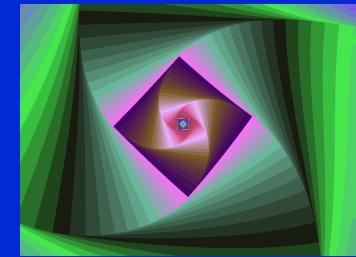
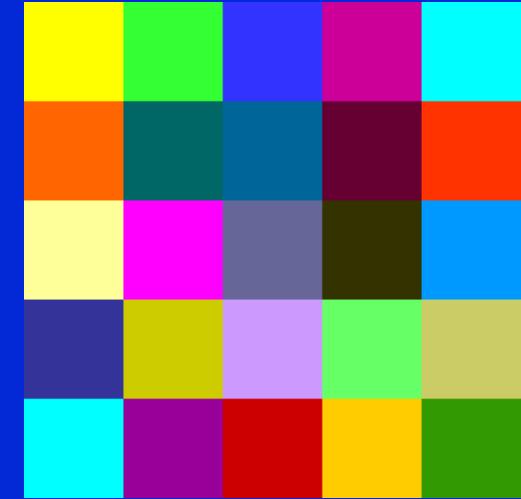
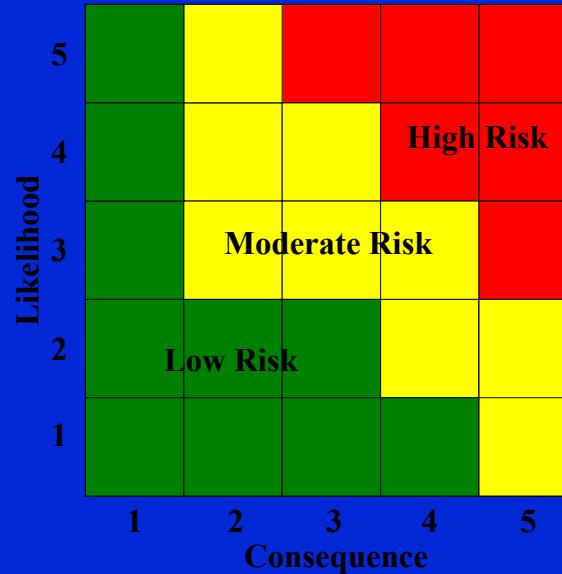


+

Cognitive Biases in the Risk Matrix



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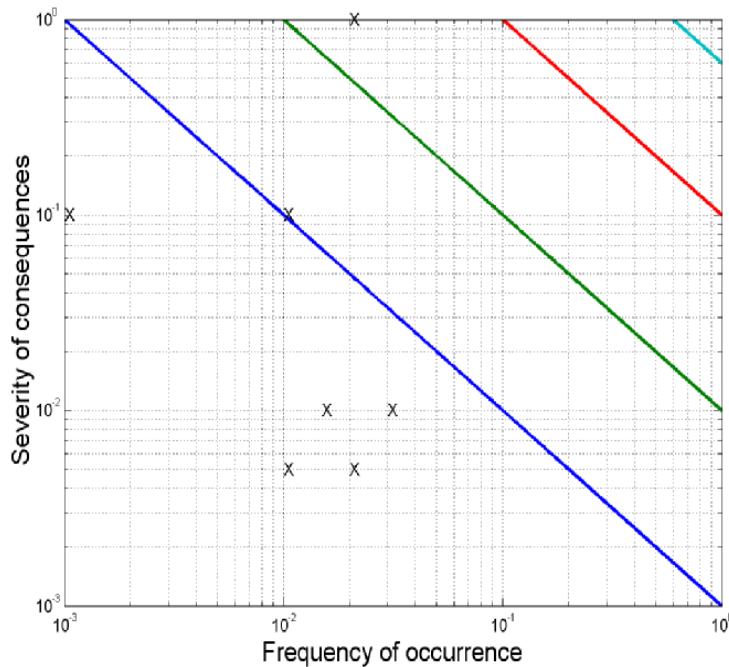
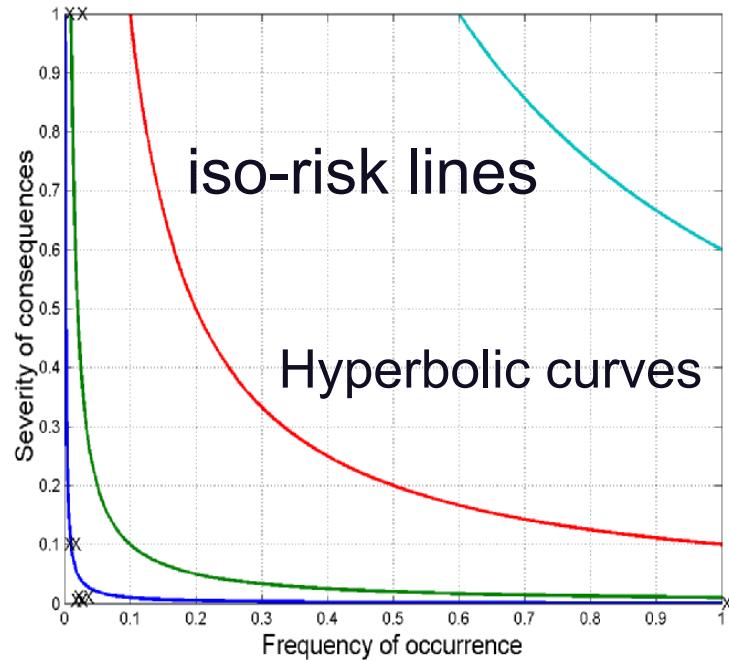
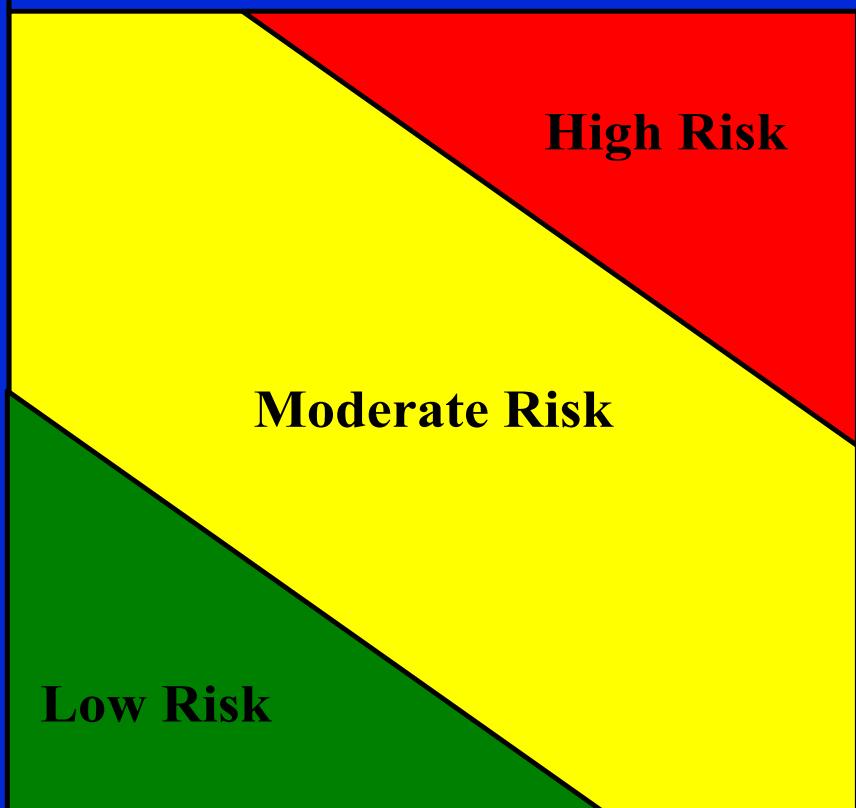
“Fear of harm ought to be proportional not merely to the **gravity** of the harm, but also to the **probability** of the event.”

Logic, or the Art of Thinking
Antoine Arnould, 1662

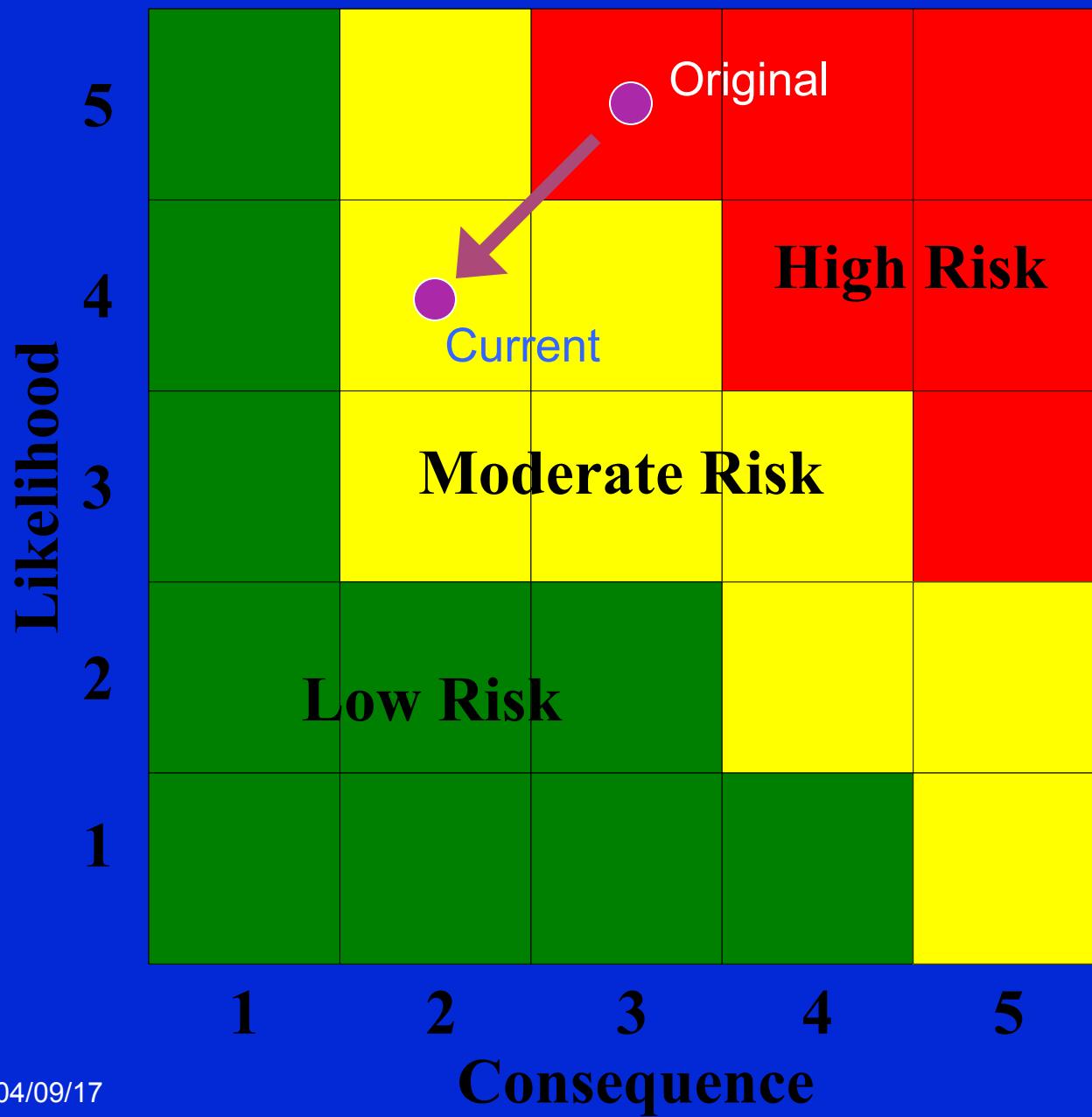
Consequence x Likelihood = Risk

Risk graphing

Likelihood of bad outcome



5 x 5 Risk “Cube”



Objective
vs.
Subjective
data

"Campfire
conversation"
piece

Present Situation

- Risk matrices are recognized by industry as the best way to:
 - consistently quantify risks, as part of a
 - repeatable and quantifiable risk management process
- Risk matrices involve human:
 - ❖ Numerical judgment
 - Calibration – location, gradation
 - Rounding, Censoring
 - ❖ Data updating
 - often approached with under confidence
 - often distrusted by decision makers

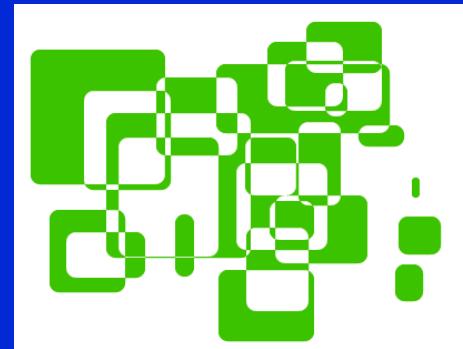
Goal

- Risk Management improvement and better use of the risk matrix
 - Confidence in correct assessment of probability and value
 - Avoidance of specific mistakes
 - Recommended actions

Heuristics and Biases

Daniel Kahneman won the Nobel Prize in Economics in 2002 "for having integrated insights from psychological research into economic science, especially concerning human judgment and decision-making under uncertainty."

