



# Communicating probability with natural frequencies and the equivalent binomial count

Scott Ferson, University of Liverpool, United Kingdom

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# Communicating probability

- Probability density function graphs
- Box and whisker plots
- Cumulative probability function graphs
- Icon arrays
- Spinners, games
- Numerical statistics
- Percentages, odds, natural frequencies

# Natural frequencies and icon arrays

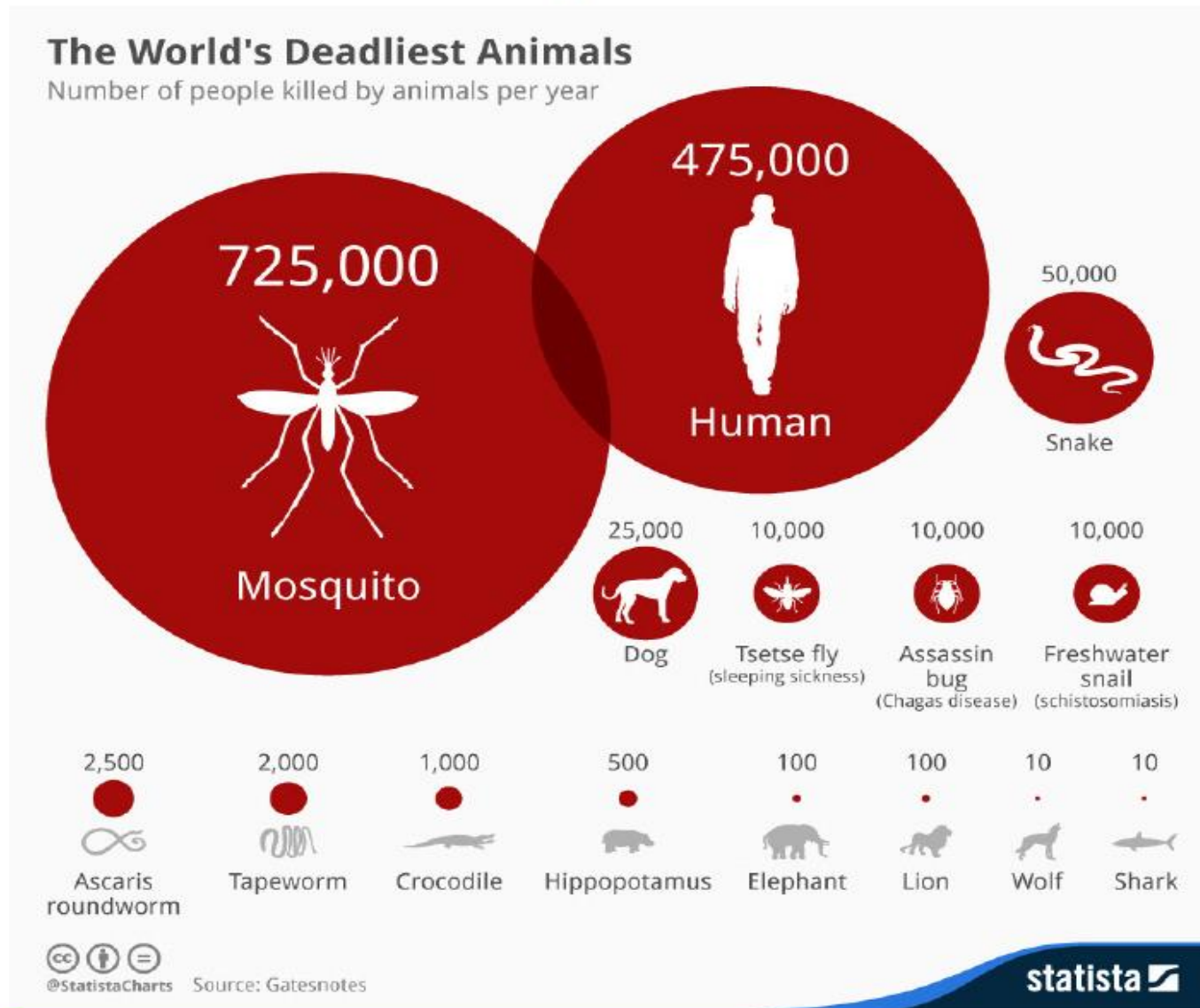
37 out of 100



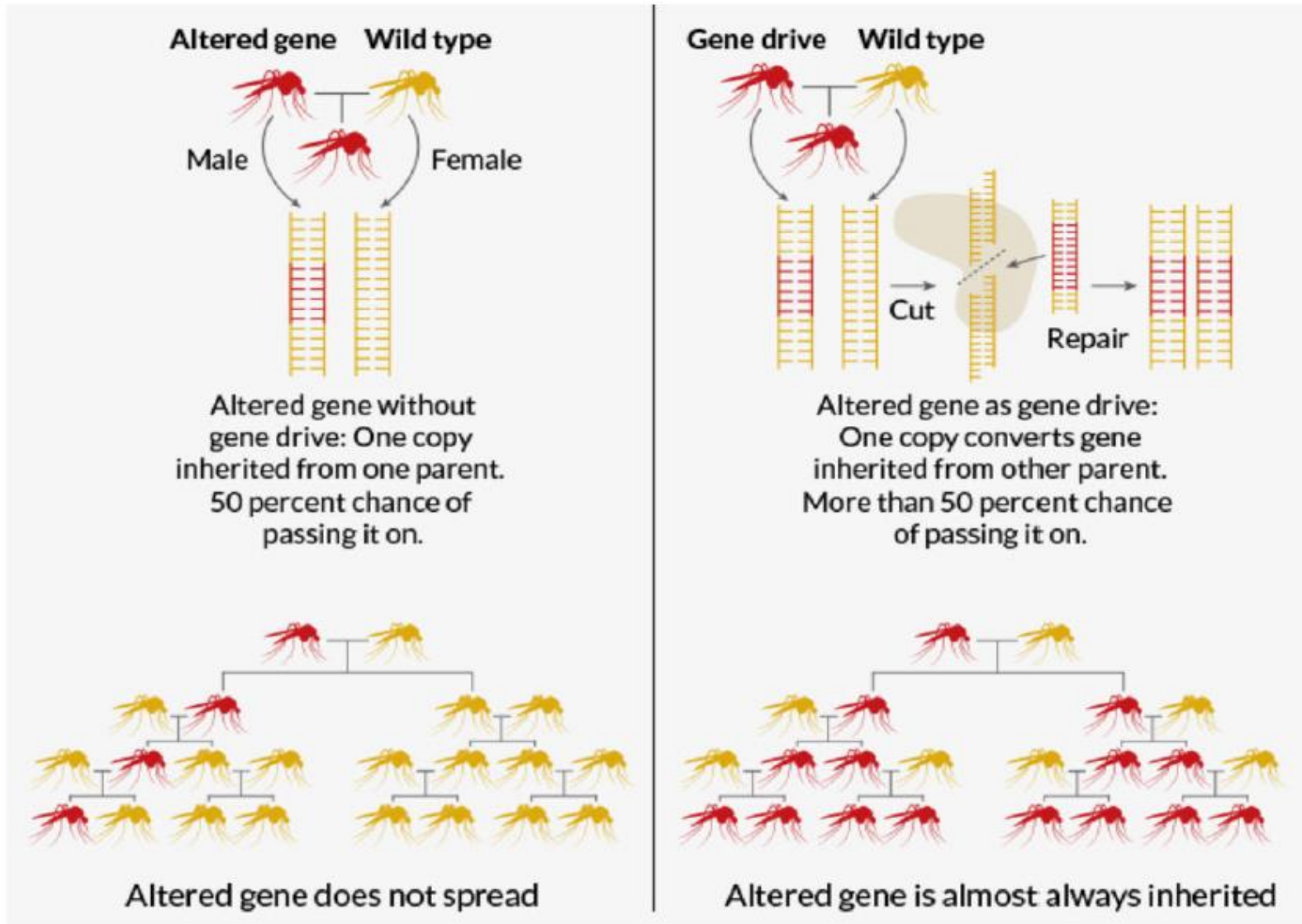
What if it's more complex than scalar  $p \in [0,1]$

