

PEOPLE ANALYTICS

*Professor Cade Massey
The Wharton School*



•Rigorously tracks the performance of all teachers, comparing it to evaluations when they were hired.

•Helps refine the most productive steps in the hiring process, where to allocate more resources, etc.

TEACH FOR AMERICA



CORNER OFFICE: LASZLO BOCK
By ADAM BRYANT
Published: June 19, 2013

In Head-Hunting, Big Data

This interview with Laszlo Bock, senior operations at Google, was conducted by Adam Bryant.

[Enlarge This Image](#)



Q. How did you get into the head-hunting field?
A. I think it's because of confluence of data and capital in business what people perennially have a problem, measure Part of the

- Systematically tracked interview predictions about new hires to figure out how good they were at it.
- Answer: Not very.
- So dramatically reduced the # of interviewers.

Google



- Believes a 1% increase in retention can save \$75-100m/year.
- 3-year study: Changing jobs increases employee “stickiness”.
- Increased internal postings of open jobs from <50% to >80%.

CREDIT SUISSE 

Many, many others

- Firms in technology, financial services, telecommunications, automotive, consumer packaged goods, energy, not-for-profit...
- ...are finding:
 - Better levers for retaining key employees
 - More diagnostic methods for hiring
 - Who their most valuable employees are
 - How to compose the most productive teams
 - Etc.

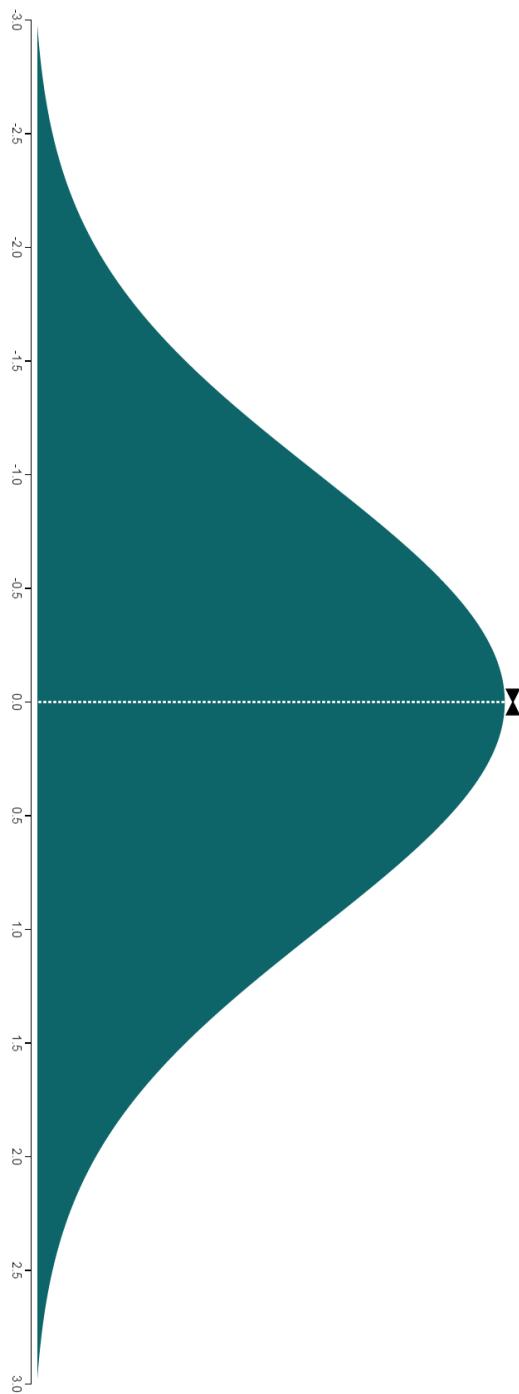
Part 1A – Performance Evaluation

- Purpose of performance evaluation
 - Feedback
 - Rewards / punishment
- Performance valuation, not talent management. Tough to compare employees if not in identical situations. More on this in a later module.
- Helpful starting place, for this lecture (and often for life):
Begin by assuming all employees are equal ability.

Noise

- The fundamental challenge in performance evaluation is that performance measures are *noisy*.
- I.e., outcomes are imperfectly related to employee effort.

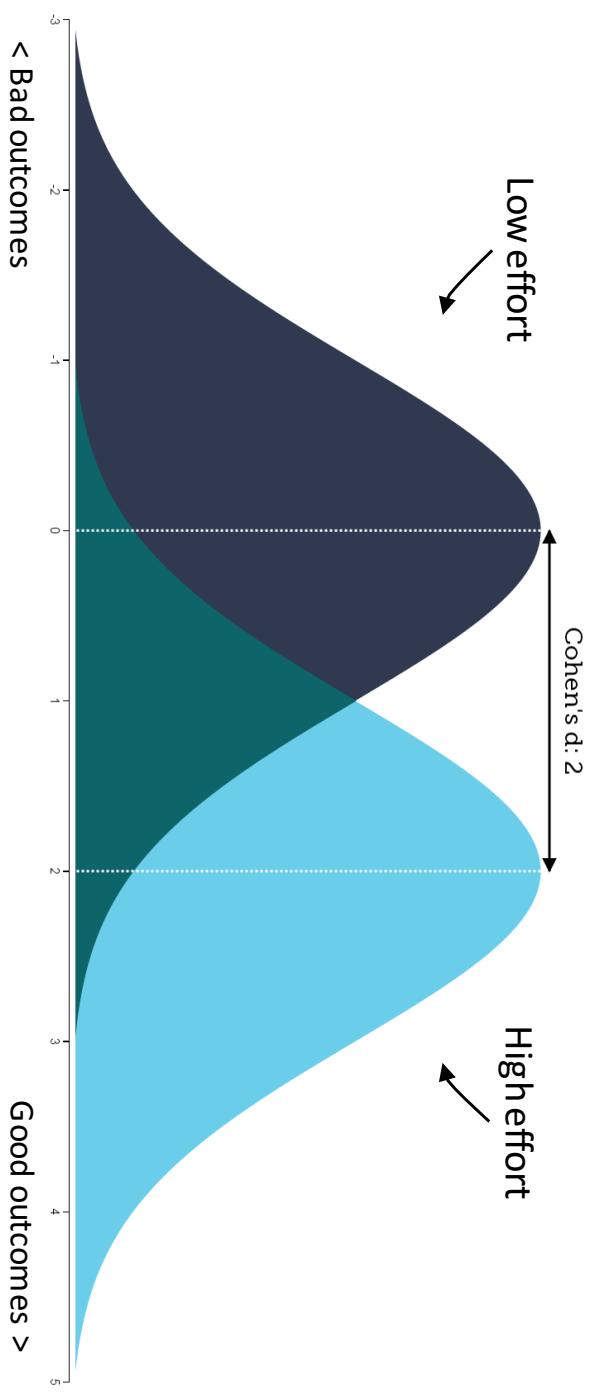
- For any given level of effort, a range of outcomes can occur due to factors outside the employee's control.
 - Competitors, team members, her boss, the economy, etc.
 - The challenge: Separating skill from luck.