Similarities between cognitive bias experiments and the risk matrix axes show that risk matrices are susceptible to human biases.



Anchoring

- First impression dominates all further thought
- SS# Bias
- Questionnaire included last two digits of SS#
 - Answers to questions biased
 - ❖ High last two digits showed higher estimations
 - ❖ Lower last two digits showed lower estimations
- Obviating expert opinion
 - The analyst holds a circular belief that expert opinion or review is not necessary because no evidence for the need of expert opinion is present.

Heuristics and Biases

Presence of cognitive biases

– even in extensive and vetted analyses –
can never be ruled out.

Innate human biases, and exterior circumstances, such as the framing or context of a question, can *compromise* estimates, judgments and decisions.

It is important to note that subjects often maintain a *strong sense* that they are acting rationally while exhibiting biases.

Terminology

Subjective Parameters

Likelihood (L)

Consequence (C)

Subjective
Probability, $\pi(p)$

Utility (negative),
 $U^-(v)$

Y axis

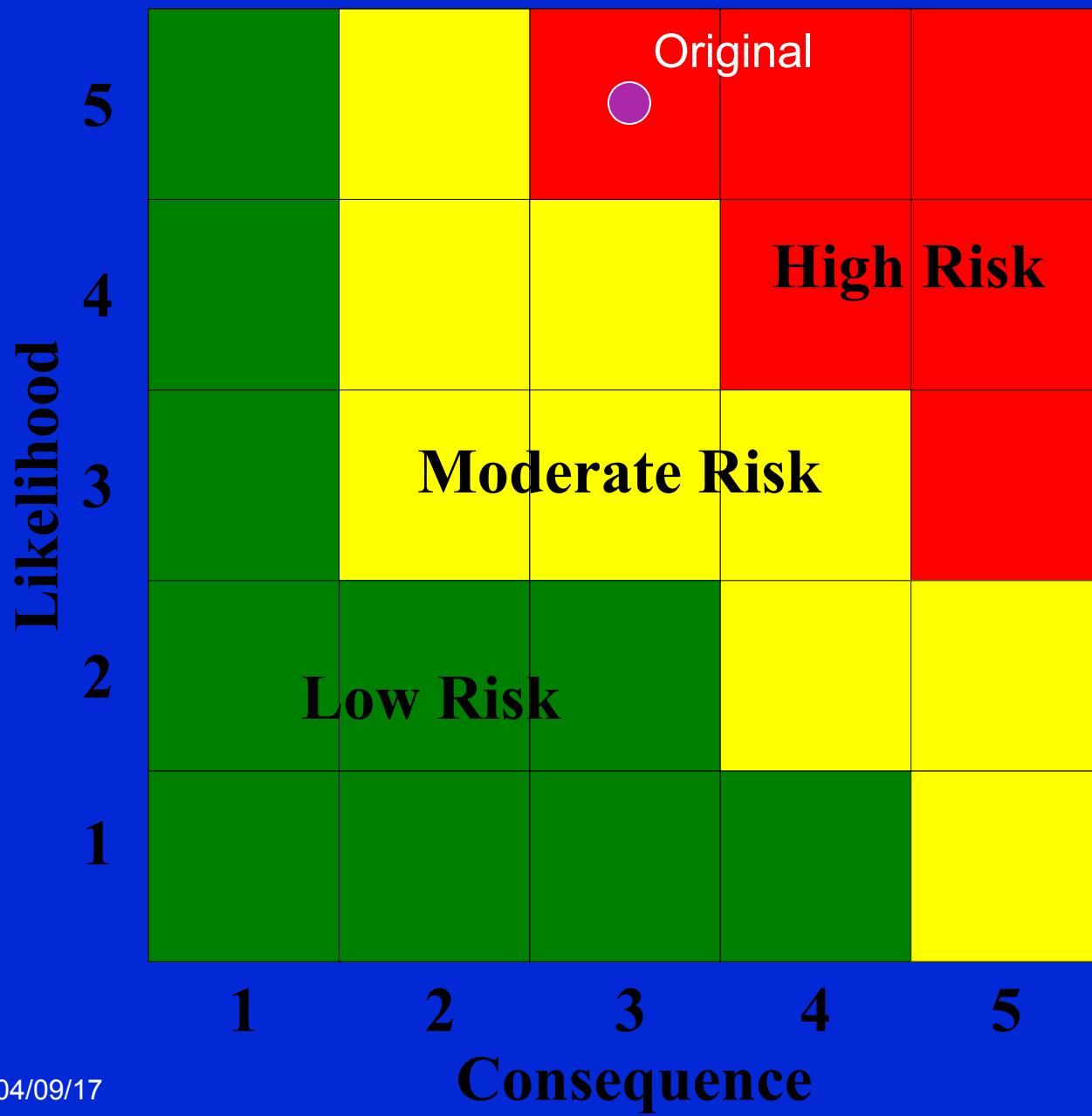
X axis

Objective Parameters

Objective Probability,
 p

Objective
Value, v

5 x 5 Risk “Cube”



Objective
vs.
Subjective
data

Communication
Tool

Likelihood

1. **Frequency of occurrence** is objective, discrete (coin toss)
2. **Probability** is continuous, fiction
 - ❖ "Humans judge probabilities poorly" [Cosmides and Tooby, 1996]
3. **Likelihood** is a subjective judgment
(unless mathematical)

Consequence, C

- Objective Consequence determination is costly
 - Total life-cycle cost increased by determination
- Mil-Std 882d

	\$ damage	Human impact	Environment	Law
Catastrophic	> \$1M	Death, Disability	irreversible damage	Violate
Critical:	\$1M - \$200K	Hospitalization to >= 3 personnel	Reversible damage	Violate
Marginal:	\$200K- \$10K	Loss of work days; injury	Mitigation damage	
Negligible:	\$10K-\$2K	No lost work day; injury	Minimal damage	

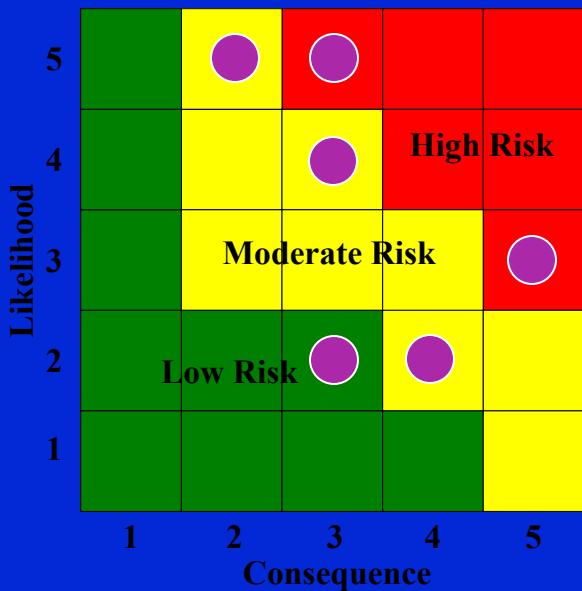
Case Study

- **Industry risk matrix data**
 - **1412 original and current risk points (665)**
 - ❖ Two programs used for raw data
 - ❖ Time of first entry known
 - ❖ Time of last update known
 - **Cost, Schedule and Technical Impact known**
 - **Risk Subject not known**
- **Biases revealed**
 - **Likelihood and consequence judgment**

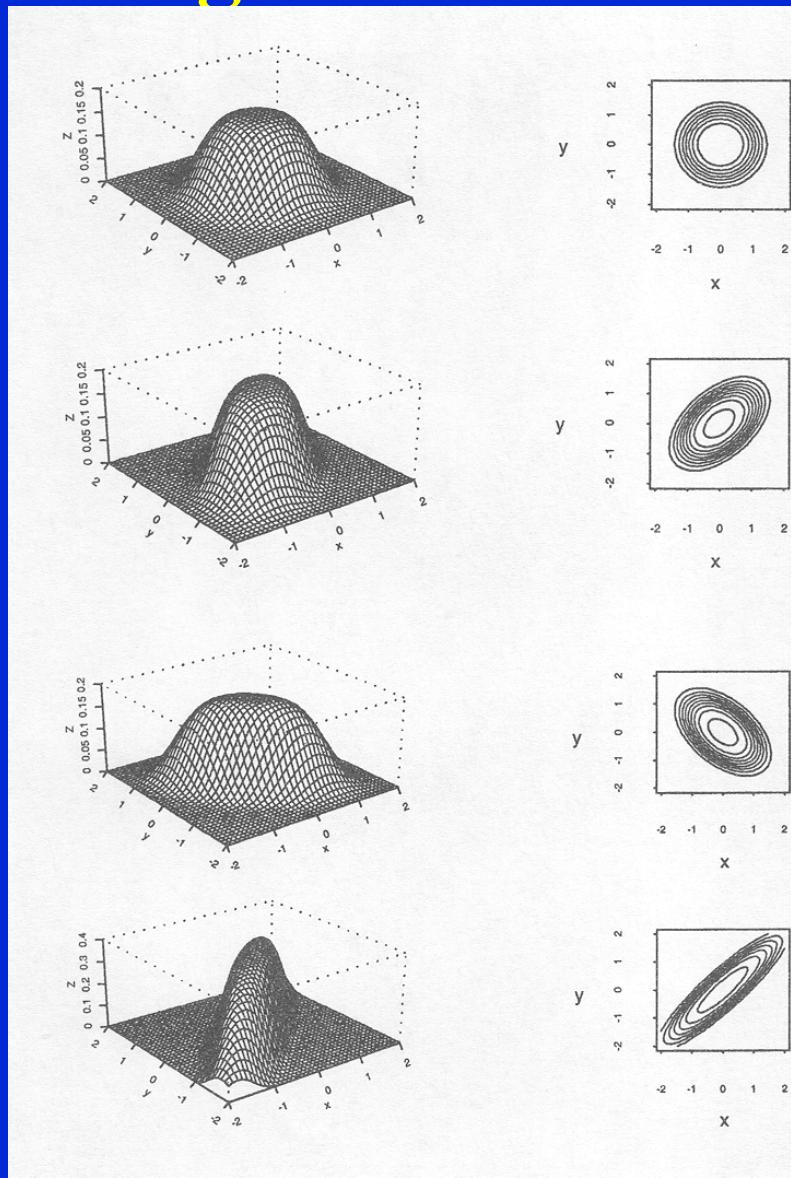
Magnitude vs. Reliability [Griffin and Tversky, 1992]

- Magnitude perceived more valid
 - Data with outstanding magnitudes but poor reliability selected and used
- Suggestion:
 - Data with uniform source reliability
 - ❖ Speciousness of data
- Observation: Risk Matrices are magnitude driven, without regard to reliability

Expected Distribution for original risk points in Risk Matrix?