- What if it is a distribution?
- What if it is a collection of distributions?
  - Second-order distribution
  - Robust Bayesian analysis
  - Probability box
  - Envelop of distributions from disagreeing experts

# What is the world's most dangerous animal?



# Medelian inheritance versus gene drive



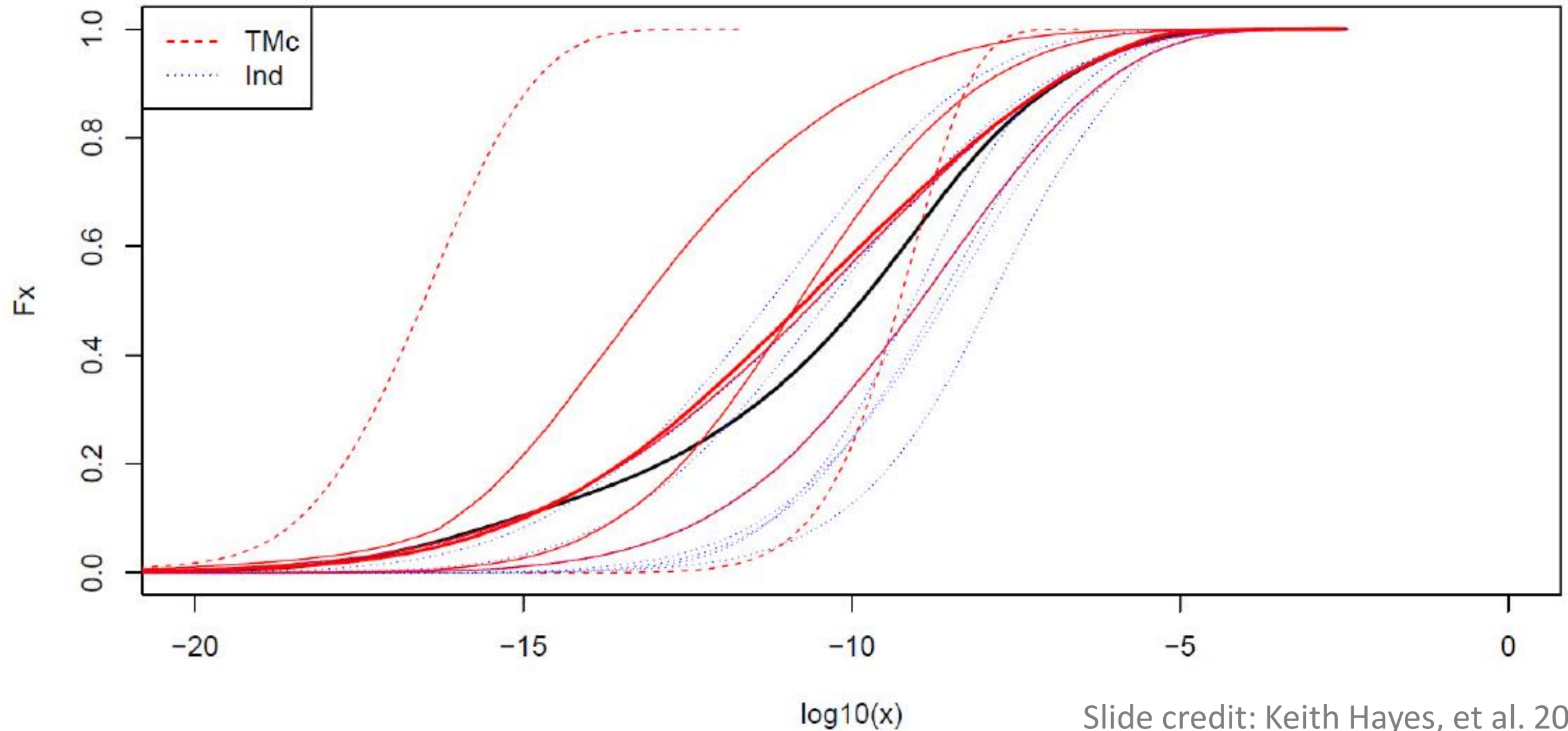
Source: Saey (2015)

Slide credit: Keith Hayes, et al. 2017



# Risk result: HGT to non-target Eukaryotes

FT3



# Confidence structure (c-box)

- P-box-shaped estimator of a (fixed) parameter
- Gives confidence interval at *any* confidence level
- Can be propagated just like p-boxes
- Allows us to *compute with confidence*



# Example: binomial rate $p$ for $k$ of $n$ trials

$$p \sim \text{env}(\text{beta}(k, n-k+1), \text{beta}(k+1, n-k))$$

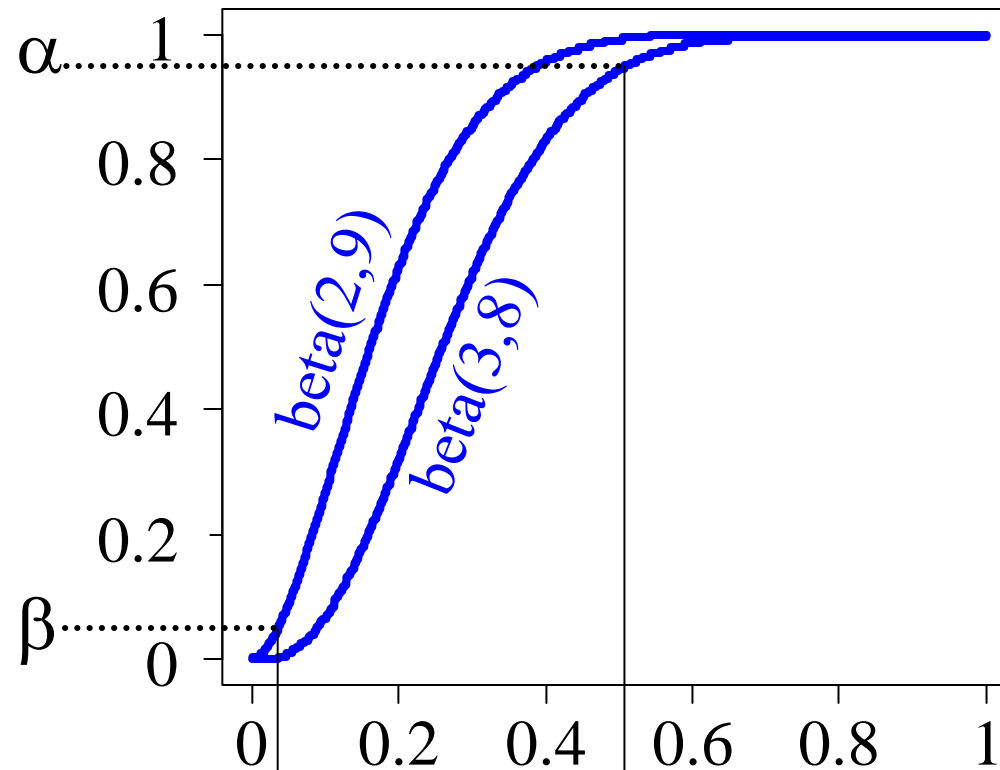
*Notation  
extends  
the use of  
tilda*