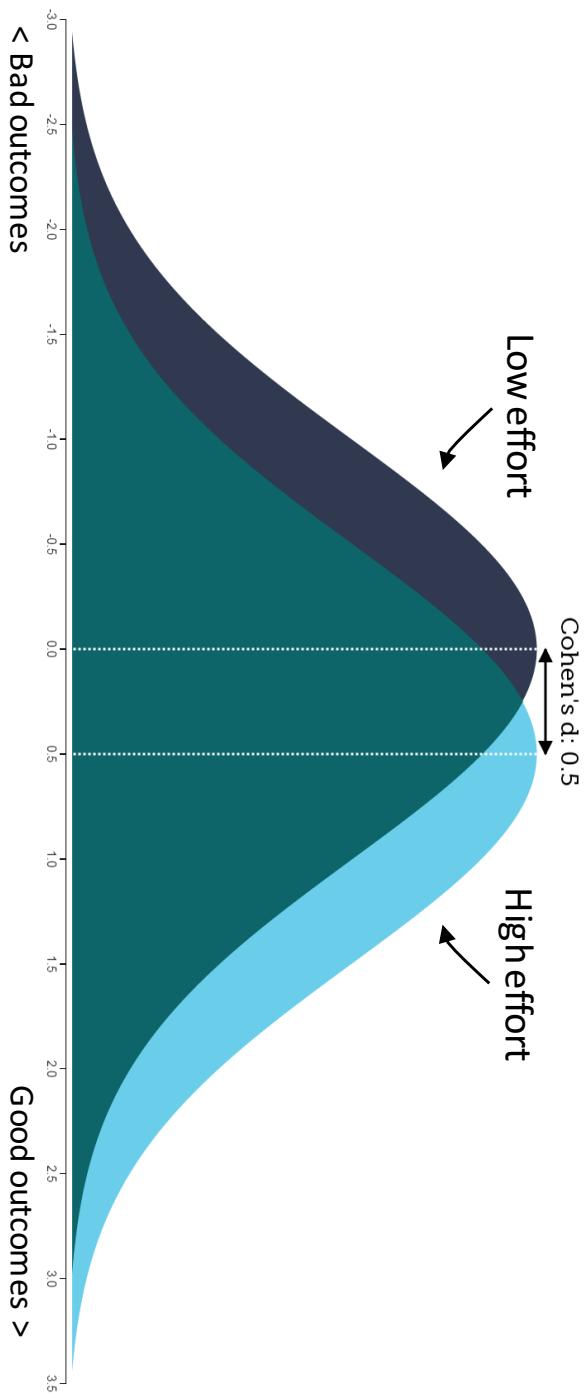
- University of Chicago economist John Huizinga:

“This distinction between ascertaining skill and luck shows up all the time,” he added. “Who do you give your money to for investing? How much do you pay a certain employee? It’s everywhere. It isn’t just about sports. It’s about life.”

- In some environments this isn't so hard...



- ...but in others it's brutal.



Module Plan

- Extended example
- Four key issues:
 - Regression to the mean
 - Sample size
 - Signal independence
 - Process vs. Outcome
- Summary

Module Plan

- **Extended example**

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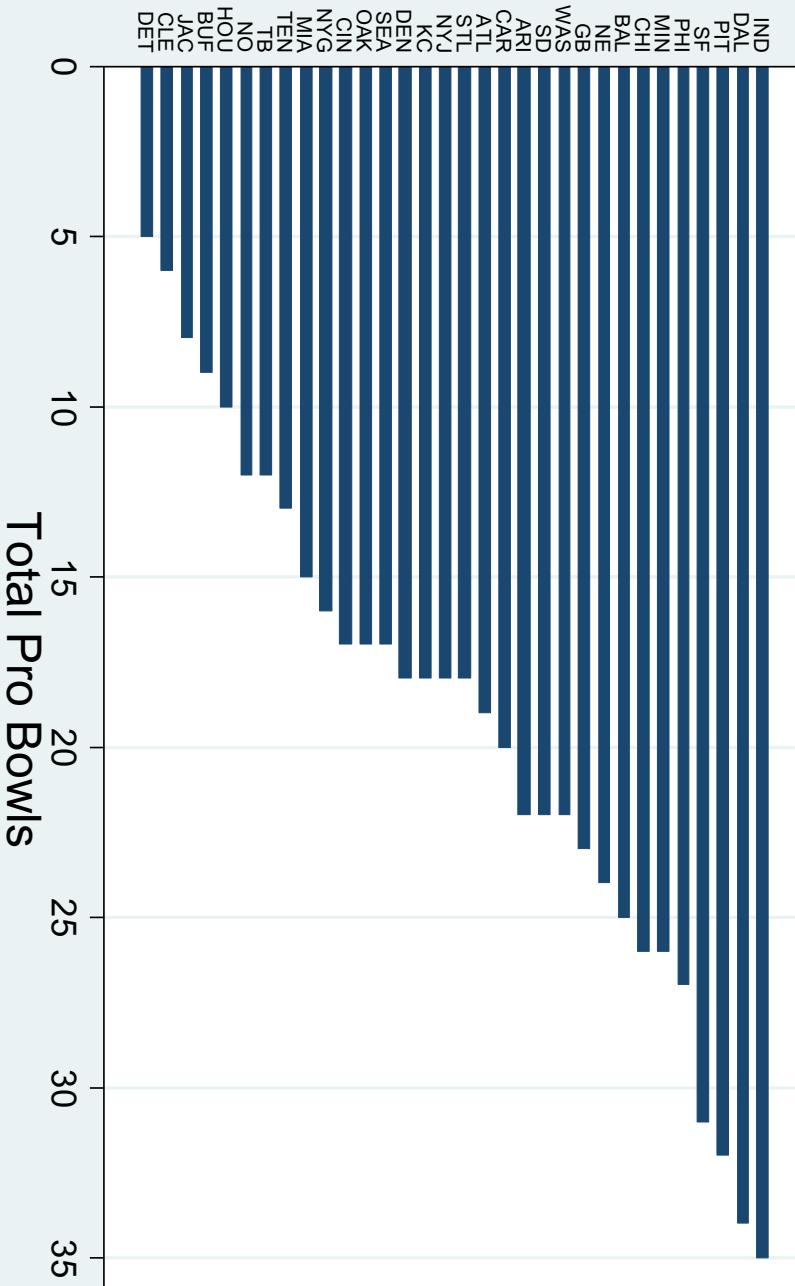
The NFL Draft



The NFL Draft

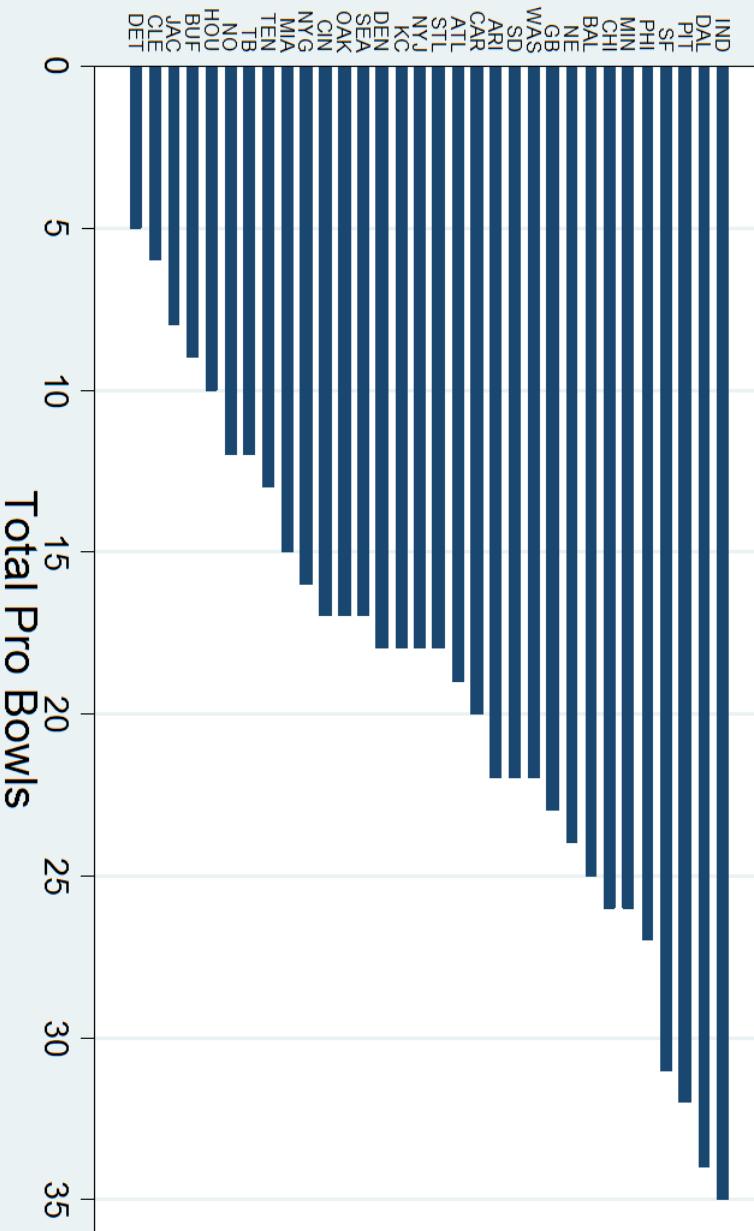
- Teams take turns drafting college players.
 - One of their main sources of talent
 - Players are obligated to play for the team that drafts them for ~4 years.
- The draft is comprised of 7 “rounds”
 - A round consists of each team picking once
 - They pick in reverse order of the previous year’s records
 - Currently 32 teams, so ~224 players drafted
- Picks can be traded, and are

