





**STANFORD SOCIAL INNOVATION REVIEW PRESENTS** 

### DATA ON PURPOSE

**TELLING GREAT STORIES WITH DATA** 

February 9, 9:55 – 11:00 a.m.

#### Grabbing the Attention of Stakeholders Through Powerful Data Storytelling Techniques



Sacha Litman Managing Director, **Measuring Success** 







### Early Career Personal Example



What did I do wrong?





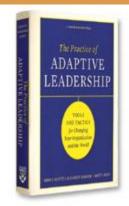
#### Adaptive vs. Technical Challenges

Adaptive Leadership, Heifetz and Linsky

### Technical



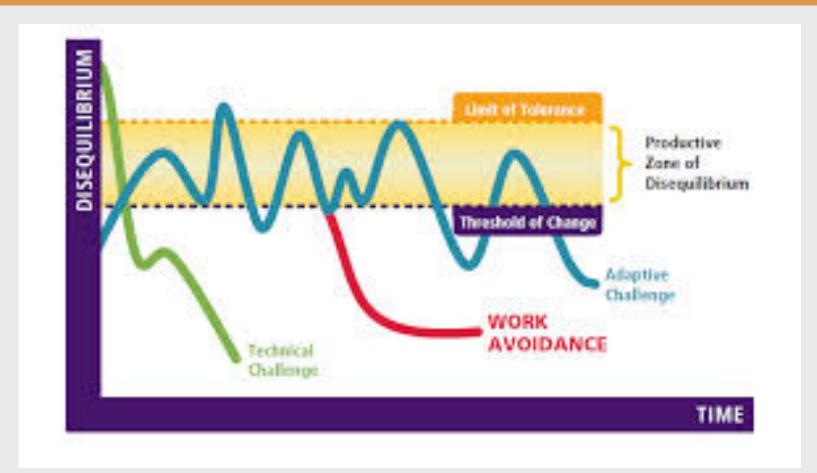








### Must Stay in Productive Zone of Disequilibrium









### Data Analytics Can Be Highly Disruptive to an Organization

#### Stakeholder interests

- Who benefits from status quo?
- Whose job or role is at stake?
- "WIIFM" What's in it for me?
- What is "FUD" Fear, Uncertainty, Doubt

#### Roles in system

Role you play wearing "Data" mantle





## My initial career example: What did I do wrong?





### Where Did Resistance Come From? Data = culture change

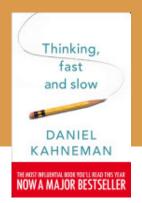
# Anecdote Trumps Data

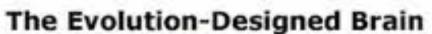




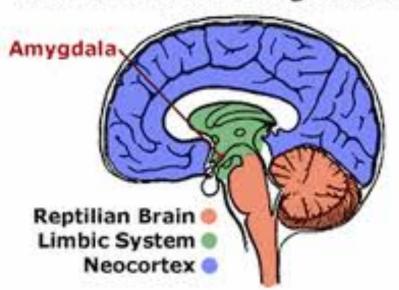
### Why the Cognitive Dissonance?

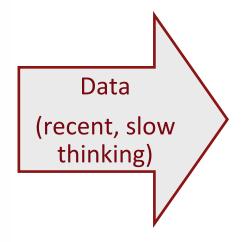
Daniel Kahneman, Thinking Fast and Slow





Emotion (primitive, fast thinking)











#### Data Storytelling Technique

Anchoring your data with a story or anecdote

### Root it in the Amygdala by Telling a Story











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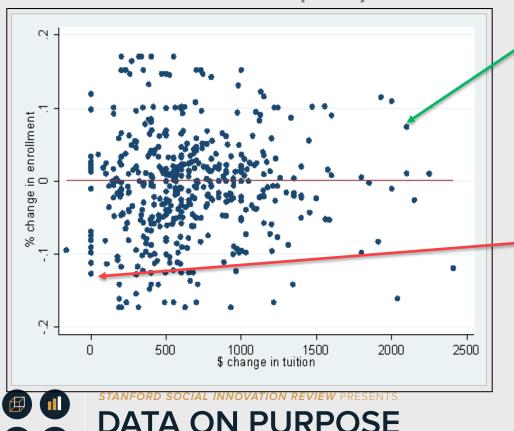
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### Data Is Unsatisfying When Does Not Support Anecdotes





ING GREAT STORIES WITH DATA

This school raised tuition 10% and enrollment grew due to high perceived value

This school kept tuition flat and enrollment fell anyway because poor perceived value

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### How Might You Alternatively Display a Regression To Tell a Story?