Data

$k = 2$  successes

$n = 10$  trials

$(\alpha - \beta)$  100%  
confidence  
interval for  $p$  →



If  $1 - \alpha = \beta$ , result is identical to classical Clopper–Pearson interval

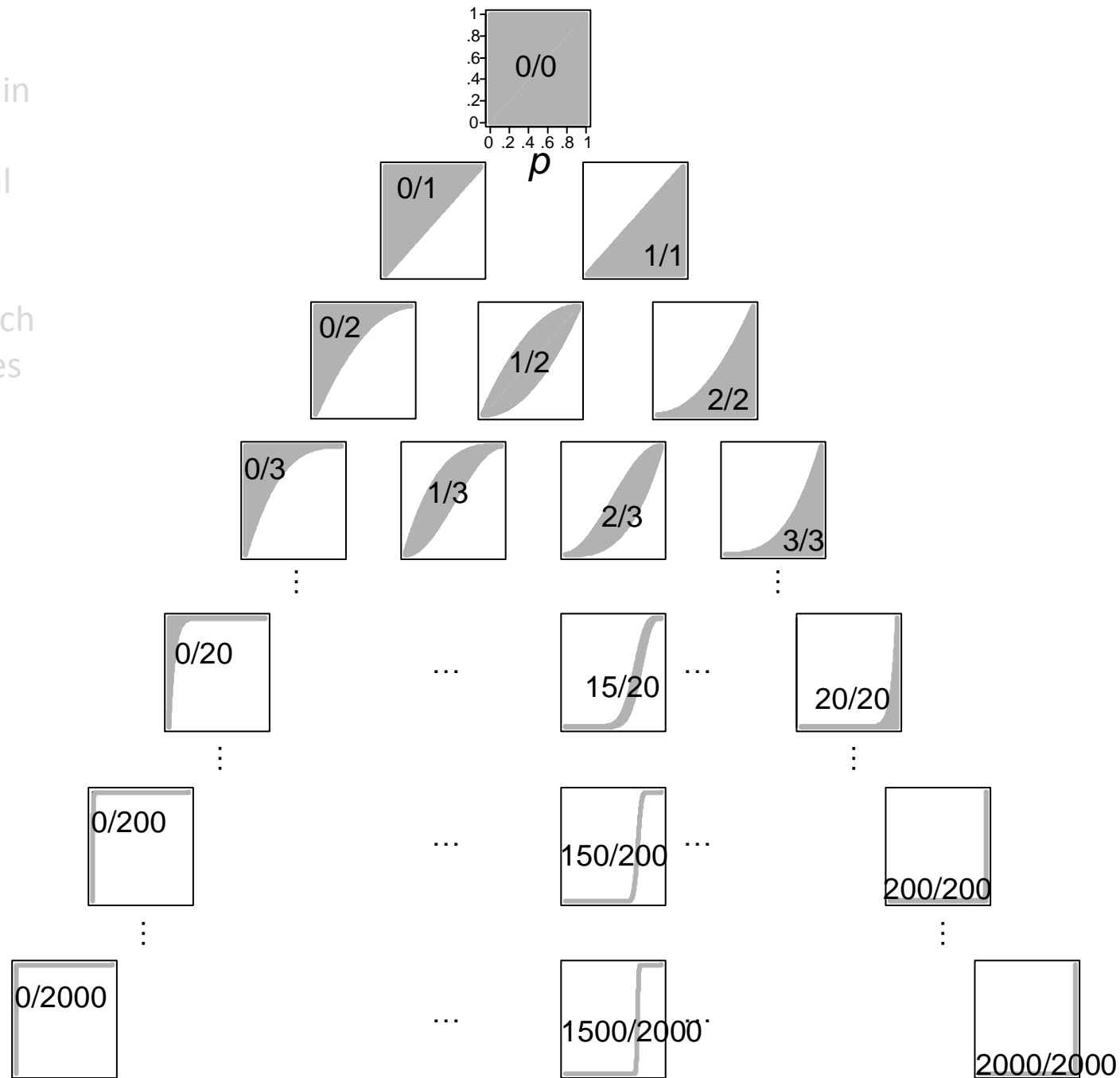
# C-boxes

- Bayesian (specifies a class of priors in the uninformative case)
- But also have frequentist coverage properties
- Don't optimize anything; they *perform*
- Characterize inputs from limited or even no information