Pro Bowls by DRAFTING TEAM



1997-2007 drafts, through the 2009 season.

Pro Bowls by DRAFTING TEAM



1997-2007 drafts, through the 2009 season.

Updated: April 21, 2010, 4:44 PM ET

Not feeling the NFL draft



By Rick Reilly
ESPN.com
Archive

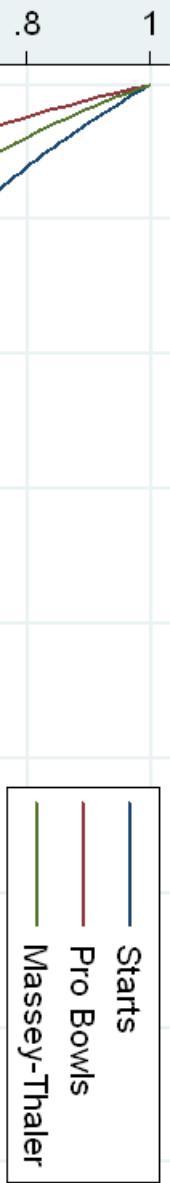
"Indianapolis Colts...not just with Peyton Manning... Dwight Freeney, Edgerrin James and Reggie Wayne were genius picks, too."

"The Cleveland Browns...screwed the Chihuahua. Their run of No. 1 picks from 1999 to 2002 is the single worst stretch of drafting since the Iraqi Republican Guard. Were they using an Ouija board?"

Reason for Skepticism re Skill

- Massey & Thaler (2013)
 - Overconfidence in the NFL draft
- Baron & Hershey (1988)
 - Outcome bias
- Rabin (2002)
 - Law of Small Numbers
- Less difference in skill than people believe
 - This fictitious variation is “the most important economic consequence of the law of small numbers”

Performance of Players Drafted 1991-2004

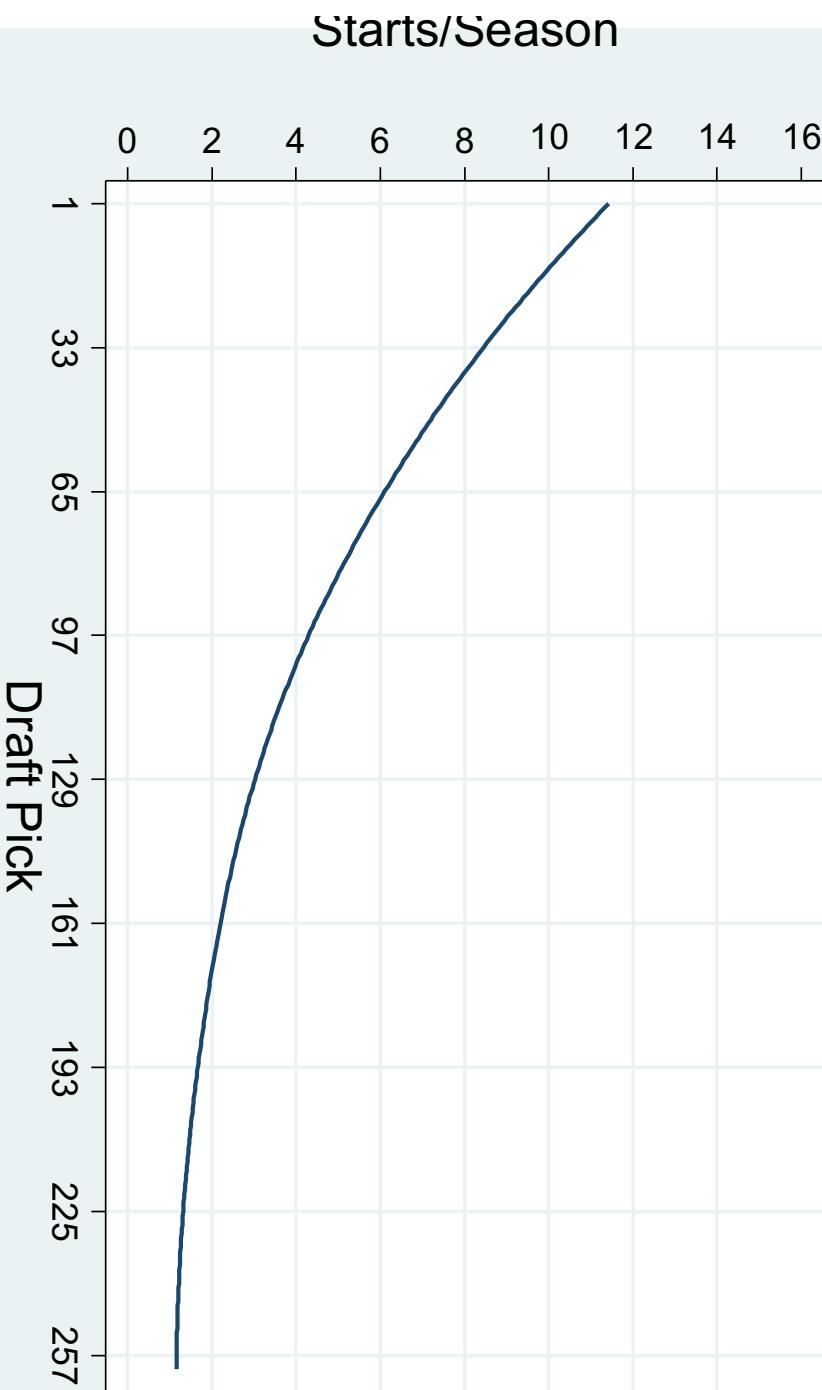


Lowess curves. Player's 1st 5 years.

Skill and Chance in the Draft

- Clearly there is skill involved.
- But are there differences in skill?
- Are some teams better than others at picking players?