Uniform:



Bivariate Normal

1. Estimation in a Pre-Define Scale Bias

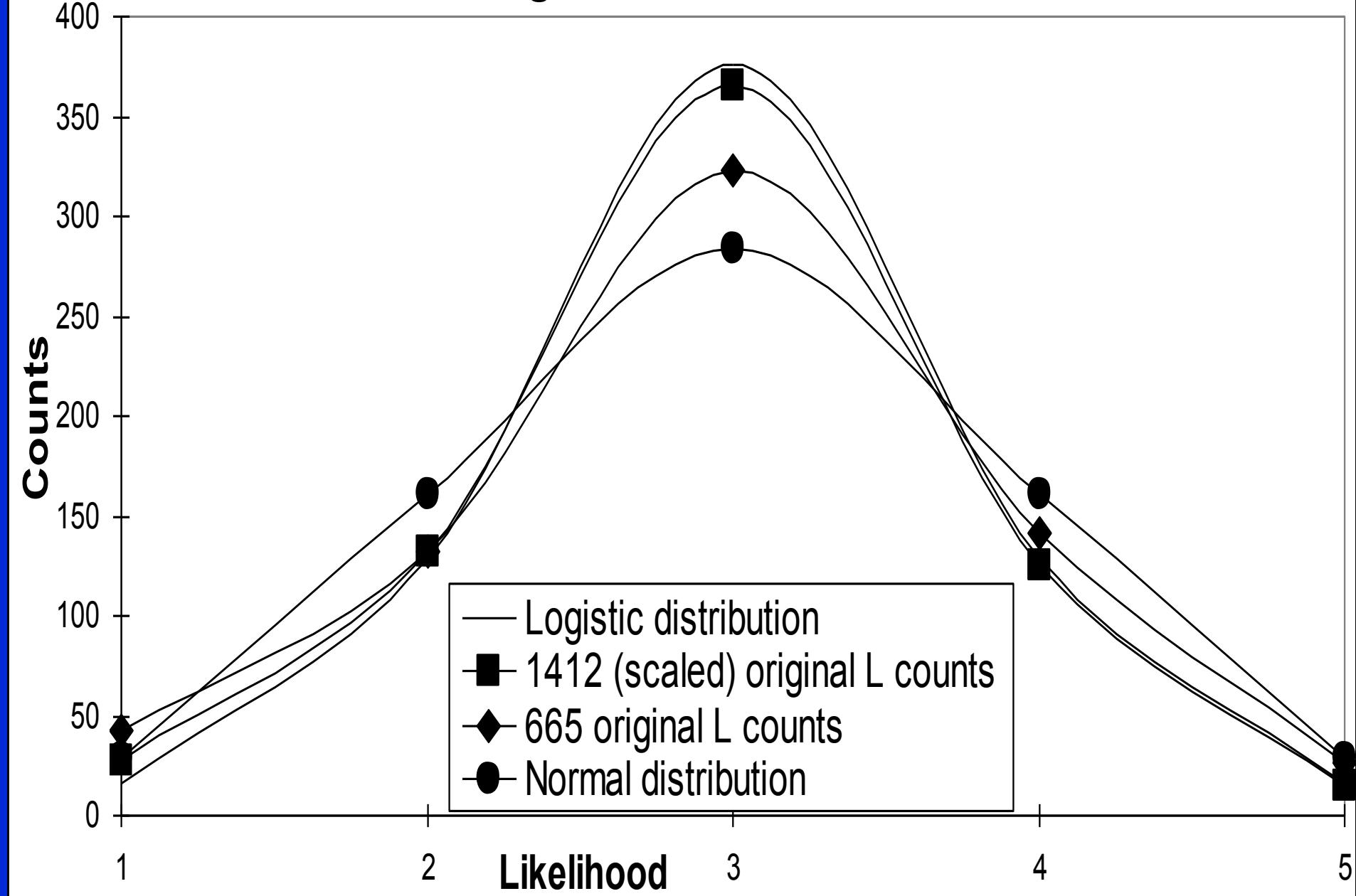
- Response scale effects judgment [Schwarz, 1990]
- Two questions, random 50% of subjects:
- Please estimate the average number of hours you watch television per week:



- Please estimate the average number of hours you watch television per week:

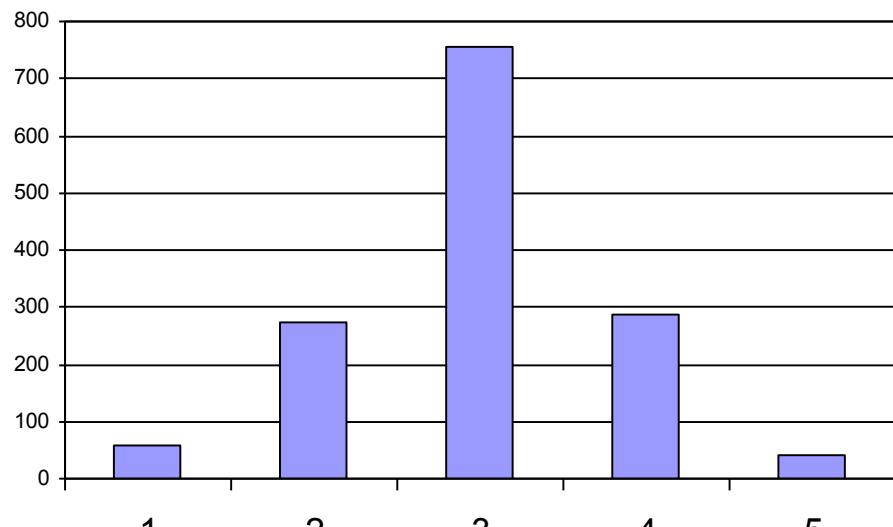


Likelihood: Logistic, 1412, 665, and Normal

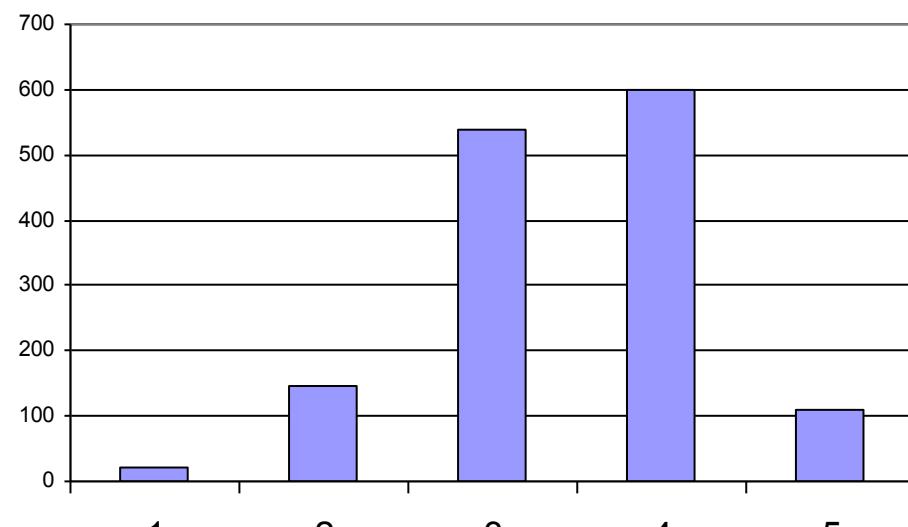


Effect of Estimation in a Pre-Defined Scale

Likelihood count in original risks



Consequence count in original risks



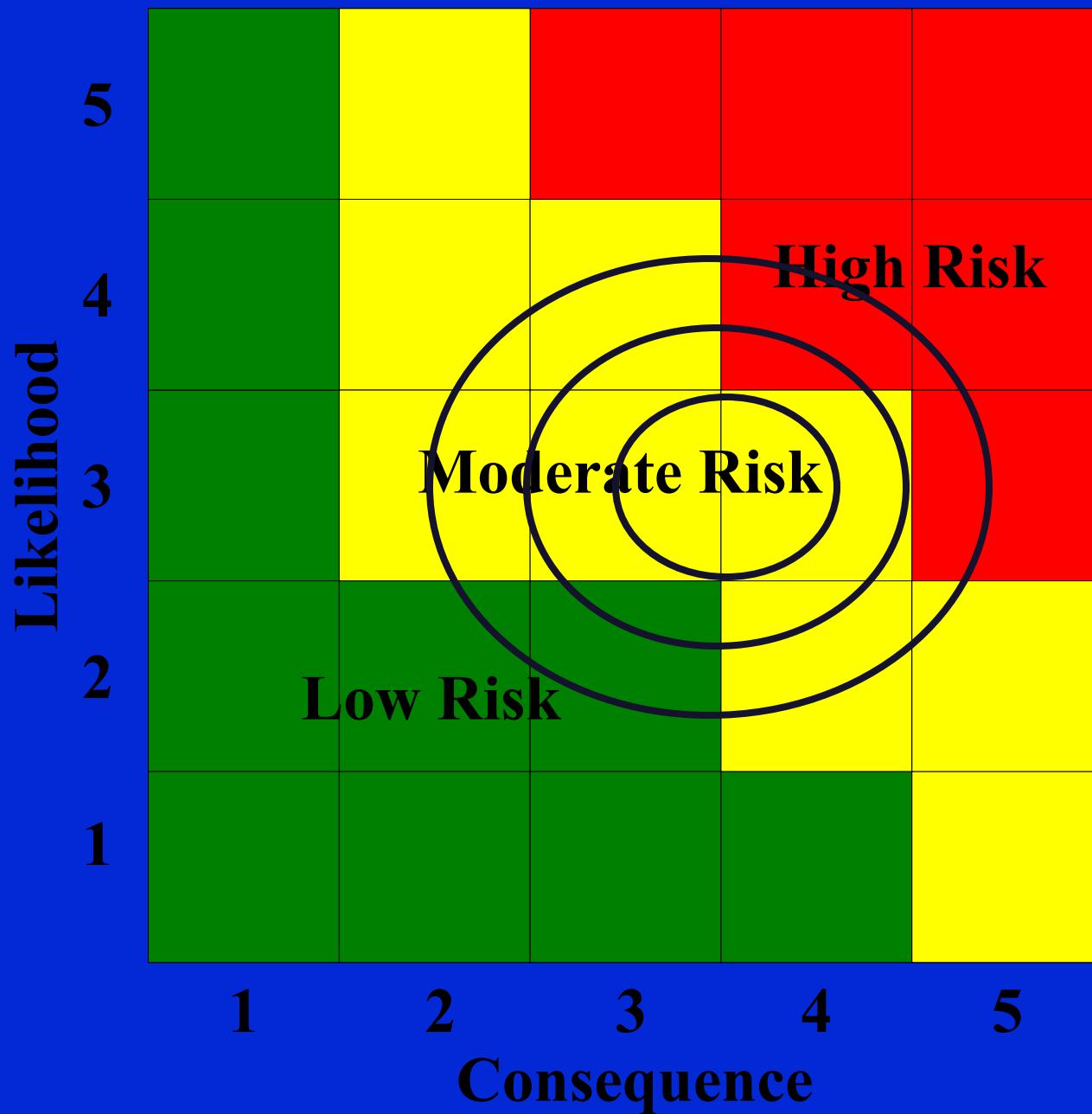
'People estimate probabilities poorly'
[Cosmides and Tooby, 1996]