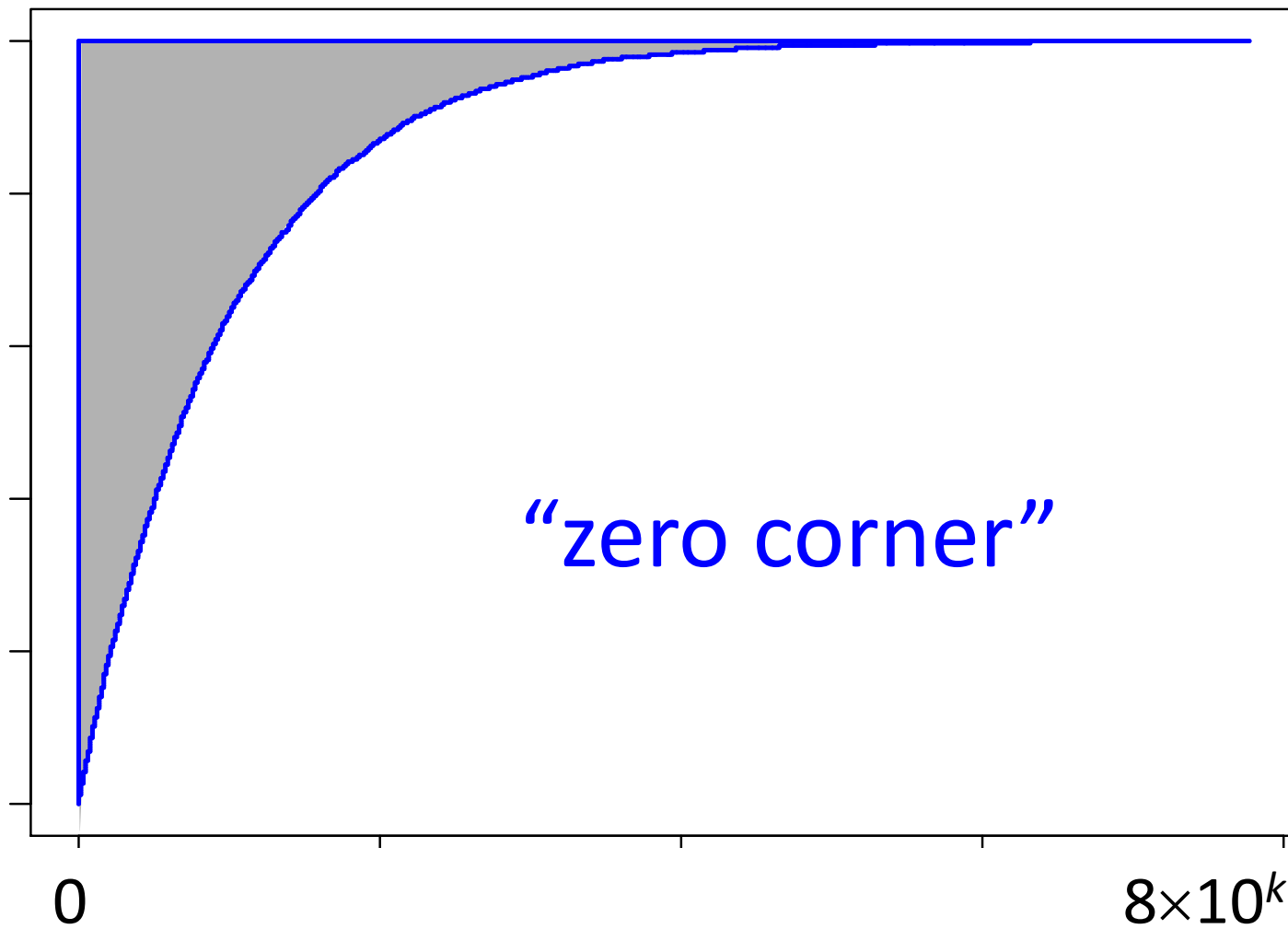
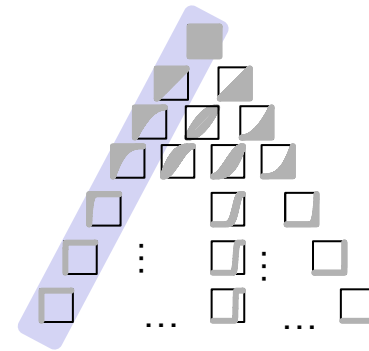
Notice that the c-boxes in every row partition the vacuous 'dunno' interval

We call the first (or first and second) c-box in each row the "corner" c-boxes

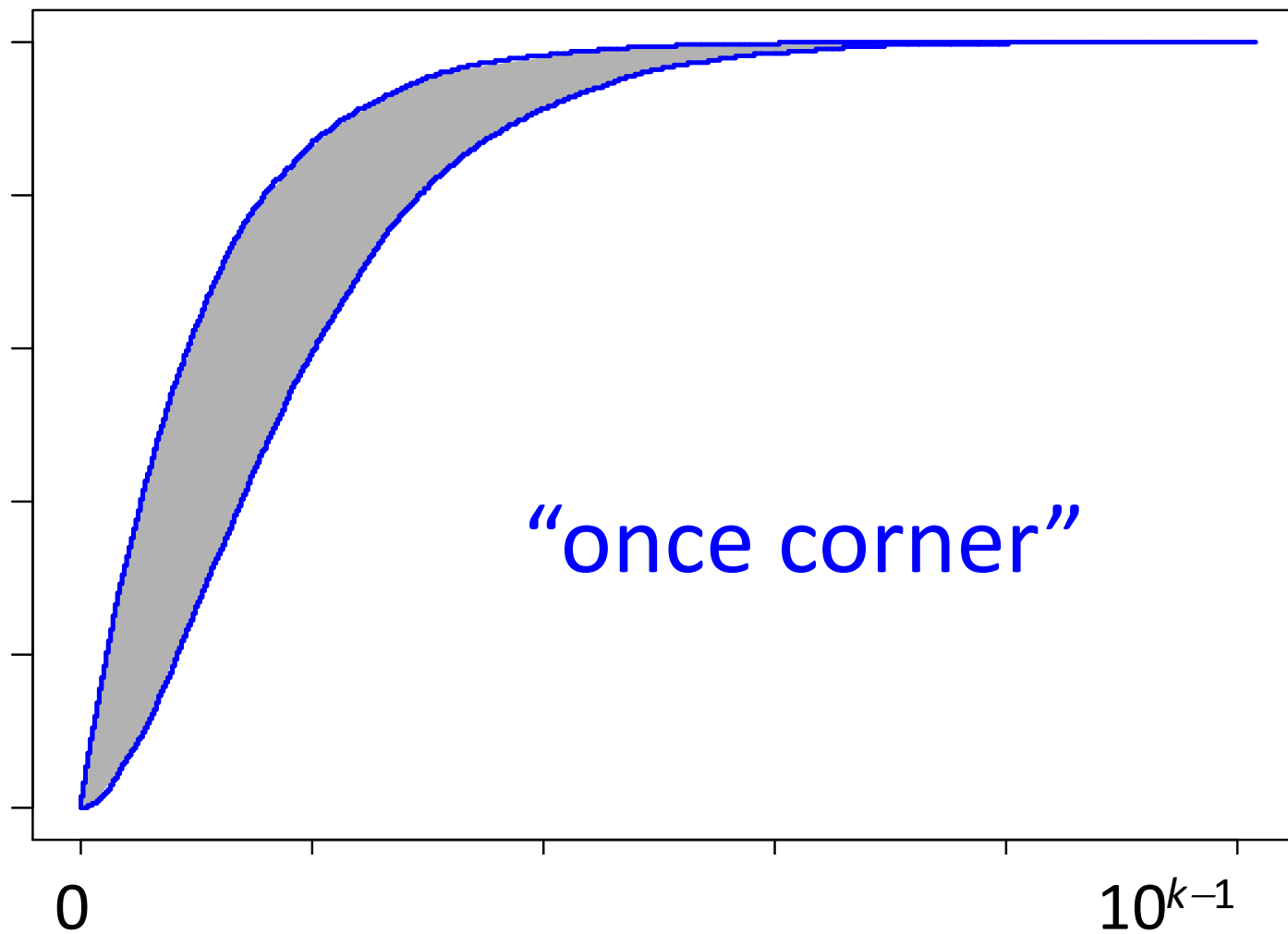
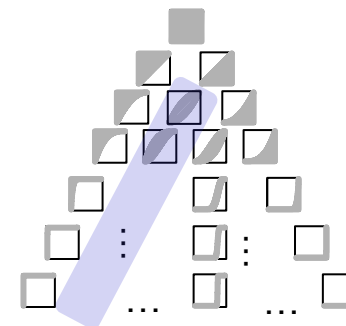
They correspond to the rare events of concern