Draft-pick Performance *vs.* Expectation

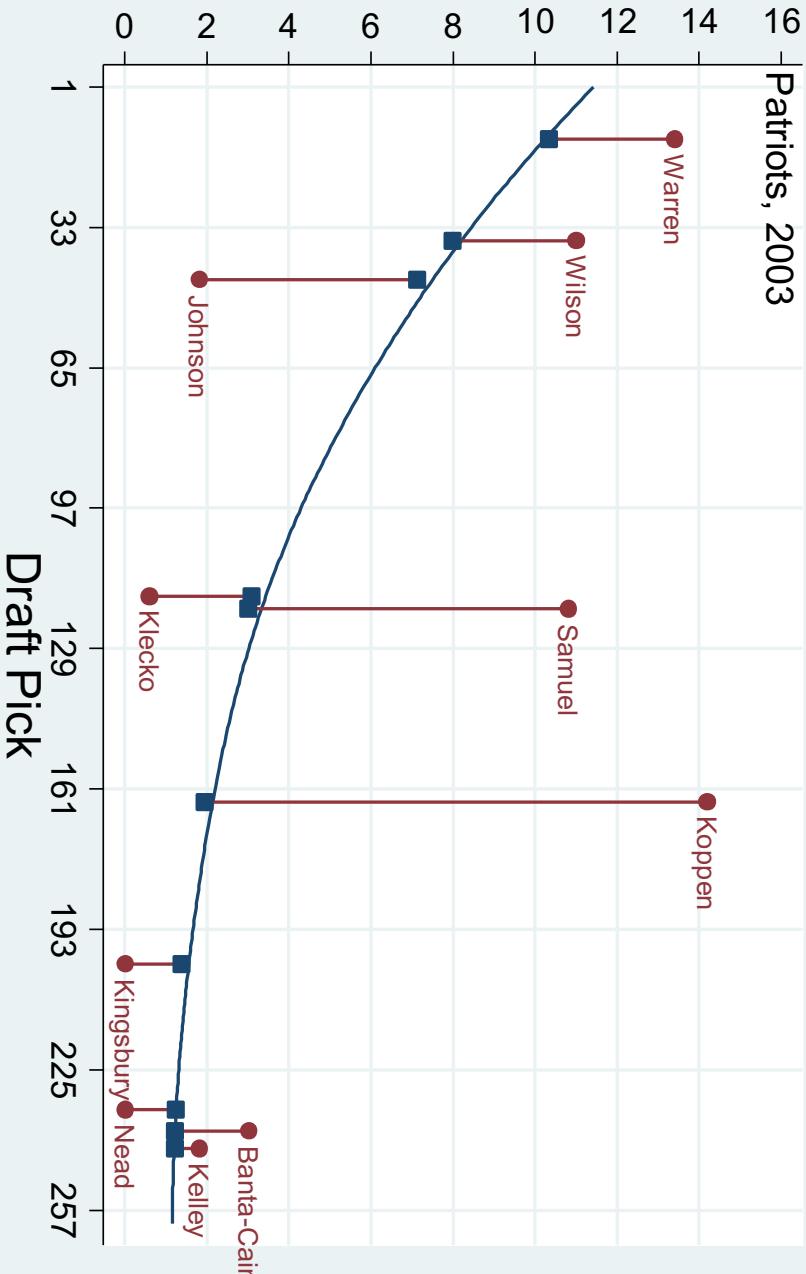


Performance evaluated over player's 1st 5 years.

Draft-pick Performance vs. Expectation

Patriots, 2003

Starts/Season



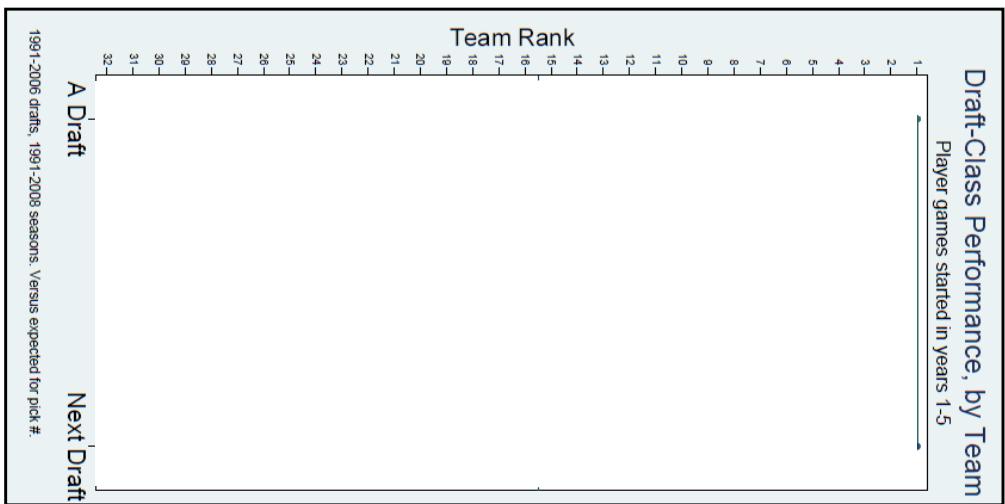
Performance evaluated over player's 1st 5 years.

Skill Test: Does Good Performance Persist?

- E.g., Pittsburgh Steelers famous draft class of 1974
 - 4 Hall of Fame players: Jack Lambert, Lynn Swann, John Stallworth, Mike Webster
- How did they do in the 1973 draft?
 - 2nd-round pick never played a game
 - 3rd & 4th round picks were average at best
- How about 1975?
 - Not one game started by 21 picks
 - Picks in each of the top 6 rounds, who only played in a total of 24 games for the team