Predicted Staff Compensation

Regression A	nalysis			
	Actual Total Compensation	Predicted Total Compensation	Total Compensation \$ Difference	Total Compensation % Difference
Guest	\$110,000	\$97,486	\$12,514	12.8%
Simmonds	\$56,240	\$75,190	\$-18,950	-25.2%
Stewart	\$34,424	\$70,004	\$-35,580	-50.8%
Taylor	\$35,000	\$40,524	\$-5,524	-13.6%
Теггу	\$50,620	\$43,456	\$7,164	16.5%
Todd	\$30,000	\$34,341	\$-4,341	-12.6%
Vickery	\$30,970	\$38,713	\$-7,743	-20.0%
Walton	\$58,495	\$55,194	\$3,301	6.0%
Wellborn	\$30,640	\$42,917	\$-12,277	-28.6%
Whitley	\$27,500	\$31,595	\$-4,095	-13.0%
Williamson	\$33,000	\$31,897	\$1,103	3.5%
Yates	\$45,835	\$39,168	\$6,667	17.0%
Kane	\$46,575	\$52,319	\$-5,744	-11.0%











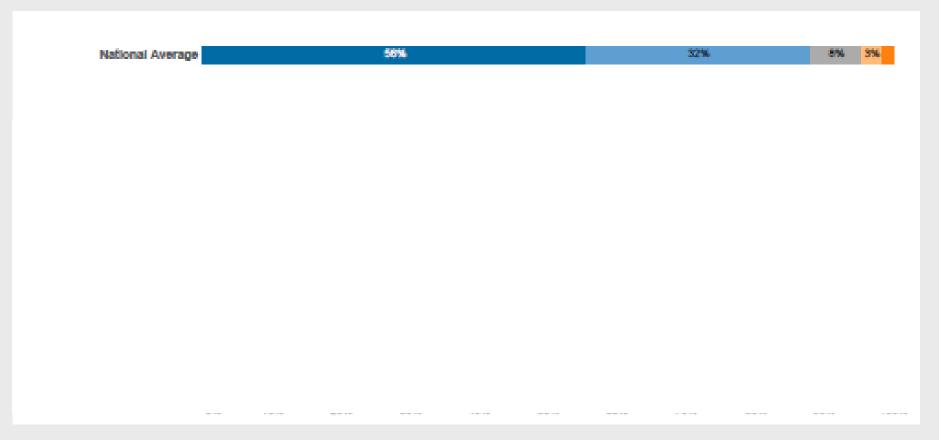
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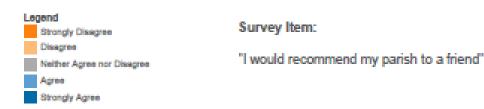
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### Net Promoter Score: Rising Tide Lifts All Boats