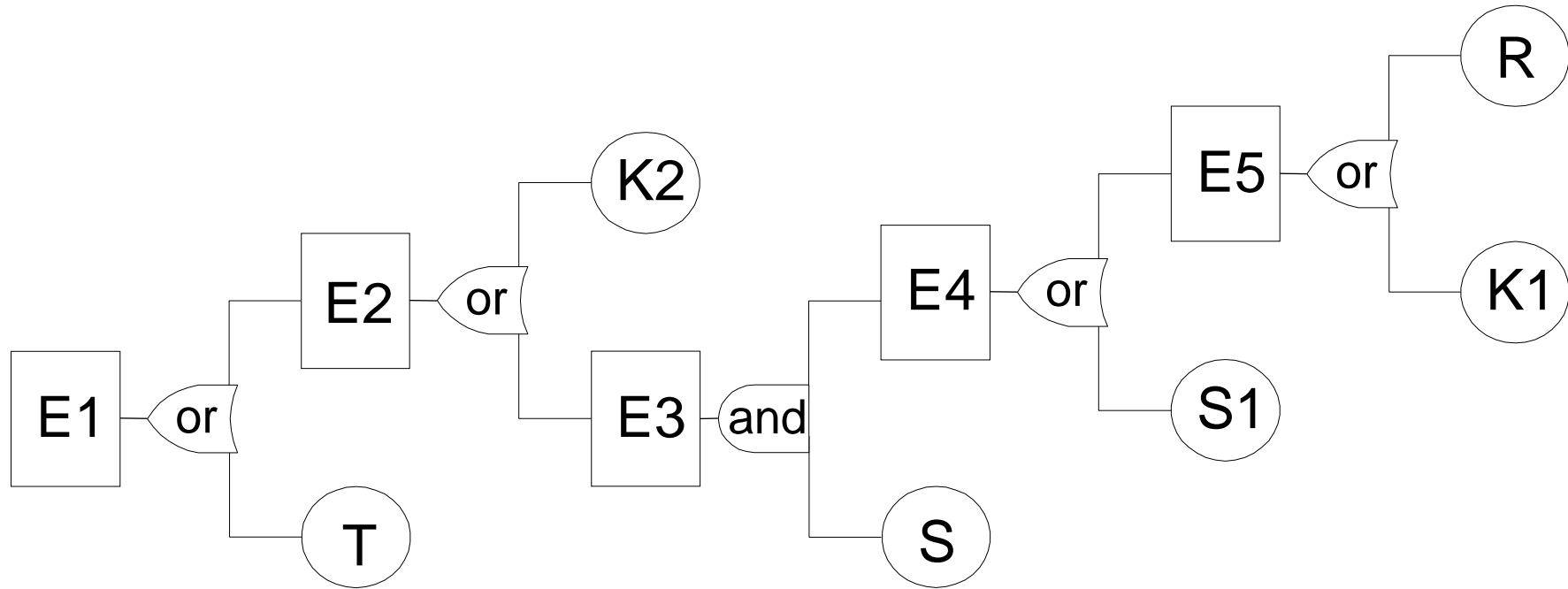
Zero out of  $10^k$  trials



One out of  $10^k$  trials



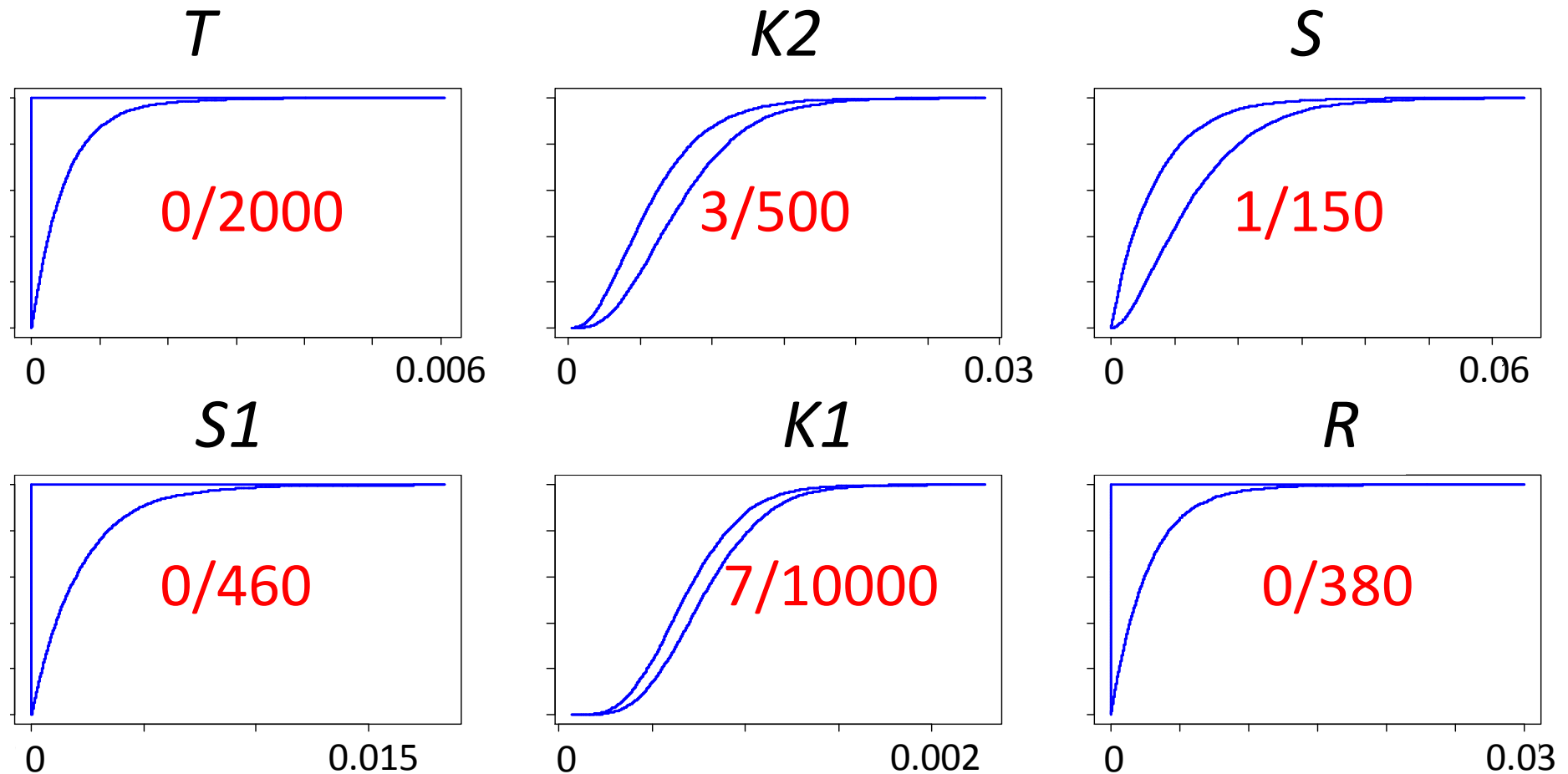
# Fault tree



$$E1 = T \vee (K2 \vee (S \& (S1 \vee (K1 \vee R))))$$