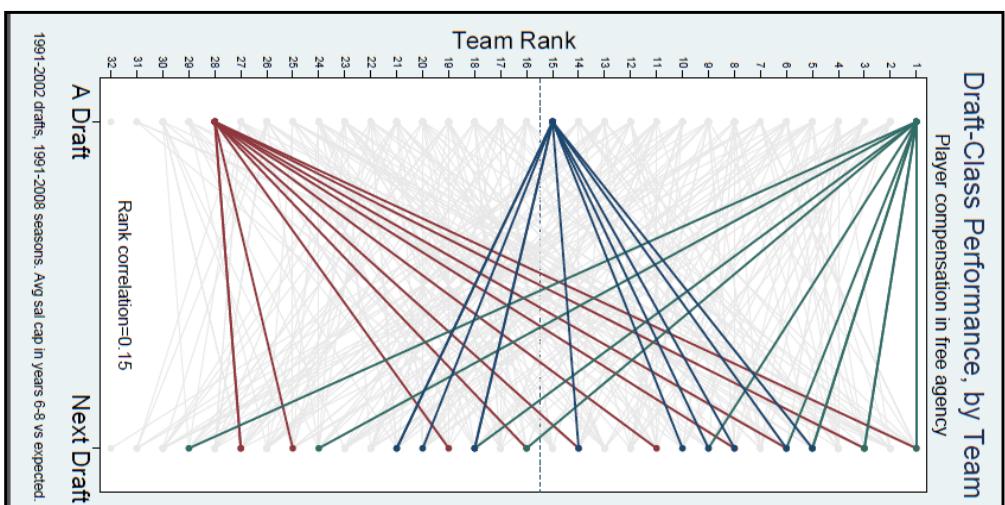
Draft-Class Performance, by Team

Player games started in years 1-5



Draft-Class Performance, by Team

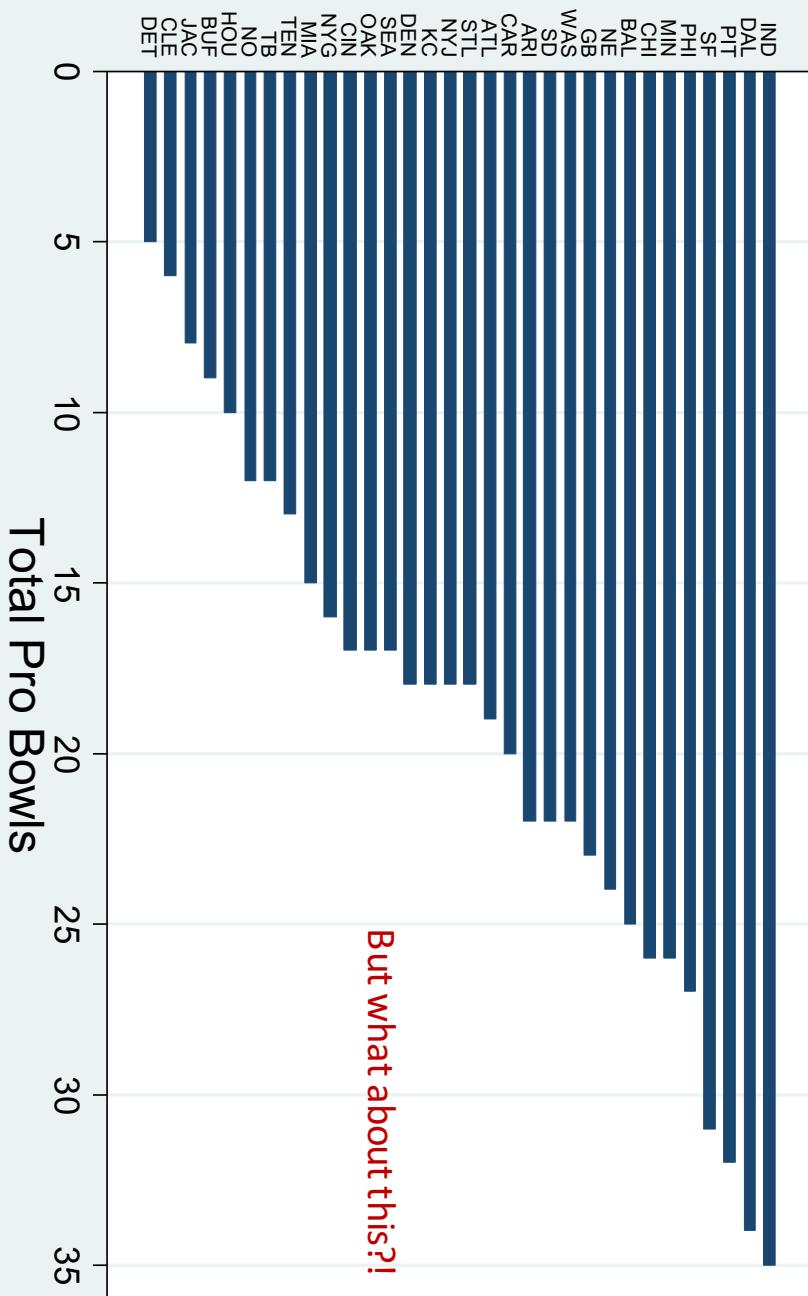
Player compensation in free agency



This is a very robust pattern

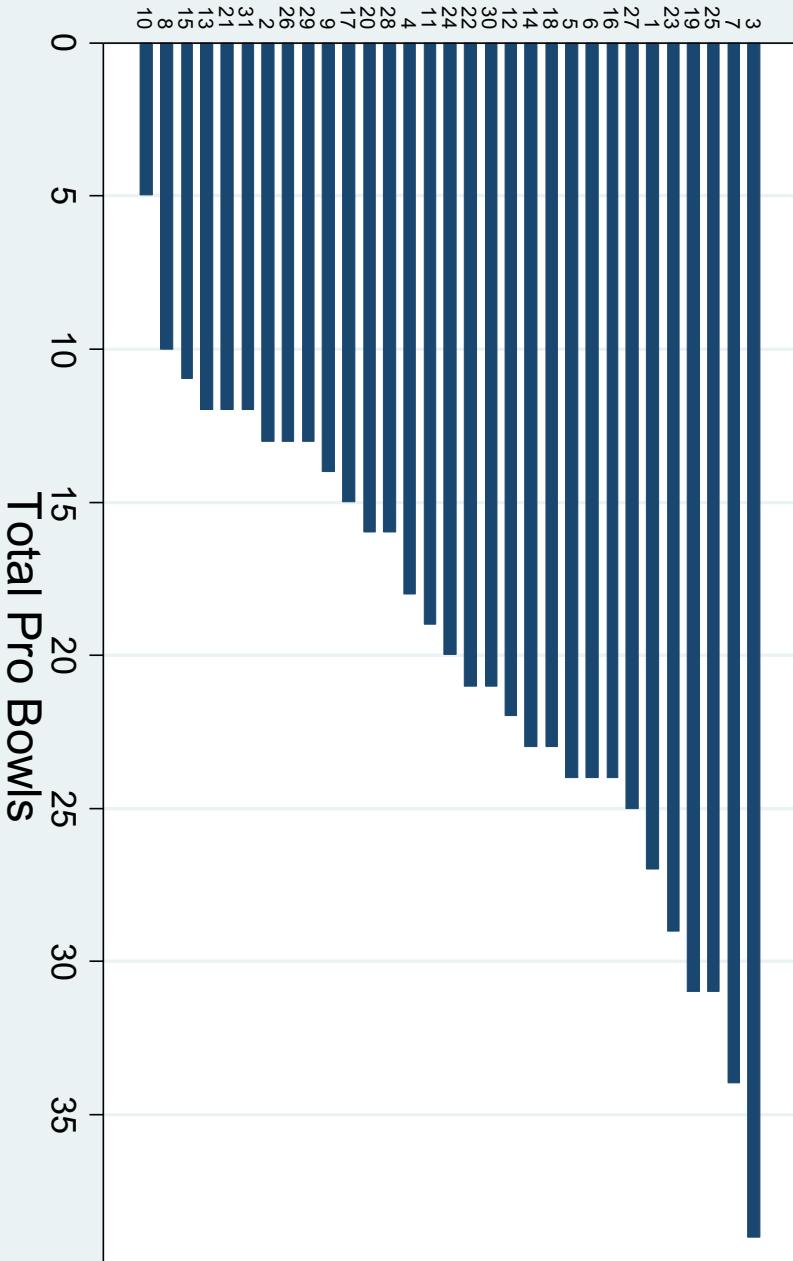
- Performance Statistic
 - Games started vs. pro bowls vs. compensation
- Player's career stage
 - 1st 5 years vs. free-agent years (>5th)
- Additional norming
 - Player position
- Decision-making unit
 - Actual individual in charge of a team's draft
- Draft stage
 - 1st 3 rounds vs. Last 4 rounds

Pro Bowls by DRAFTING TEAM



1997-2007 drafts, through the 2009 season.

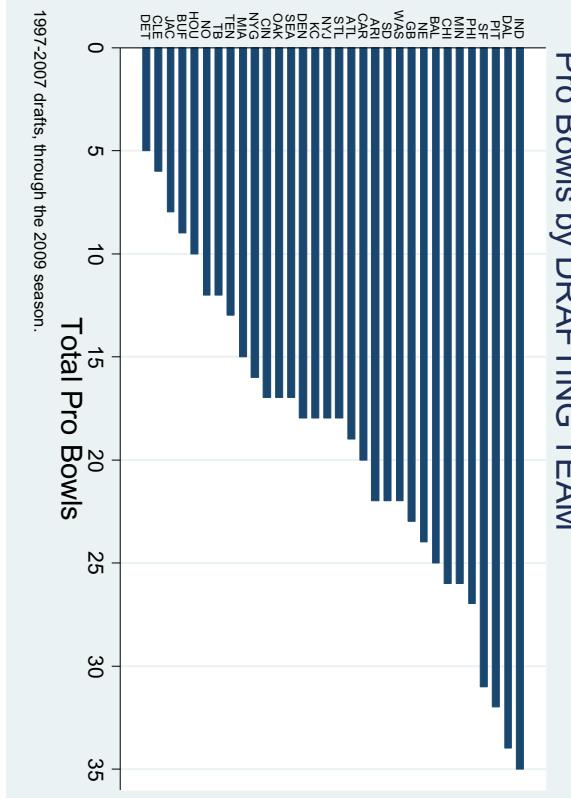
Pro Bowls by PLAYER BIRTHDAY (day of month)