Consequence/Severity amplifiers

Consequence Amplifiers

- Lack of control
- Lack of choice
- Lack of trust
- Lack of warning
- Lack of understanding
- Manmade
- Newness
- Dreadfulness
- Personalization
- Recallability
- Eminency

5 x 5 Risk Matrix



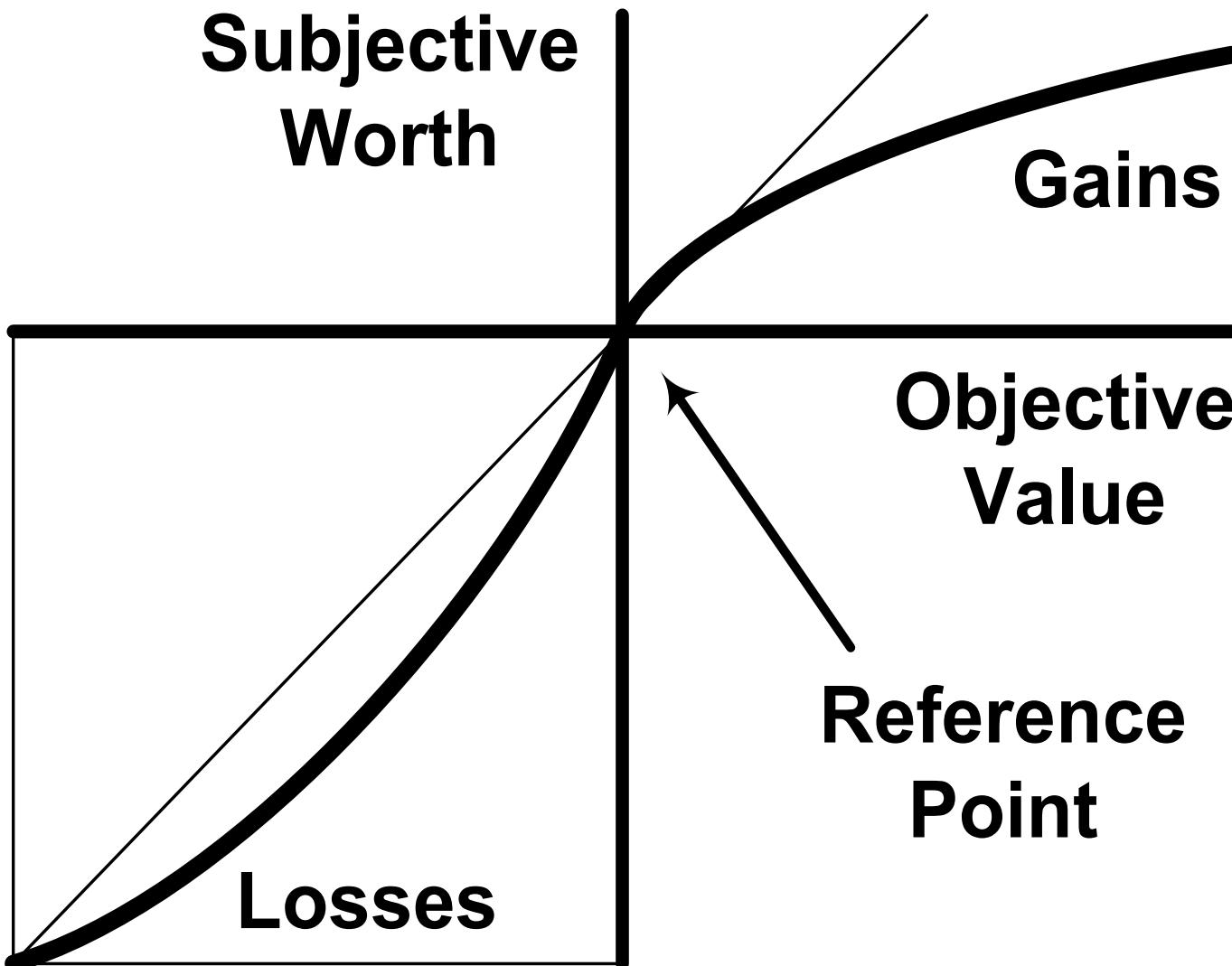
Situation assessment

- **5 x 5 Risk Matrices seek to increase risk estimation consistency**
- **Hypothesis: Cognitive Bias information can help improve the validity and sensitivity of risk matrix analysis**

Prospect Theory

- Decision-making described with subjective assessment of:
 - Probabilities
 - Values
- Prospect Theory:
 - Probabilities and values are subjectively assessed

Gains and losses are not equal*



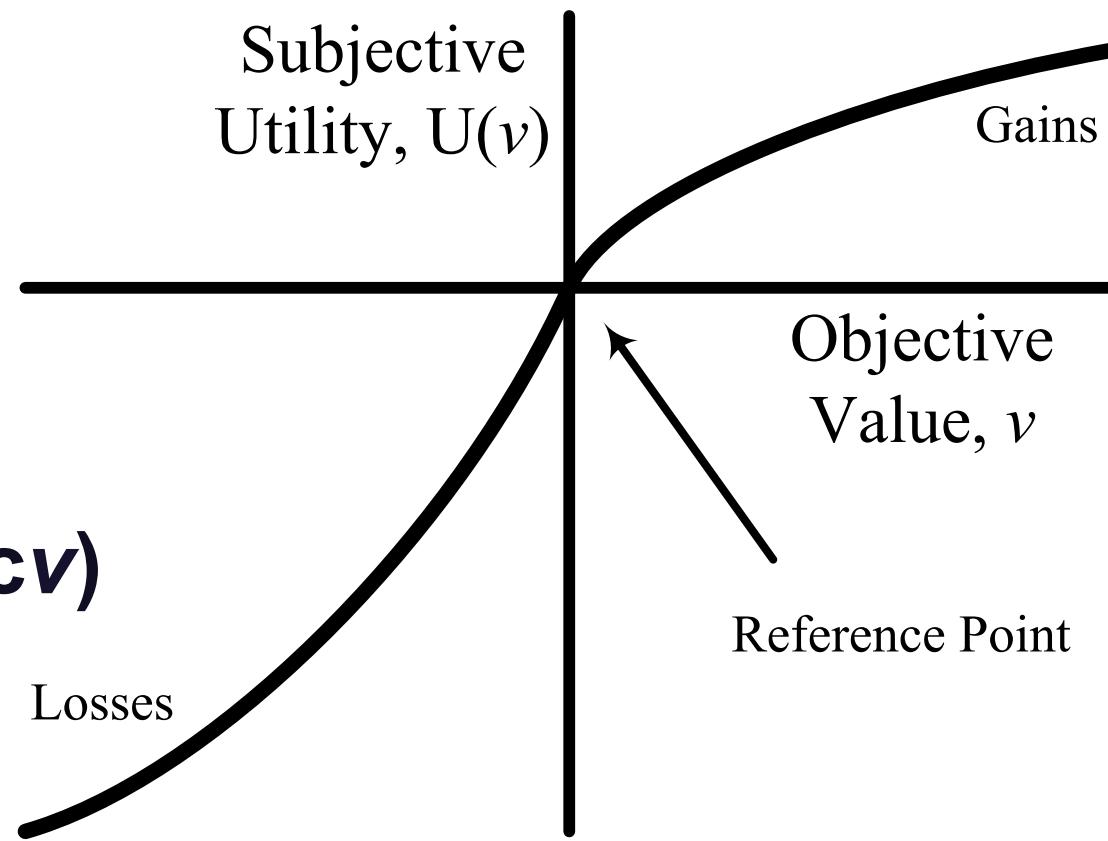
Subjective Utility

- Values considered from reference point established by the subject's wealth and perspective
 - Framing
- Gains and losses are subjectively valued
 - 1-to-2 ratio.
- For gains:
$$U^+(v) = \ln(1 + v)$$
- For losses:
$$U^-(v) = -(\mu) \ln(1 - cv)$$