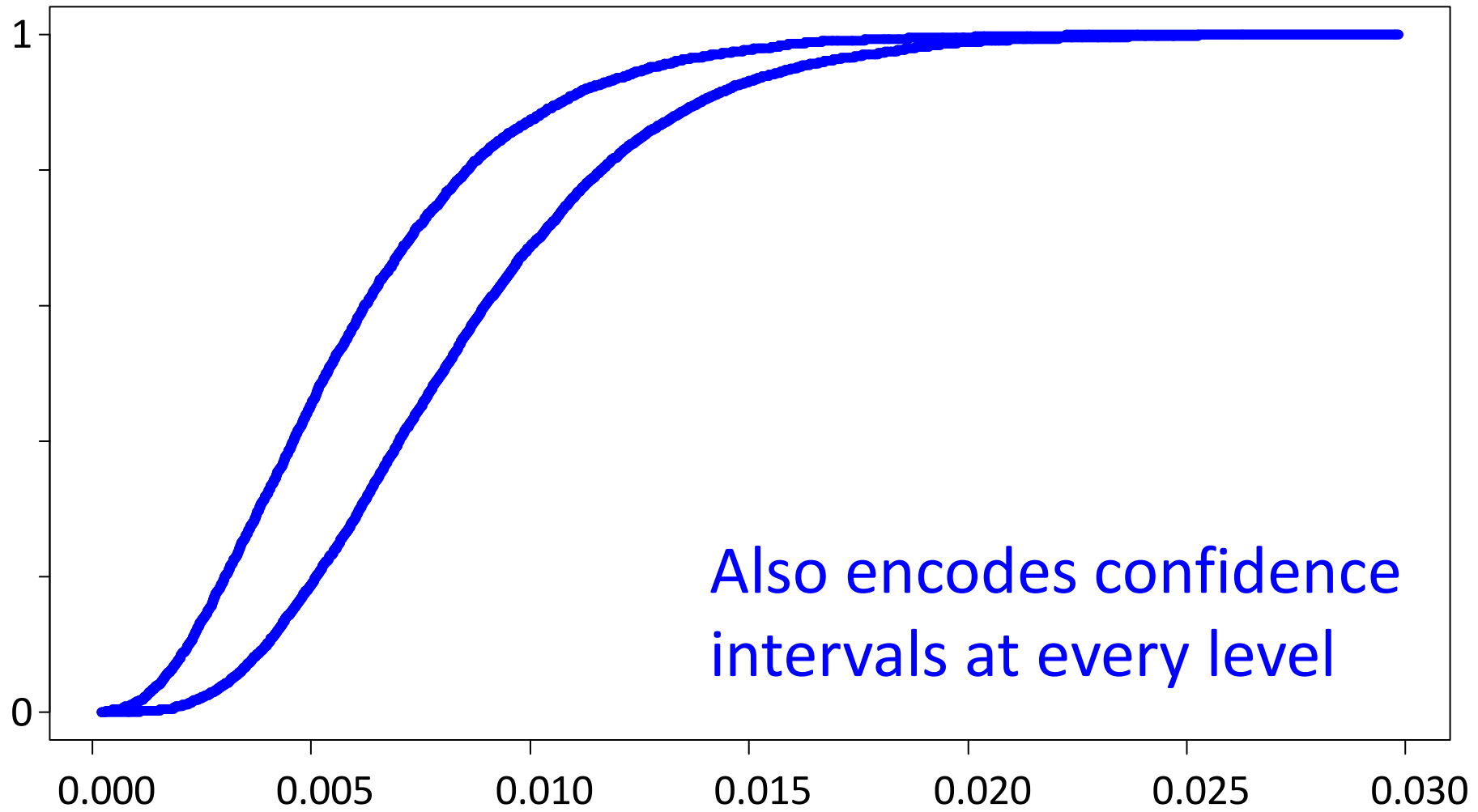


# Fault tree inputs



The blue c-boxes are posteriors for the inputs from a robust Bayes analysis based on the red data