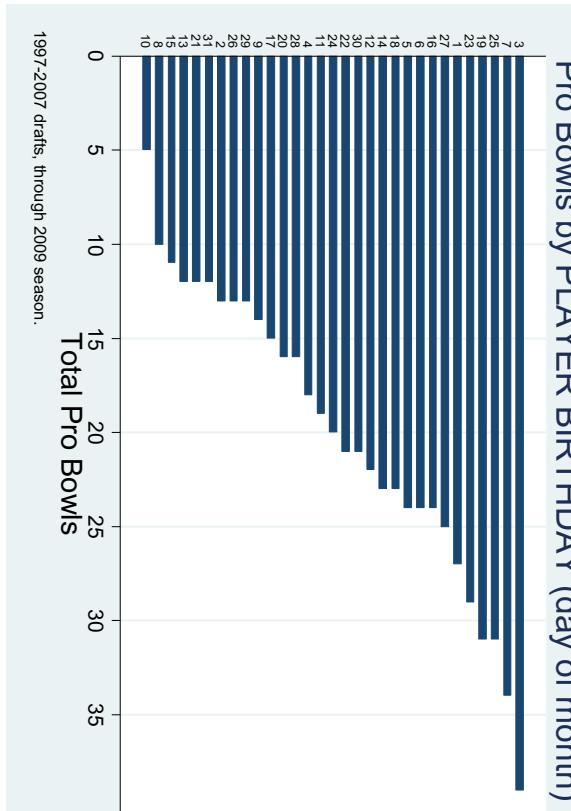
1997-2007 drafts, through 2009 season.

Fictitious Variation

Pro Bowls by DRAFTING TEAM



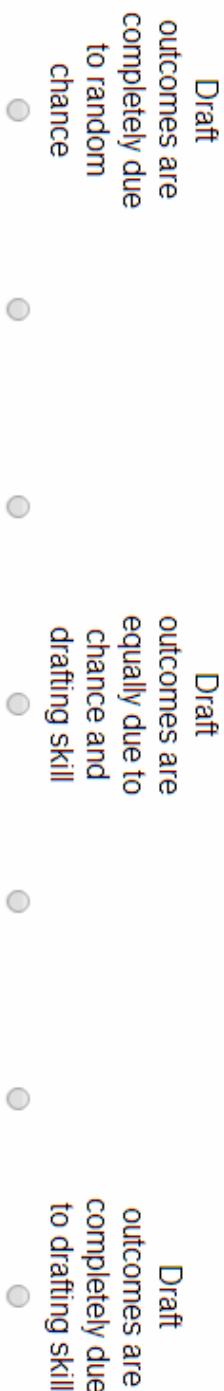
Pro Bowls by PLAYER BIRTHDAY (day of month)



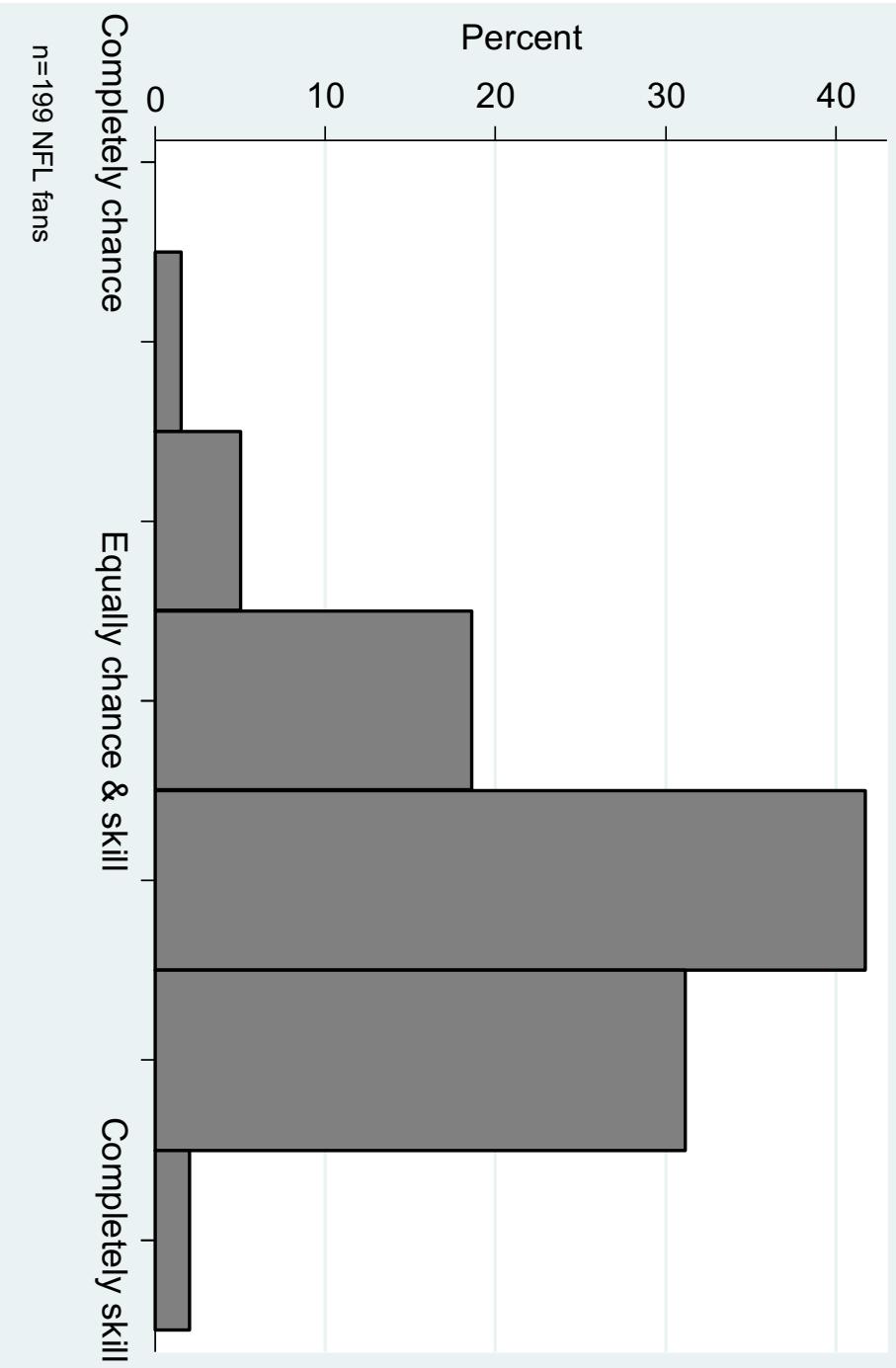
Even dramatically disparate outcomes can be purely the product of chance.

Perceptions of Chance (with Berkeley Dietvorst)

Please indicate to what degree you believe the success of NFL teams in making selections during the NFL draft is due to the drafting skill of each team versus random chance. Indicate to what degree differences between teams in terms of draft outcomes -- e.g., how well players perform long-term -- are due to attributes of the teams' management versus luck.



Are NFL Draft Outcomes Luck or Skill?



Michael Lewis, on why Moneyball versus “more serious” topics



- “If professional baseball players, whose achievements are endlessly watched, discussed and analyzed by tens of millions of people, can be radically mis-valued, who can't be?”
- “If such a putatively meritocratic culture as professional baseball can be so sloppy and inefficient, what can't be?”