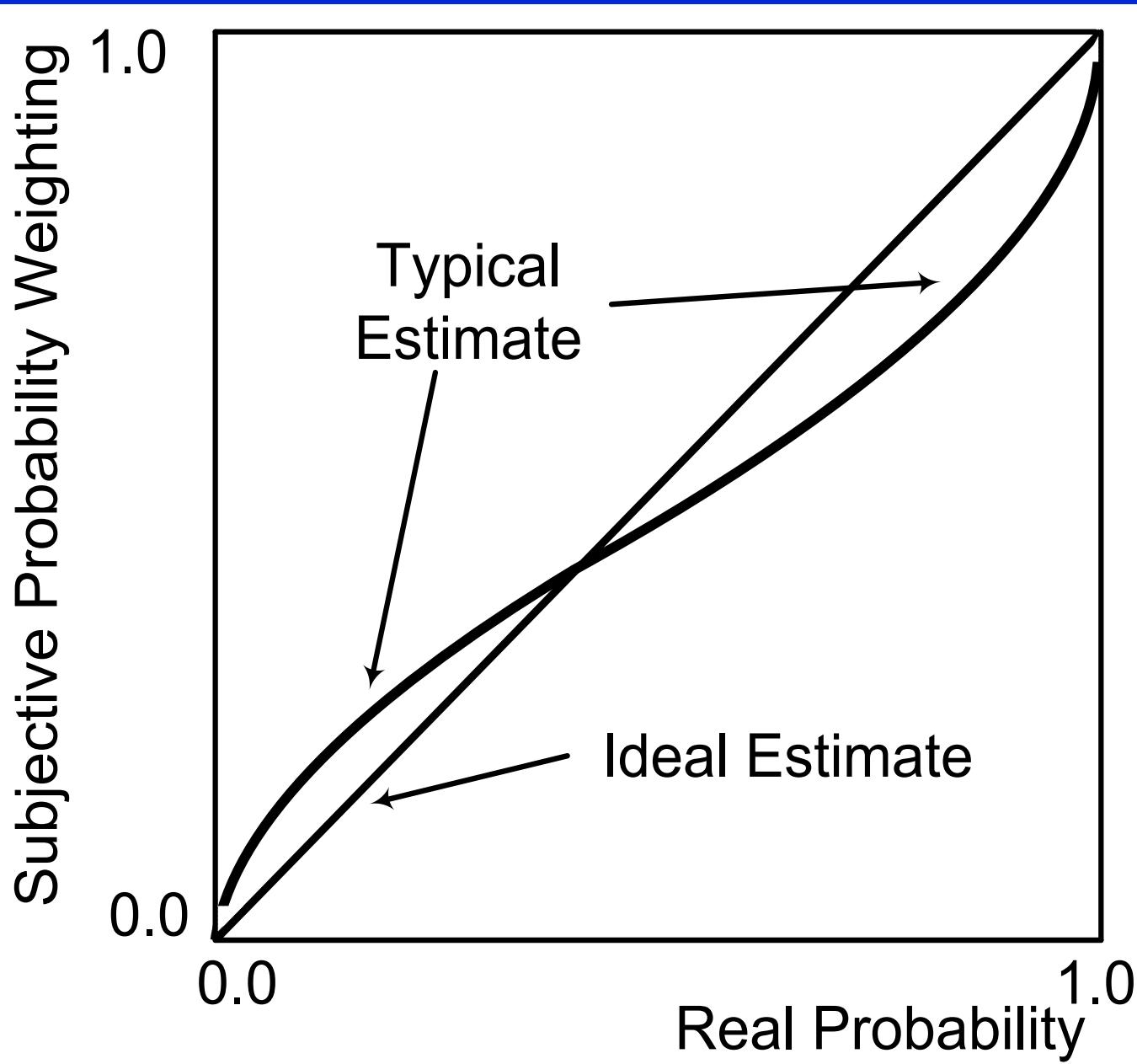
$$\mu = 2.5$$

$$c = \text{constant}$$

$$v = \text{objective value}$$



Humans judge probabilities poorly*

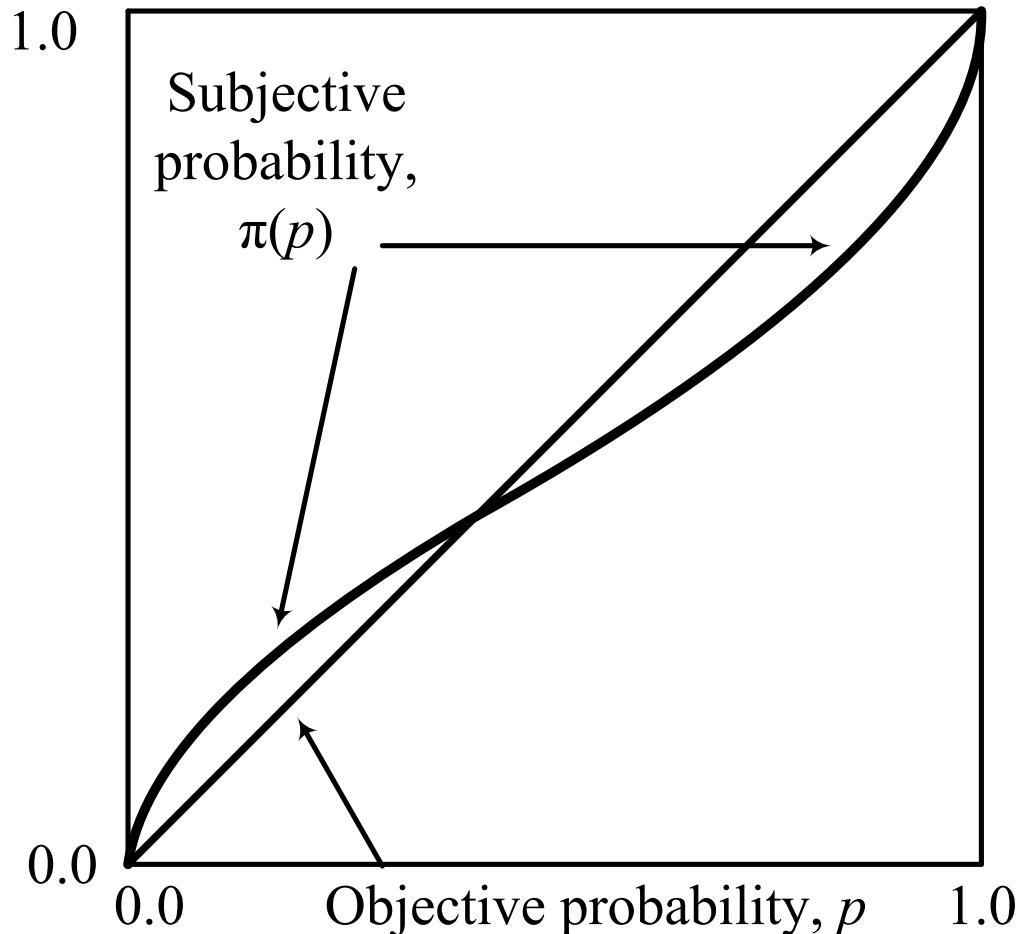


Subjective Probability, $\pi(p)$

- **small probabilities overestimated**
- **large probabilities underestimated**

$$\pi(p) = \frac{(p^\delta)}{[p^\delta + (1-p)\delta]}^{(1/\delta)}$$

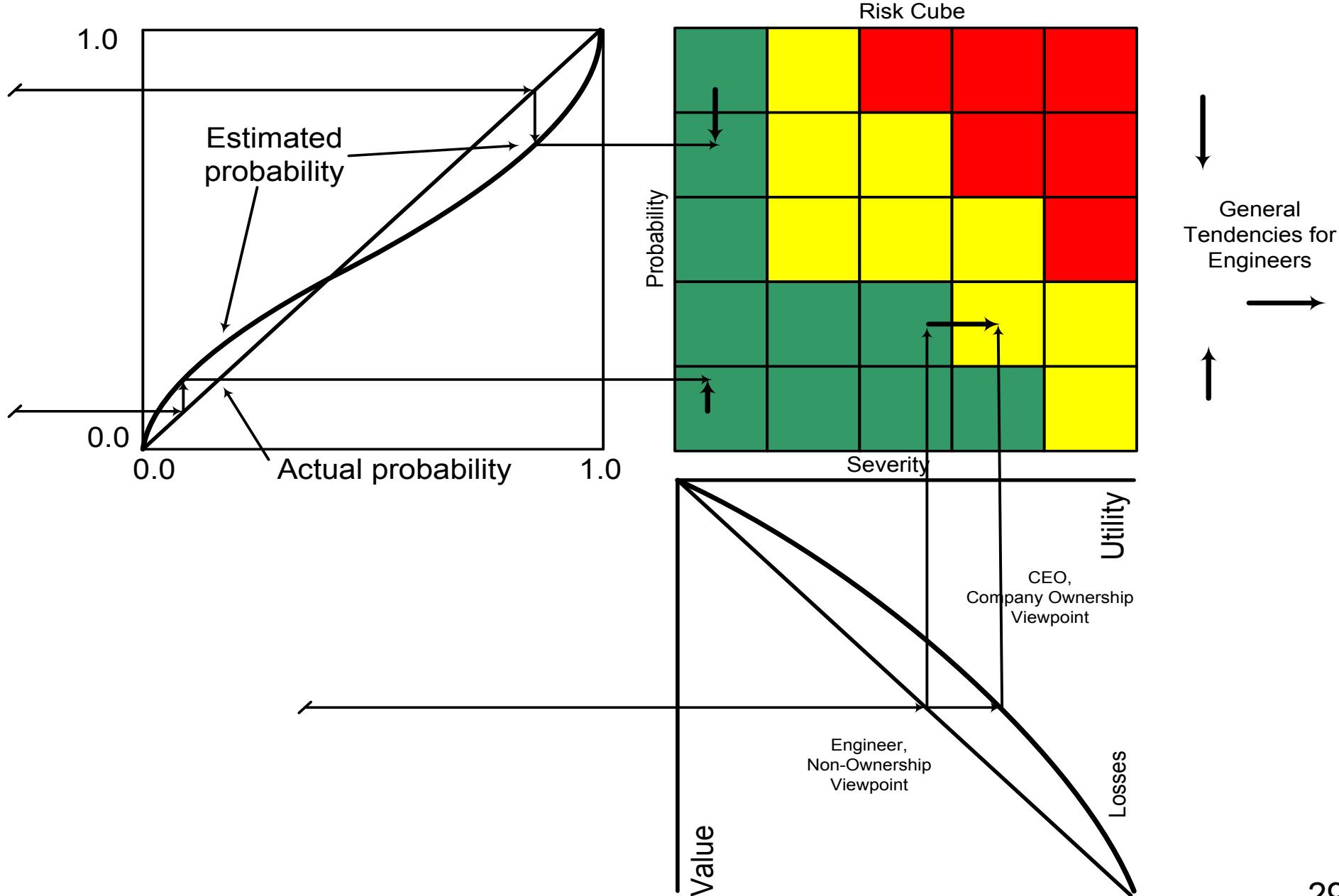
p = objective prob.
 $0 < \delta \leq 1$



When $\delta = 1$, $\pi(p) = p$ = objective probability

usual value for δ :
 $\delta = 0.69$ for losses
 $\delta = 0.61$ for gains

Implication of Prospect Theory for the Risk Matrix



ANALYSES AND OBSERVATIONS OF INITIAL DATA

**Impediments for the appearance of cognitive
biases in the industry data:**

- 1) Our data are somewhat granular while the predictions of Prospect Theory are for continuous data
- 2) Qualitative descriptions of 5 ranges of likelihood and consequence
 - non-linear influence in the placement of risk datum points

**Nevertheless, the evidence of cognitive
biases emerges from the data**



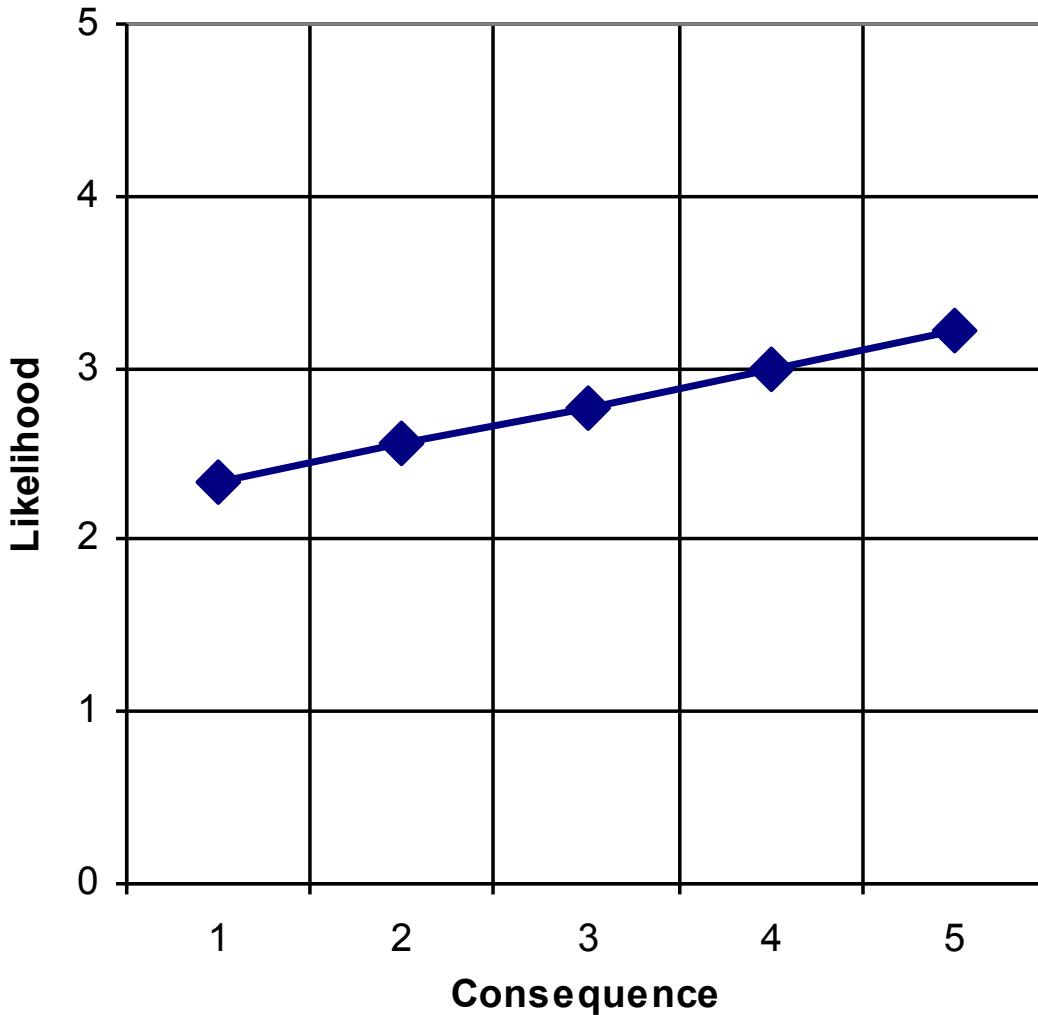
2. Diagonal Bias

- Anticipation of later moving of risk points toward the origin
- Risk points withdrawn from the origin upward and rightward along the diagonal