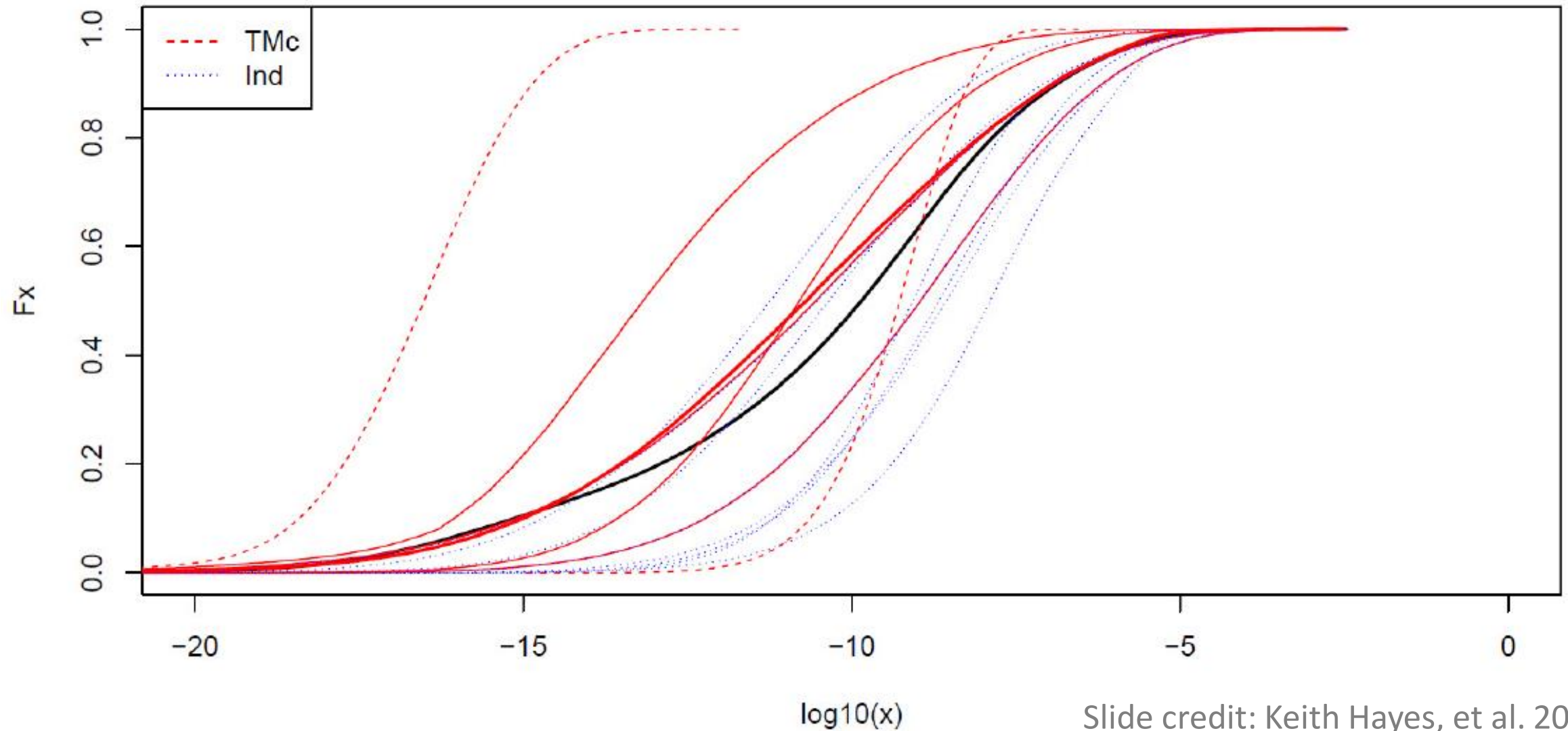
Top event  $E1$



Also encodes confidence intervals at every level

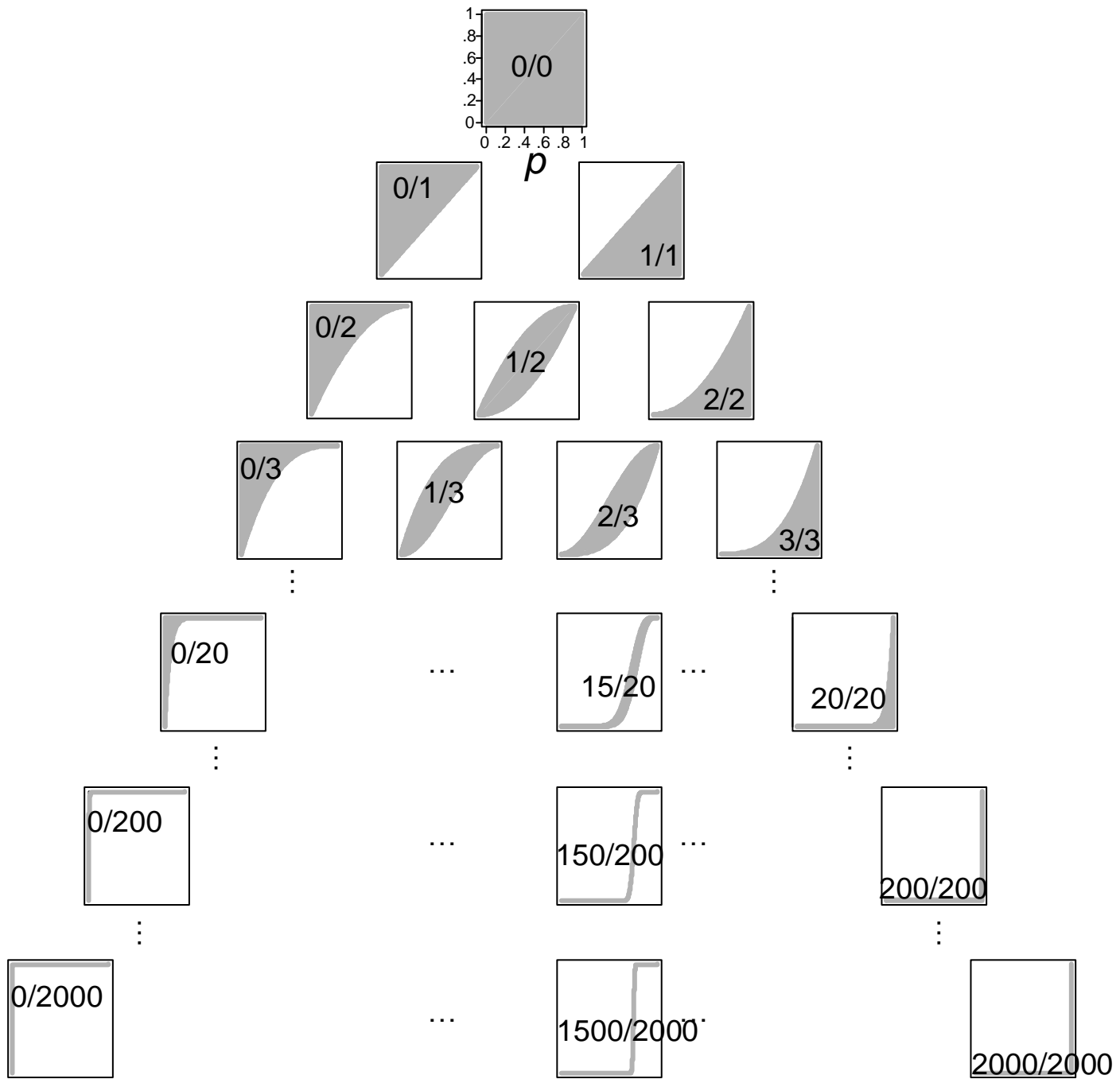
# Risk result: HGT to non-target Eukaryotes

FT3

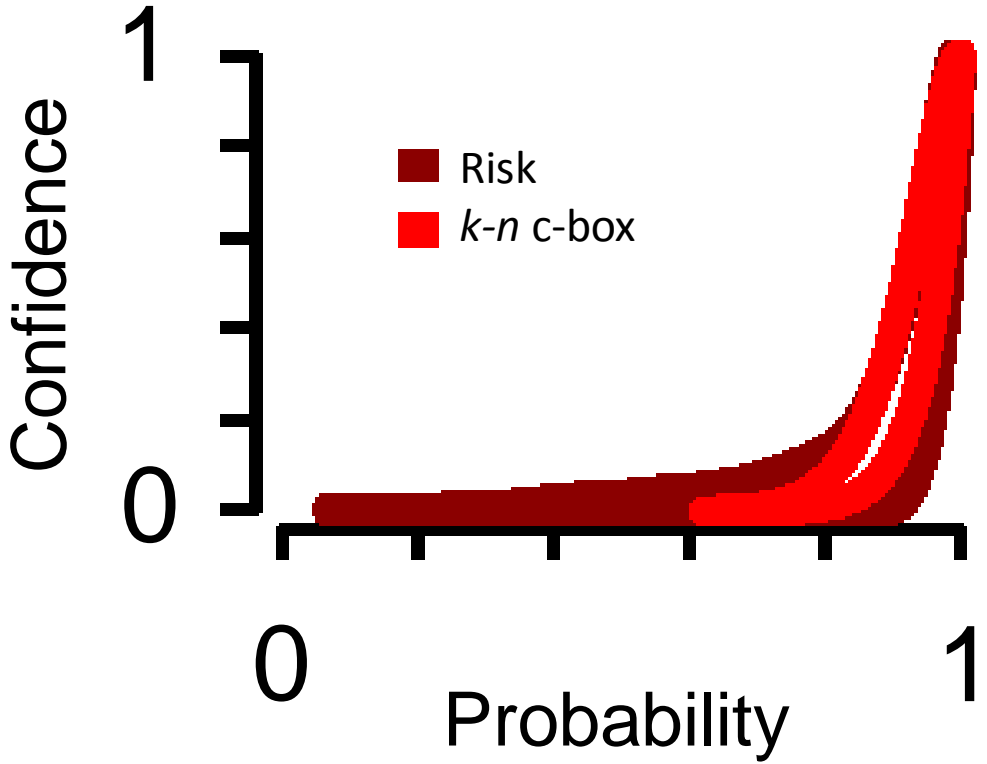


# Equivalent binomial count

- An imprecisely computed risk can be expressed as a p-box on  $[0,1]$
- Transform it into a natural language expression “ *k out of n* ”
- These are *natural frequencies*
  - Ratio  $k/n$  implies magnitude of risk
  - Large uncertainties imply small denominators
- People can understand them



Match a calculated risk to a c-box



Express risk as “*k* out of *n*”