Module Plan

- Extended example
- Four key issues:
 - Regression to the mean
 - Sample size
 - Signal independence
 - Process vs. Outcome
- Summary

Issue #1: Regression to the Mean

A Simple Model

- There are two components to performance:
 - In **informal terms**: Real Tendency + Luck
 - In **more formal terms**: $y = x + e$,
 - x = true ability, and
 - e = error, randomly distributed around 0.
- What happens when we sample on extreme performance?
What underlies extreme success and failure?
 - Extreme success = f(superior ability, positive error)
 - Extreme failure = f(inferior ability, negative error)
- Consequences?

- A study was recently conducted examining the performance of the 283 stock mutual funds that existed during the 1990s. The study divided the 1990s into an early period (1990-1994) and a late period (1995-1999). Below are the 10 funds that had the highest rate of return in the early period (with their names disguised), ranked from 1 to 10. Predict their rank for the late 1990s.

Ignoring Regression-to-the-Mean (2. Var. Neglect)

Early 1990s Fund & Rank	Late 1990s Estimated Rank (median)	Late 1990s Actual Rank
A 1	10	129
B 2	20	134
C 3	20	261
D 4	28	21
E 5	44	210
F 6	37	53
G 7	42	183
H 8	31	105
I 9	31	275
J 10	25	54

Total # funds = 283 Avg. = 25, r=.51

Avg. = 142.5, r = -.03

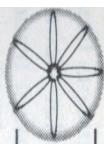
Examples

- Officer in the Israeli Air Force— “Punishment is more effective than praise. Whenever I punish a pilot after a really poor flight, I see better performance the next time. Whenever I praise a pilot after an excellent flight, I see worse performance the next time.”

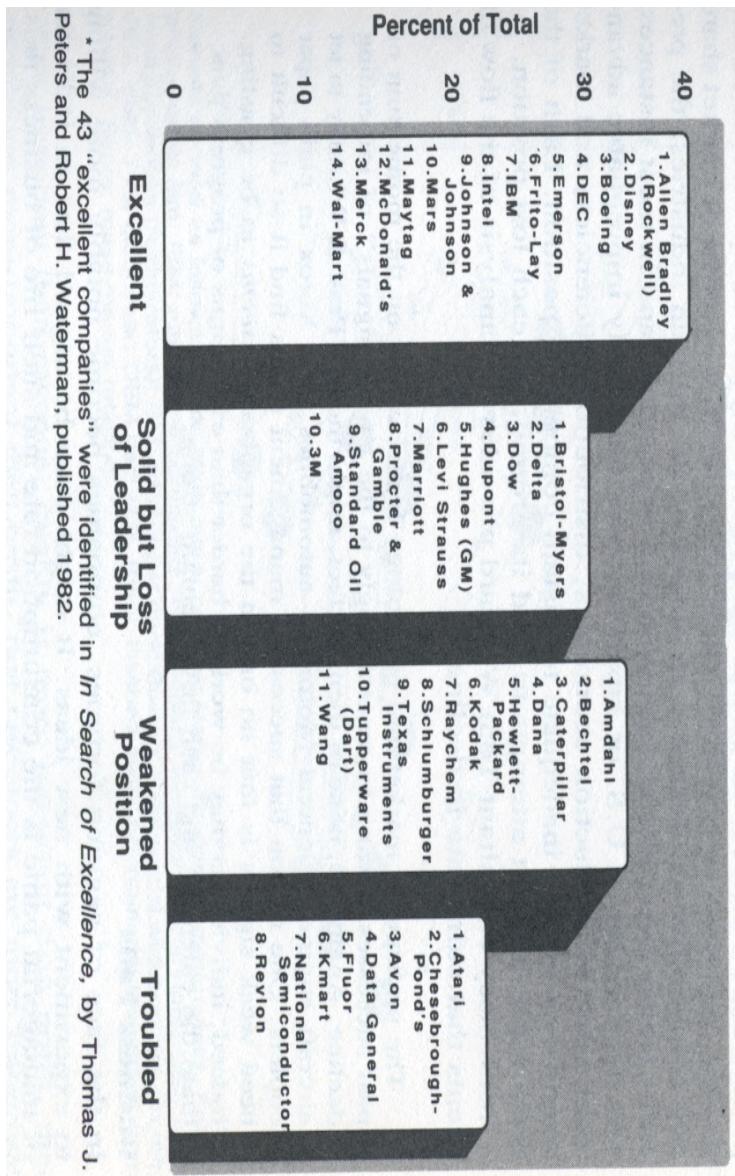
- Peters and Waterman’s book, In Search of Excellence. They selected 43 high performing companies in the early 1980s, and looked to see what practices they used (some that they discovered were the organizational equivalent of “brushing teeth”)

Excellent Companies. 5 Years Later

FIG. 1-1



Status of the 43 "Excellent"
Companies Five Years Later*



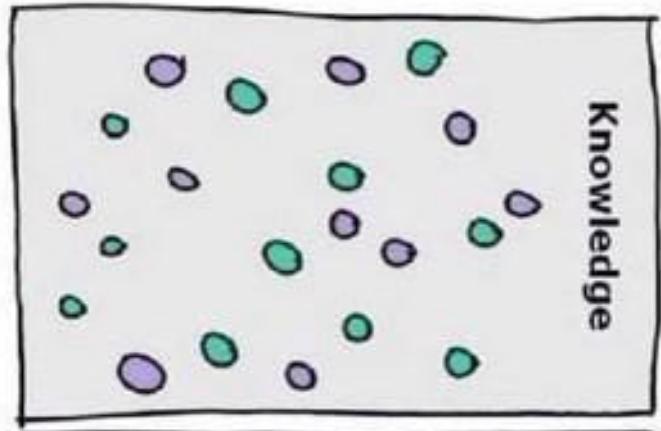
* The 43 "excellent companies" were identified in *In Search of Excellence*, by Thomas J. Peters and Robert H. Waterman, published 1982.

Regression to the Mean

- Anytime you sample based on extreme values of one attribute, any other attribute that is not perfectly related will tend to be closer to the mean value.
- “Attributes” can be:
 - Performance at different points in time
 - E.g., last year’s stock returns and this year’s
 - Different qualities within the same entity
 - E.g., a person’s running speed and language ability

What gets in the way of seeing this?

- Among other things:
 - Outcome bias
 - Hindsight bias
 - Narrative seeking
- In short, we make sense of the past
 - We find a story that connects all the dots
 - Chance plays too small a role in these stories



Issue #2: Sample size

Extrapolating from Small Samples

• Your firm has two plants, one large and one small, which mass produce a standard computer chip. Other than the amount they produce, the two plants are identical in all essential regards. Both use the same technology to produce the same product. When properly functioning, this particular technology produces one percent (1%) defective items. Whenever the number of defective items from one day's production exceeds two percent (2%), a special note is made in the quality control log to "flag" the problem. At the end of the quarter, which plant would you expect to have more "flagged" days in its quality control log? Please mark one.