Regression on 1412 Original Points

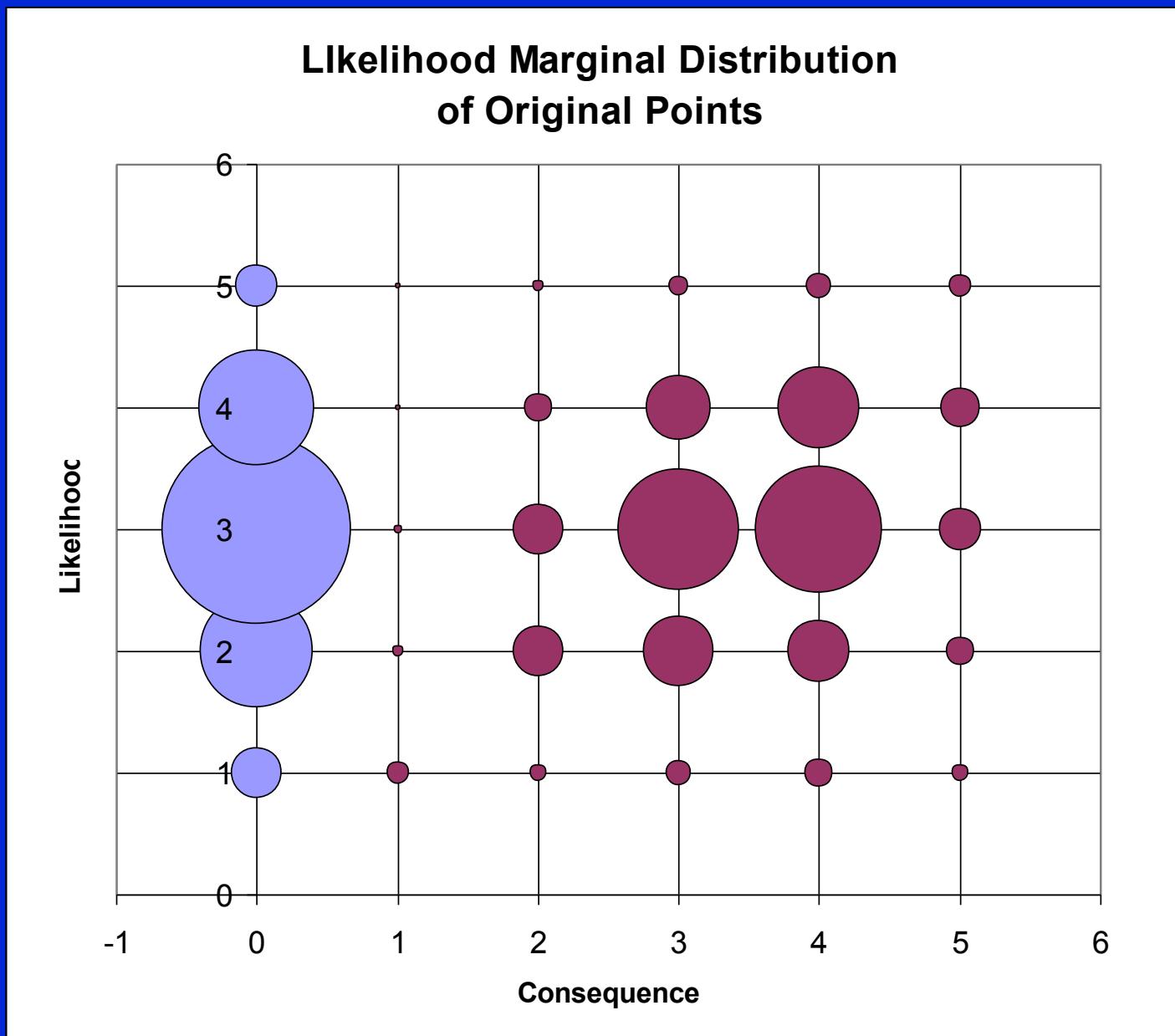
Intercept	Slope	R
2.2	0.22	0.22

Diagonal Bias in Original Points



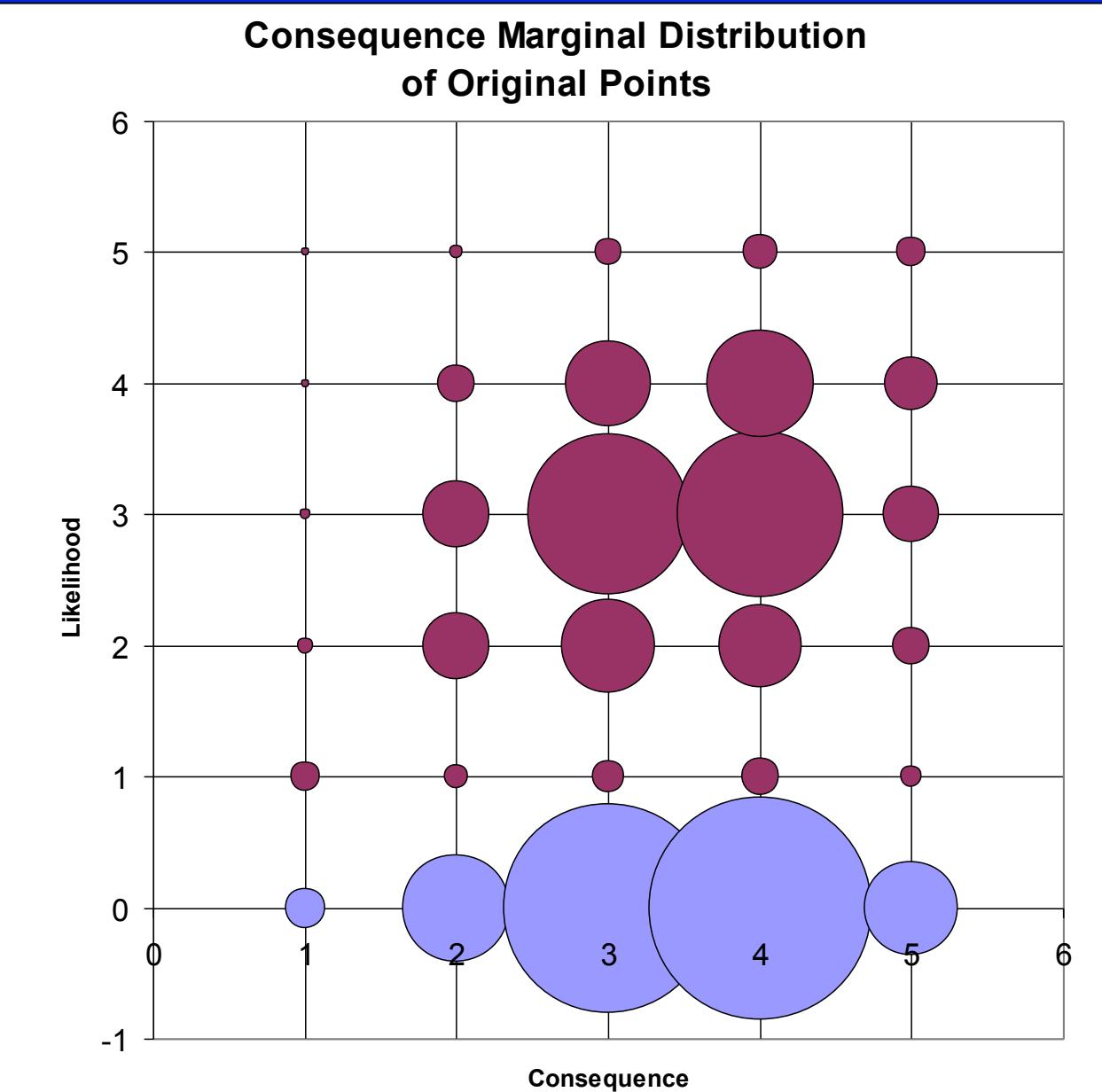
3. Probability Centering Bias

- Likelihoods are pushed toward $L = 3$
- Symmetric to a first order



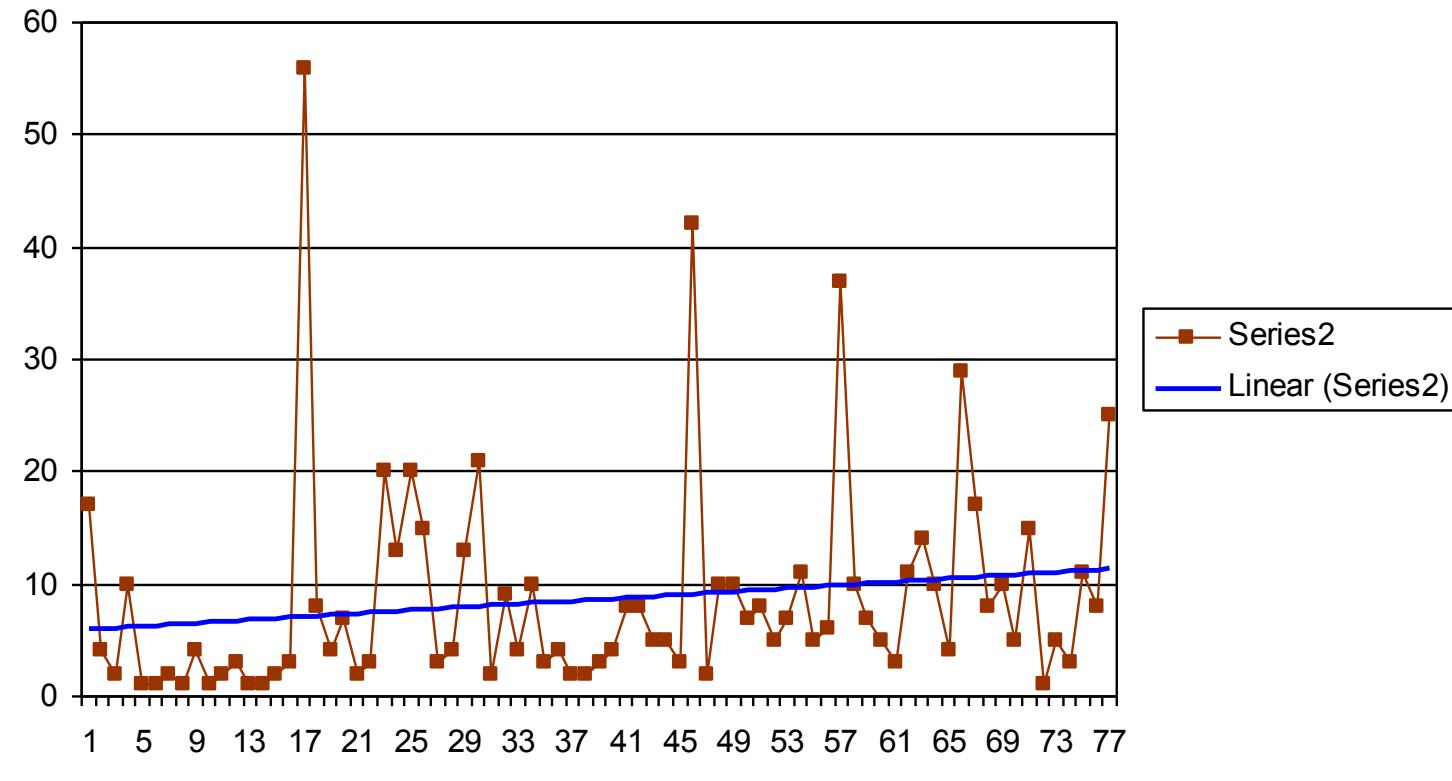
4. Consequence Bias

- Consequence is pushed higher
- Engineer identifies with increased risk to entire corporation
- 'Personal' corporate risk



Open Risks by Month

Units



Time (Months)

Likelihood mitigation recommendations

- **Engineers and Management**
 1. **Technical risk highest priority**
 2. **Schedule risk communicated well by management**
 3. **Cost risk likelihood less frequently communicated by management.**
- **Higher cognizance of cost risk will be valuable at the engineering level**

Likelihood mitigation

1. Technical

2. Schedule

3. Cost

Consequence Mitigation

- **Engineers:**

- 1. Schedule consequences effect careers**
- 2. Technical consequences effect job performance reviews**
- 3. Cost consequences are remote and associated with management**

- **Higher cognizance of cost risk would be valuable at the engineering level**

Consequence mitigation

1. Schedule

2. Technical

3. Cost

CONCLUSION

- First time that the effects of cognitive biases have been documented within the risk matrix
- Clear evidence that probability and value translations, as likelihood and consequence judgments, are present in industry risk matrix data
- Steps
 - 1) the translations were predicted by prospect theory,
2) historical data confirmed predictions
- Risk matrices are not objective number grids
- Subjective, albeit useful, means to verify that risk items have received risk-mitigating attention.