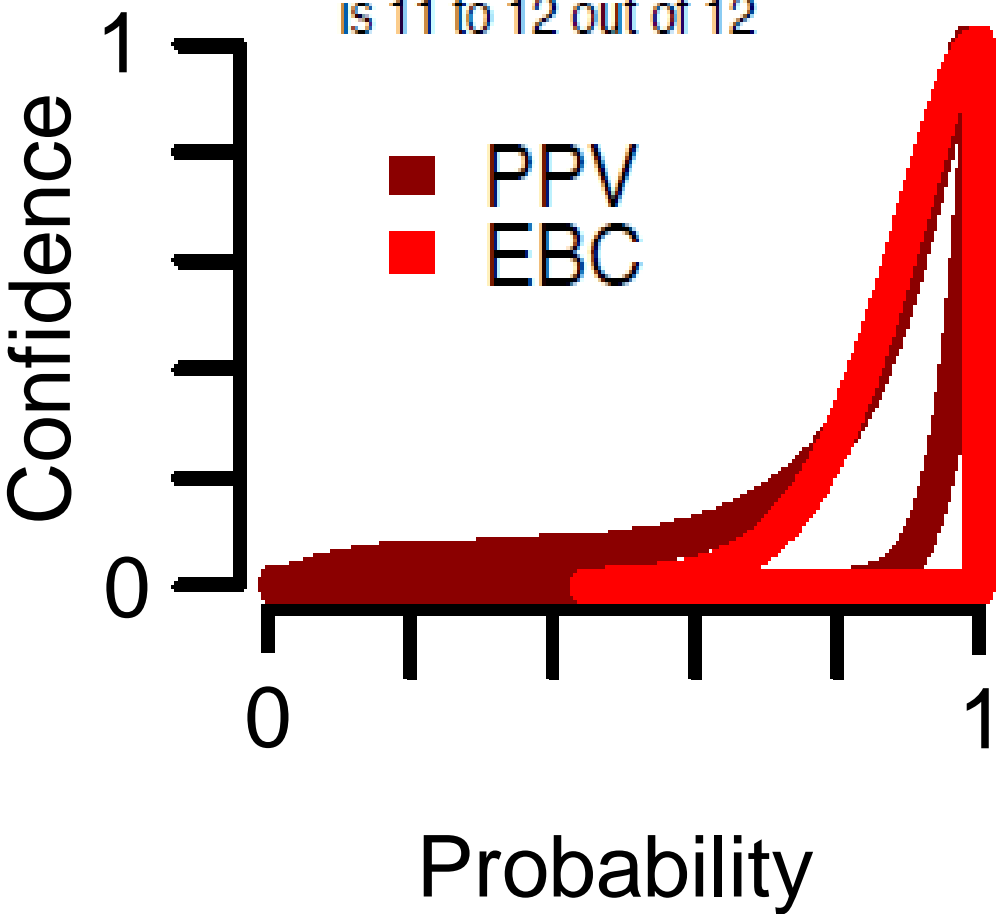


When there is a lot of epistemic uncertainty

- It might be possible to use intervals as numerators  $k$

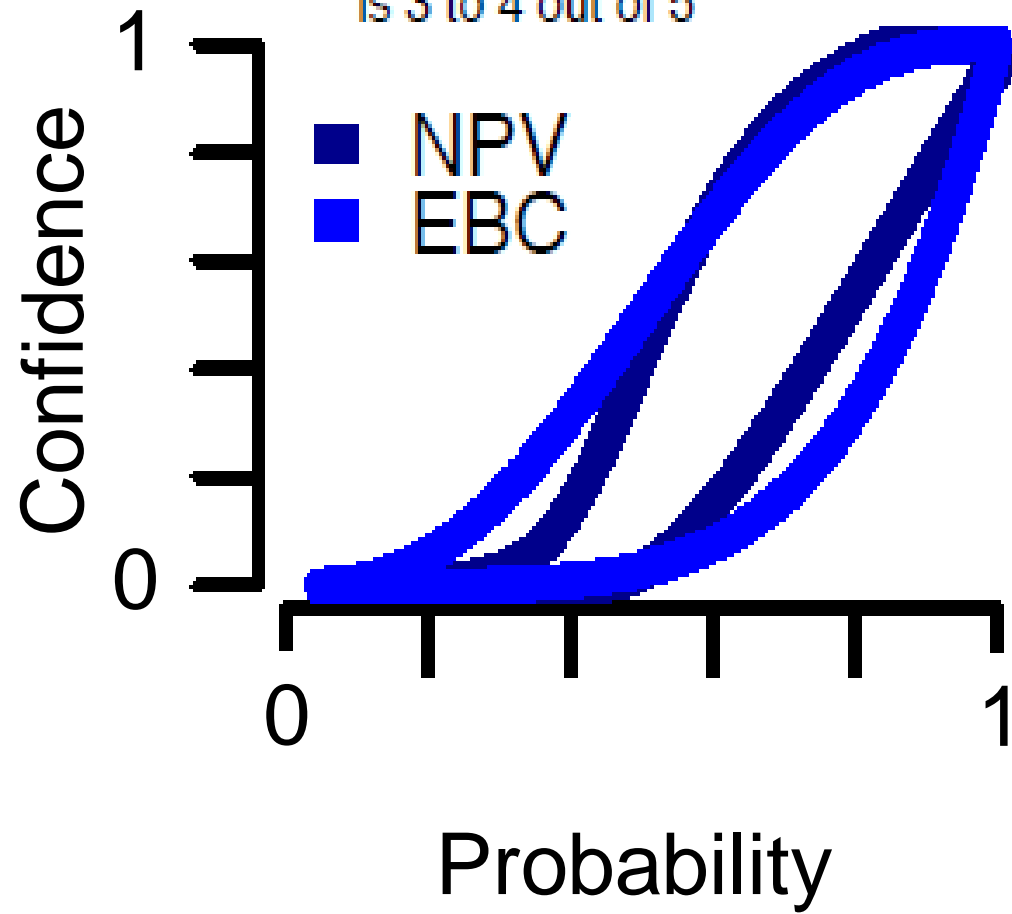
# PPV

The chance the patient is sick is 11 to 12 out of 12



# NPV

The chance the patient is well is 3 to 4 out of 5



# Do people understand?

- We used Amazon Mechanical Turk to check this
- We showed >300 “turkers” several mock sunglasses comparisons
- We checked the turkers’ preferences for identical sunglasses rated by other buyers using various schemes
- More frequent ‘excellent’ ratings should be preferred
- Larger pool of buyers rating should be more reliable

# We tested whether

- Turkers can make rational choices
- Natural frequencies are as good as or better than percentages
- Larger denominators convey more reliability
- Interval numerators can be understood

## Which product is better?

- Based on the reviews left by previous customers, which product would you buy?
- Use only the customer ratings and the number of stars left by customers to guide your decision.



Pair A was rated excellent by 2 out of 4 customers.



50% of customers rated Pair B as excellent.

Which product would you buy? Pair A, or Pair B?

# Findings

80%

10/100

66/198

[66,88]/198

50%

2/4

50%

80/100

2/6

33/100

50/100

50%

rational

same sample size

same magnitude

even with ambiguity

prefer natural frequency

unless very uncertain

These are exemplar comparisons from the study with “master turkers”



# Conclusions

- People make rational choices when given natural frequencies “ $k$  out of  $n$ ”
- Analysts can translate results into  $k$ -out-of- $n$  equivalent binomial counts
- Natural frequencies express probabilities so humans can understand them
- Also embody uncertainty about the risks which humans also care about

# Acknowledgments

- Keith Hayes
- Jason O'Rawe
- Michael Balch
- National Institutes of Health

End