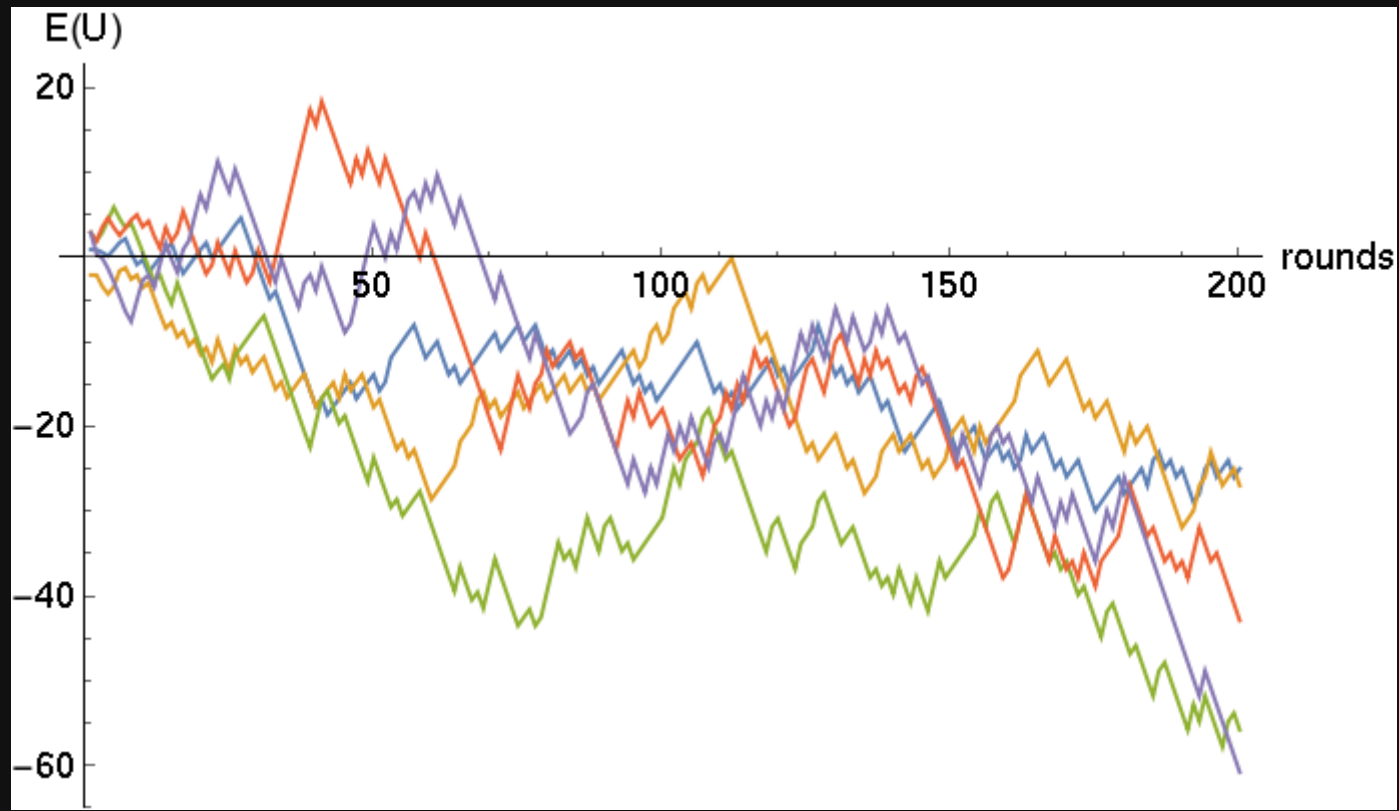
	H	T
H	3	-2
T	-2	1

$$\begin{cases} [H, U] = x \\ [H, I] = y \end{cases}$$

$$\Rightarrow E(U) = 3xy + (1 - x)(1 - y) - 2(x(1 - y) + y(1 - x))$$

# Expectation Value

Simulation :  $[U, H] \in \{0.1, 0.25, 0.5, 0.75, 0.9\}$



# Prob. Distr. Functions

Binomial

$$[k; n, p] = \binom{n}{k} p^k (1 - p)^{(n-k)}$$

# Prob. Distr. Functions

Poisson

$$[k; \lambda] = \frac{\lambda^k e^{-\lambda}}{k!}$$

# Prob. Distr. Functions

Normal

$$[x; \mu, \sigma] = (\sigma\sqrt{2\pi})^{-1} \exp(-(x - \mu)^2 / 2\sigma^2), \quad x \in \mathbb{R}$$



# Acknowledgement

Thanks for Your Attention

# References

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