Suggestions for risk management improvement

- **Objective basis of risk:**
 - Frequency data for Probability
 - \$ for Consequence
- **Long-term, institutional rationality**
- **Team approach**
- **Iterations**
- **Expert review**
- **Biases and errors awareness**

SUCCESS

OPPORTUNITY

RISK

IDE / SYS ENG 427-FUNCTION BASED RISK ANALYSIS

Risk is a necessary ingredient for SUCCESS! Effective identification and management of risk and opportunity is a must for the advancement of technology. IDE/Sys Eng 427 presents the risk tools engineers need to thrive in today's ever changing technological environment.



Course Topics

This exciting interdisciplinary engineering course will cover advanced engineering techniques in Probabilistic Risk Analysis (PRA) such as Failure Modes and Effects Analysis (FMEA), Fault Tree Analysis, Event Tree Analysis, etc. will be covered in the context of a range of interdisciplinary products. New state-of-the-art autonomous PRA techniques and risk mitigation plans will also be covered with emphasis placed on early risk identification. These topics will be investigated and im-

plemented over a wide range of products including space shuttles, nuclear power plants, automobiles, computer systems, and household products.



IDE/Sys Eng 427-
Function Based Risk
Assessment

Fall 2008
For more information:
Dr. Katie Grantham Lough
kag@mst.edu



References

- L. Cosmides, and J. Tooby, **Are humans good intuitive statisticians after all? Rethinking some conclusions from the literature on judgment under uncertainty**, *Cognition* 58 (1996), 1-73.
- D. Kahneman, and A. Tversky, **Prospect theory: An analysis of decision under risk**, *Econometrica* 46(2) (1979), 171-185.
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<http://nobelprize.org/economics/laureates/2002/index.html>.
- N. Schwarz, **Assessing frequency reports of mundane behaviors: Contributions of cognitive psychology to questionnaire construction**, *Review of Personality and Social Psychology* 11 (1990), 98-119.
- A. Tversky, and D. Kahneman, **Advances in prospect theory: Cumulative representation of uncertainty**, *Journal of Risk and Uncertainty* 5 (1992), 297-323.

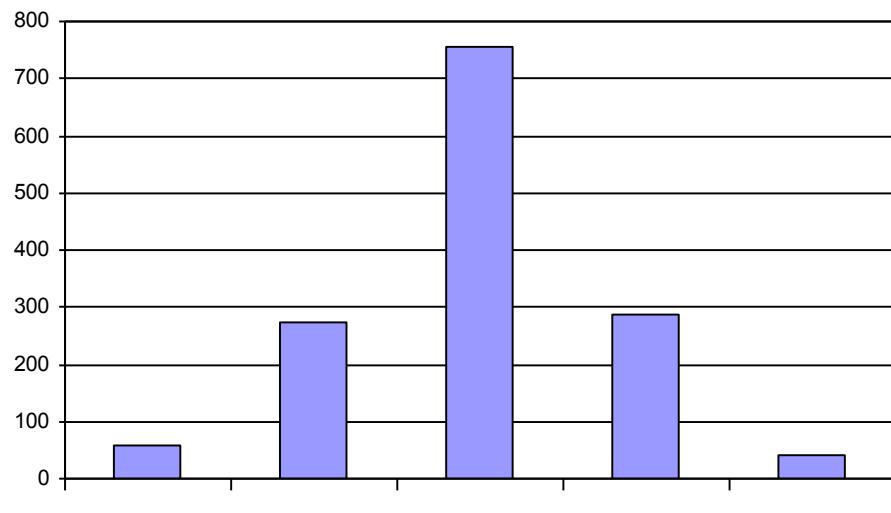
- Comments !
- Questions ?



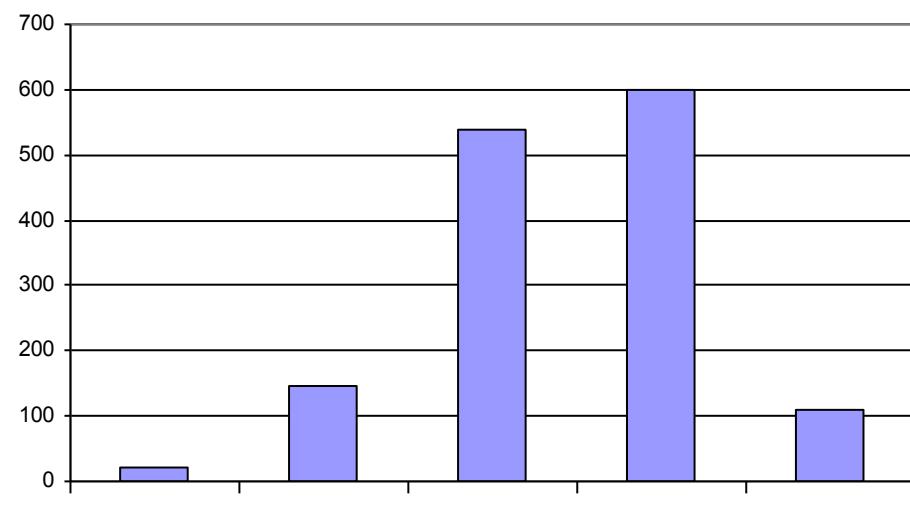
- Back up

Effect of Estimation in a Pre-Defined Scale

Likelihood count in original risks



Consequence count in original risks

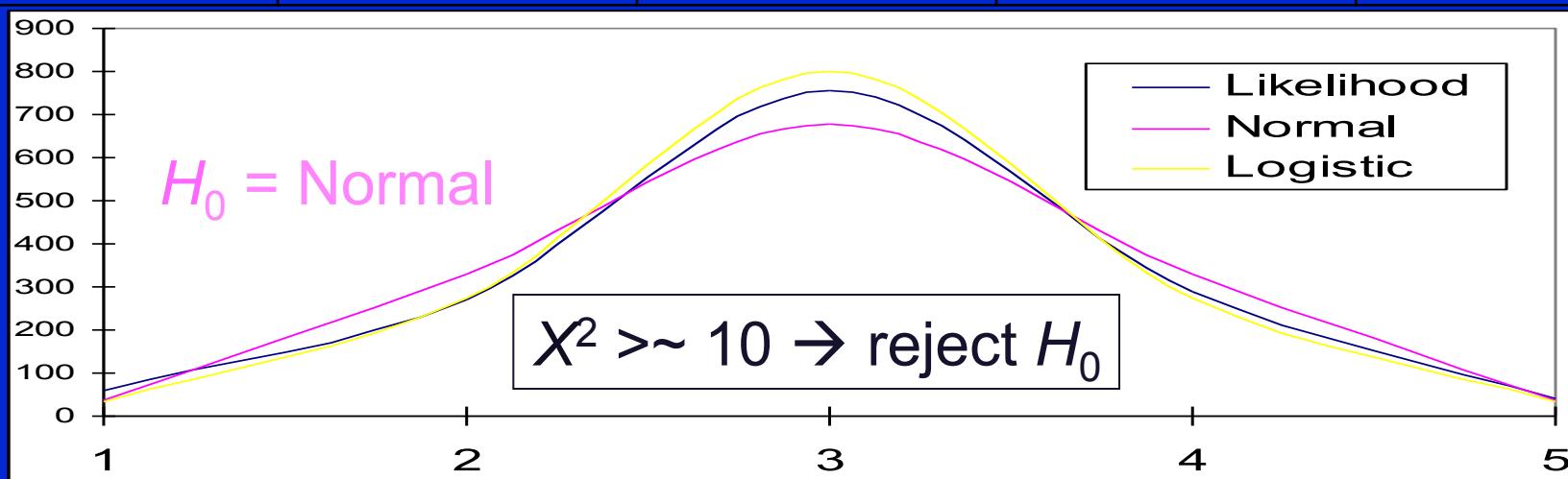


'People estimate probabilities poorly'
[Cosmides and Tooby, 1996]

Consequence/Severity amplifiers

Likelihood Marginal Distribution of Original Points

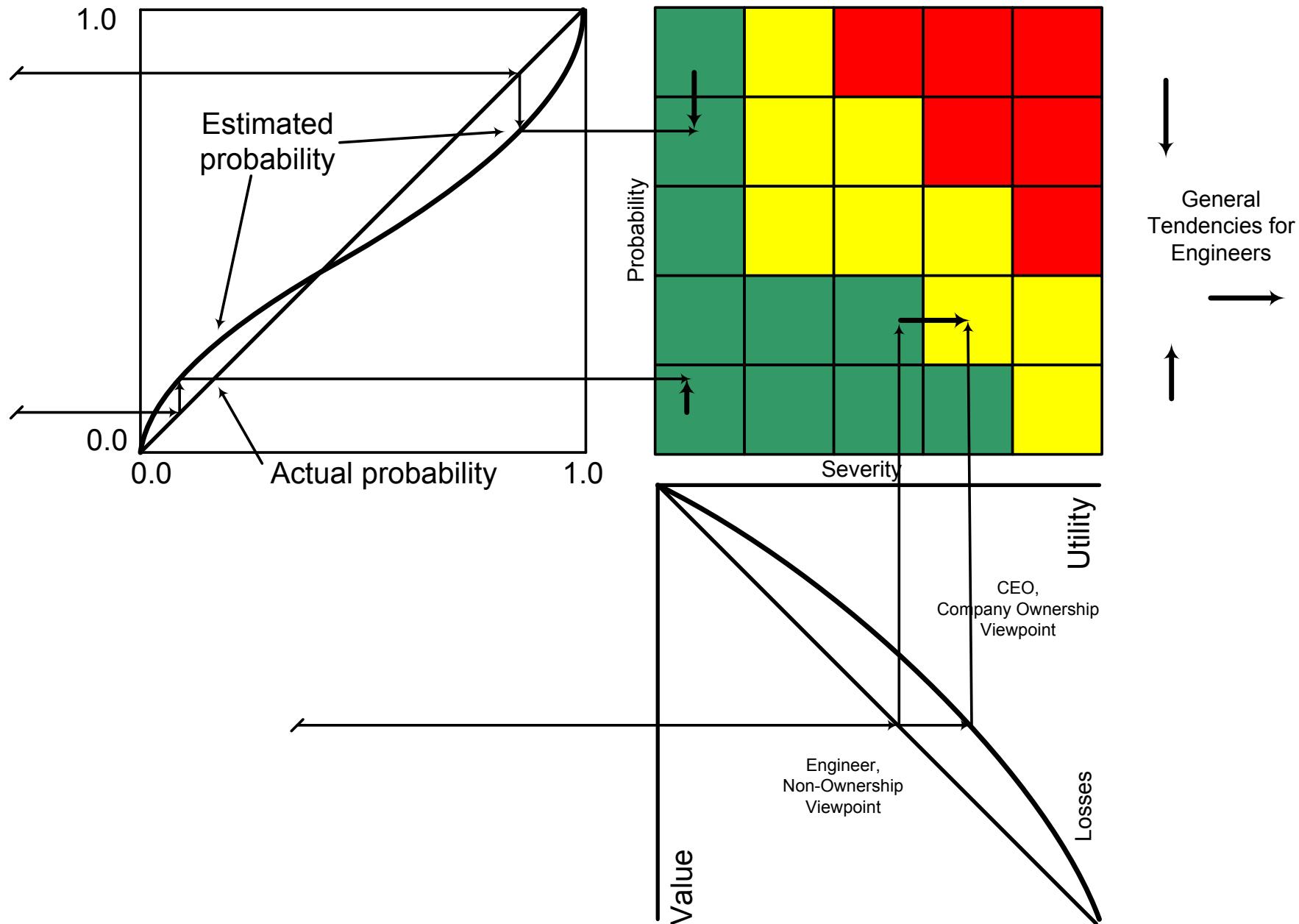
1	2	3	4	5
58	272	754	288	40
Normal distribution with $\mu = 3.0, \sigma = 0.78$				
38	330	676	330	38
$\Delta = \text{actual} - \text{normal}$				
20	- 58	78	- 42	2