•

- A) The small plant
22%
- B) The large plant
30%
- C) The same number on average
48%

Extrapolating from Small Samples

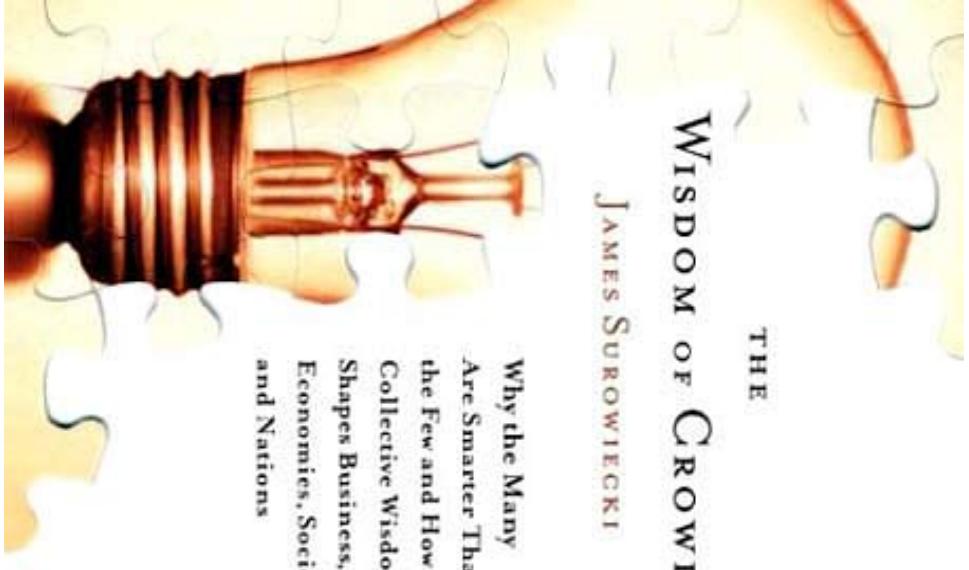
Principle: Sample means converge to the population mean as the sample size increases. (This is known as the Central Limit Theorem.) Thus, you will see more extreme values in small samples.

- When are you more likely to see a .400 season batting average in baseball – May 1 or Sept. 1?
- In which hospital are you more likely to see a dramatically higher % of boys than girls (or vice versa) born on any given day – a small community hospital (e.g., 5 births/day) or a large city hospital (e.g., 100 births/day)?

“Law of Small Numbers”

- People believe small samples closely share the properties of the underlying population.
- This means they too readily infer the population's properties (e.g., average) from the sample's.
- That is: They neglect the role variability (aka chance) inevitably plays in small samples.

Issue #3: Signal Independence



THE WISDOM OF CROWDS

JAMES SUROWIECKI

Why the Many
Are Smarter Than
the Few and How
Collective Wisdom
Shapes Business,
Economies, Societies
and Nations

The average of a large number of forecasts reliably outperforms the average individual forecast.

- Idiosyncratic errors offset each other
 - E.g., Galton's (1906) county fair contest
 - Many other examples

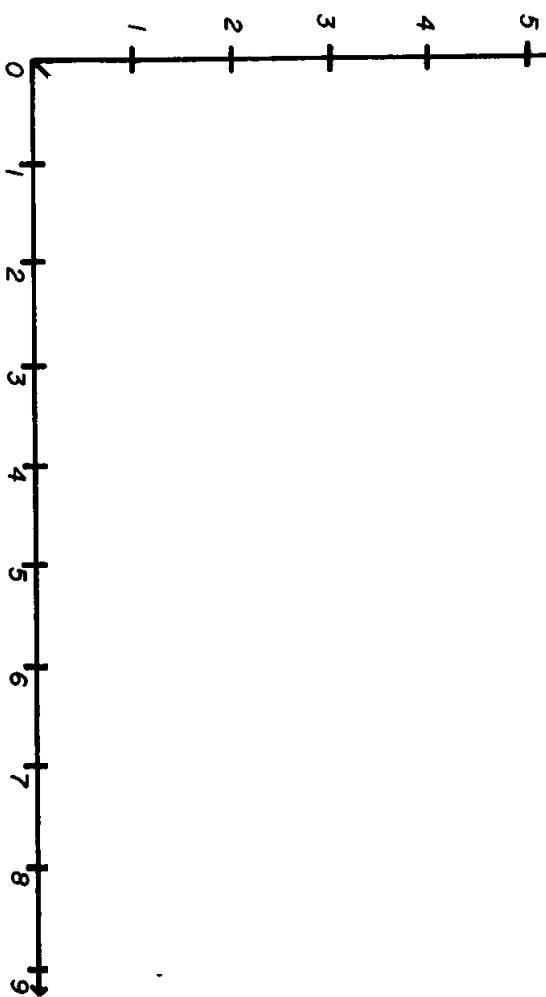
- But the value of the crowd critically depends on the independence of their opinions.
- Independent means uncorrelated.
- If correlated, the value of additional opinions quickly diminishes.

Impact of Correlation

Equivalent Number of
Independent Experts (n^*)

$$n^* = \frac{n}{1 + (n-1)\rho}$$

$n^* \rightarrow 1/\rho$
as $n \rightarrow \infty$



Clemen and Winkler (1985)

Signal Independence

- People are bad at accounting for this effect
 - Even when you tell them exactly what the correlation is, people do not properly adjust (Enke & Zimmerman, 2015).

Signal independence

- Sources of correlation between two opinions?
 - They've discussed it already!
 - They talk to the same people
 - They have the same background – from the same place, trained the same way, same historical experiences, etc.
- Need to find ways to keep opinions independent, and add independent perspectives to experienced groups.

Process vs. Outcome

Consider broader set of objectives

- Organizations generally care about how a person goes about his/her job
 - Most important: impact on others
- People consider too few objectives (Bond, Carlson & Keeney, 2008).
 - Systematically omit nearly $\frac{1}{2}$ of the objectives they later identify as personally relevant
- Leads many firms to rely on too narrow a set of performance measures.

Process vs. Outcome

- Famously hard-charging Dell Computers changed their performance evaluations in the early 2000s.
 - Before change: 100% results
 - After change:
 - 50% what an employee accomplished,
 - 50% how he/she accomplished it, as judged by those affected.

Process vs. Outcome

- The more uncertainty in the environment, i.e., the less control an employee has over exact outcomes, the more a firm should emphasize process in their evaluations.

Focus on process

Use analytics to better understand, and focus on, the processes that tend to produce desired outcomes.

- Key issue: Identify the fundamental drivers of value.



t-1: Shots



t: Goals



t-2: Possession

Focus on process

- What might be the “more fundamental drivers” in, say,
sales?
 - Bids?
 - Meetings?
 - Contacts?

The No-Stats All-Star



• "Knowing the odds, Battier can pursue an inherently uncertain strategy with total certainty. He can devote himself to a process and disregard the outcome of any given encounter. This is critical because in basketball, as in everything else, luck plays a role, and Battier cannot afford to let it distract him."

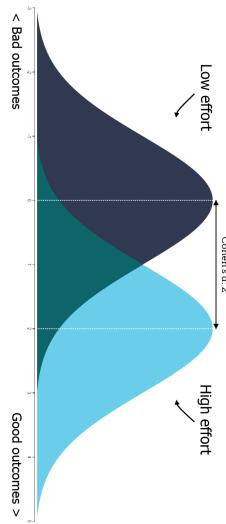
By MICHAEL LEWIS
Published: February 13, 2009

Performance Evaluation Summary

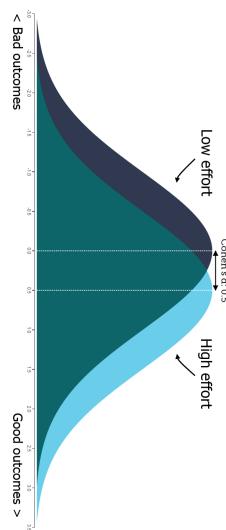
- 1) Understand your environment
 - Know you're biased
 - Account for chance
- 2) Ask the critical questions

Understand your environment

- How much lottery and how much math problem?



vs.



Know you're biased

- 1) Non-regressive predictions
- 2) Outcome bias
- 3) Hindsight bias
- 4) Narrative bias

Account for chance

The key issue: Persistence.

The more fundamental (skill-related) a performance measure is, the more it will persist over time.

The more chance-related a performance measure is, the more it will regress to the mean over time.

Critical Questions

- Are the differences persistent or random? I.e., how do we know this isn't just good/bad luck?
- Is the sample large enough to draw strong conclusions?
How can we make it larger?
- How many different signals are we really tapping into here? How can we make them as independent as possible?
- What else do we care about? Are we measuring enough? What can we measure that's more fundamental?