$$\chi_N^2 = 36, df = 4$$

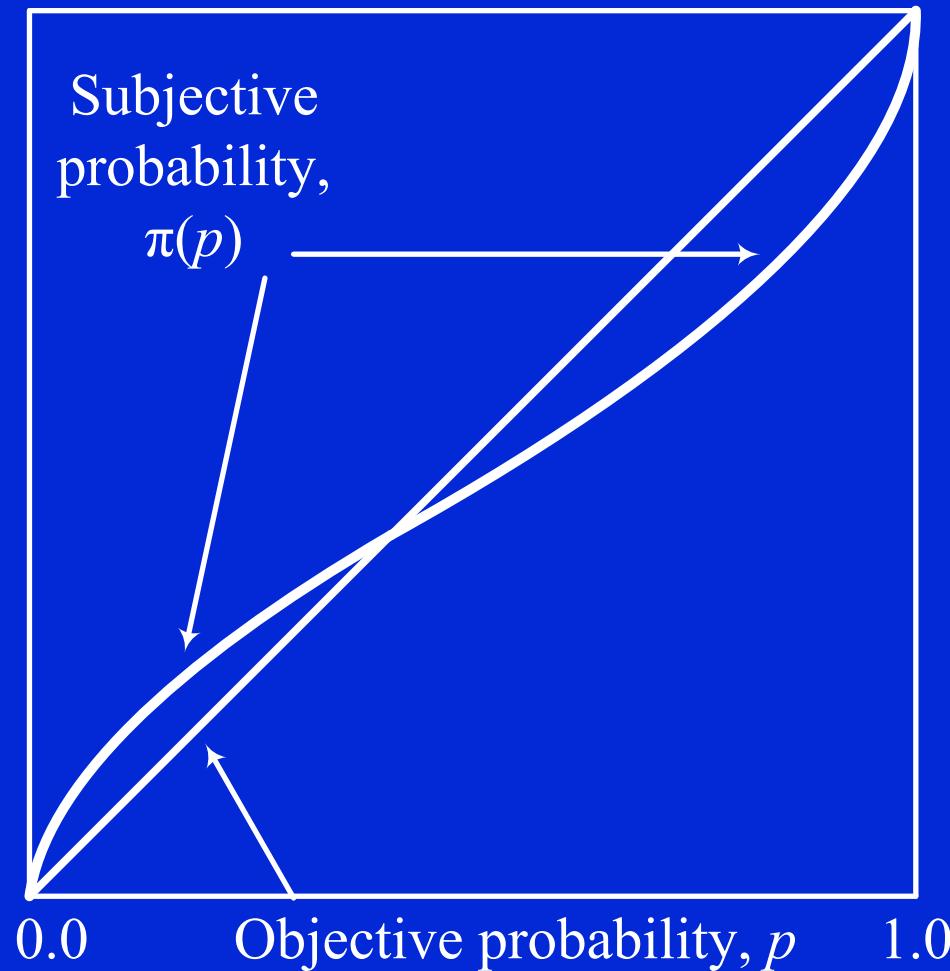
$(\chi^2 = 22, \text{Logistic})$

Implication of Prospect Theory for the Risk Matrix



3a. Asymmetrical Probability Bias

- Subjective probability transformation
- $\pi(p)$ predicts that likelihood data will be pushed toward $L = 3$
 - Large probabilities translated down more than small probabilities are translated up
- Reduced amount of large subjective probabilities, comparatively



1	2	3	4	5
58	272	754	288	40

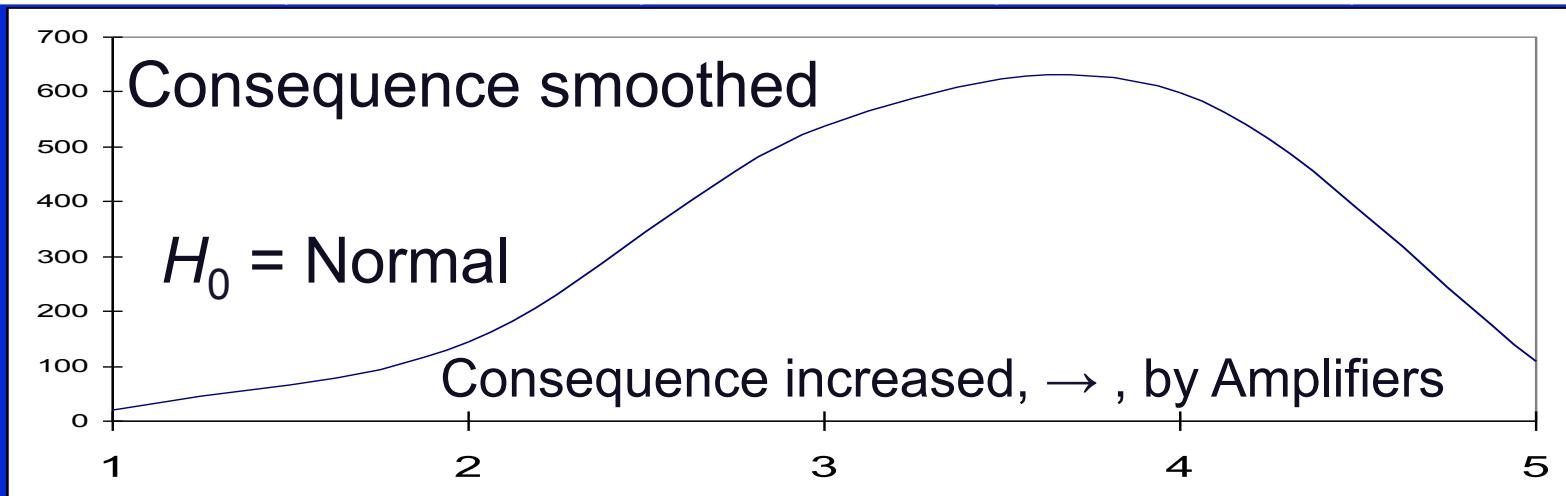
Statistical Evidence for Consequence Bias

Max at C = 4

C = 1 significantly less than C = 5 counts

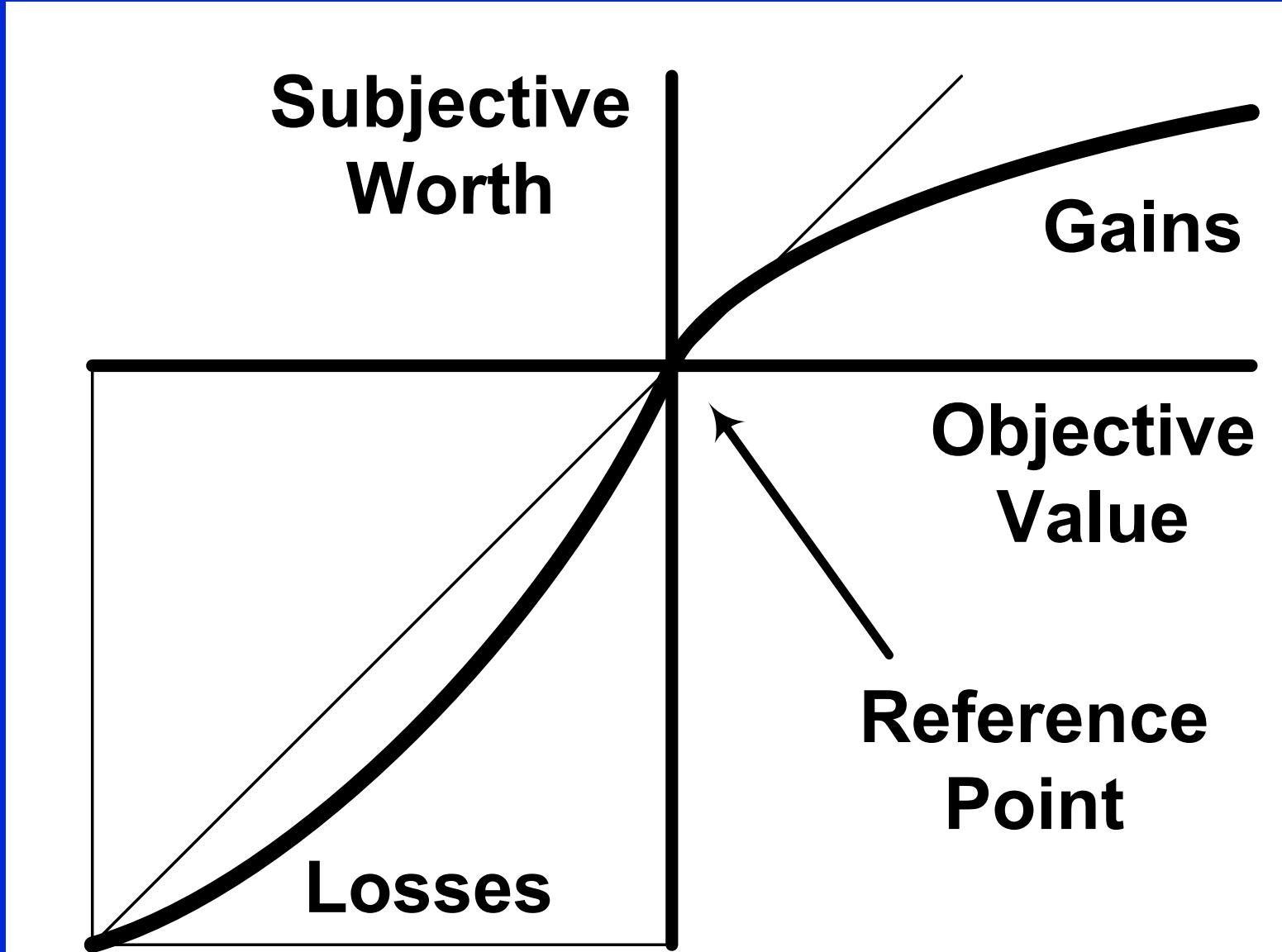
C = 2 significantly less than C = 4

Consequence Original Data Points				
1	2	3	4	5
20	145	538	599	110



Normal distribution comparison: $\chi^2 = 600$, $df = 4 \Rightarrow 0.0$ probability

Consequence translation



Data Collection Improvements

- Continuum of data from
 - Risk management to
 - (Issue management)
 - Opportunity management
- Different databases
 - years of data in each
- Time
 - Waterfall Risk charts

Future work

- Confirmation of the presence of
 - probability biases, and
 - value biases in
 - risk data from other industries or companies
- Real world effects on industry from using biased risk mitigation data
 - \$'s, not risk points
 - Sequential ramifications
 - Prospect Theory risk gambles
- Inform decision makers about how cognitive biases affect risk assessment