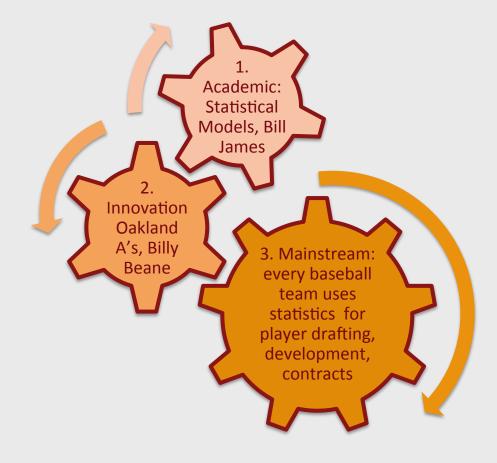






#### Good Movies All Use This Idea











### Moneyball Scene: Note the Anecdotal Stories to Drive Home the Data!

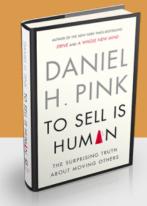


### Moneyball Scene: Note the Anecdotal Stories to Drive Home the Data!



### Pixar Films Classic Story Arc

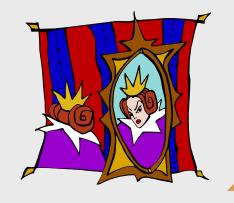
Dan Pink, To Sell is Human





<State the problem>

Every day,
<insert bad
thing –
anecdote that
is not
supported by
data>



One day, <insert good thing – a hypothesis you tested> Because of that, <result>

Because of that, <another good result>

Until finally, <happy ending>















The Oakland A's had no money to spend and lost to teams with bigger payrolls and better players

Every day, their scouts tried but failed to sign the best players by conventional statistics due to lack of funds

One day, the general manager decided to employ a radical tactic: statistics to identify undervalued players

**Because of** that, they signed players that other teams did not want but the data showed would result in wins

**Because of** that, the team started to win and gain confidence

Until finally, they won their division for many years in a row despite the smallest payroll



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### Hypothesis Testing: What drives jumps in donor giving?

#### Pipeline challenges

- 10,000 mid-level donors (\$1-10k) in database in 2000
- 65 of them made the jump to major gift levels (\$25k+) by 2010

#### Time allocation

- Limited number of major donor cultivators, with busy portfolios
- Which of 10,000 donors to focus on to make the jump?

#### Methodology

- Stage 1: Surveyed 8,000 donors across 35 affiliates
- Stage 2: Propensity match (compare 2 groups):
  - 65 Donors who made the jump
  - 300 Similar net worth donors who did not make the jump

















### Activities Potentially Associated with Giving Increases

Factor	Correlation: +,, or none	Anecdotal Justification
Seat on the board or committee		
Frequency of event attendance		
Program volunteer		
Planned or endowment gift		
Solicitation by volunteer vs. professional		
Frequency of cultivation meetings		
Spousal involvement in your org		





#### Let's Test Your Intuition....

Area	Correlation: +,, or none	Additional Data on Major Donors
Seat on the board or committee		
Frequency of event attendance		
Program volunteer		
Planned or endowment gift		
Solicitation by volunteer vs. professional		
Frequency of cultivation meetings		
Spousal involvement in your org		

### How Many Did You Get Right?

80% of Client Anecdotes
(across 1200 clients) are Not
Supported by the Data...

... So diffuse the power of anecdotes by turning them into hypotheses you can test





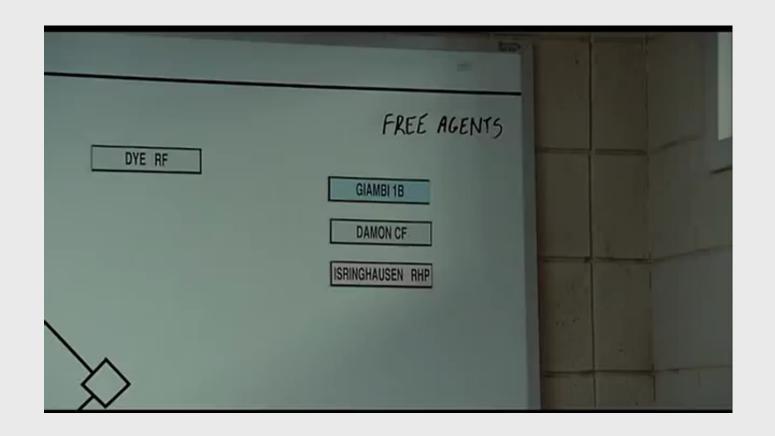
### Did Your Mind Seize on New Anecdotes to Fit the Data Findings?

... Discussing the 80% of hypotheses that were not substantiated can be more impactful than showing the 20% that were





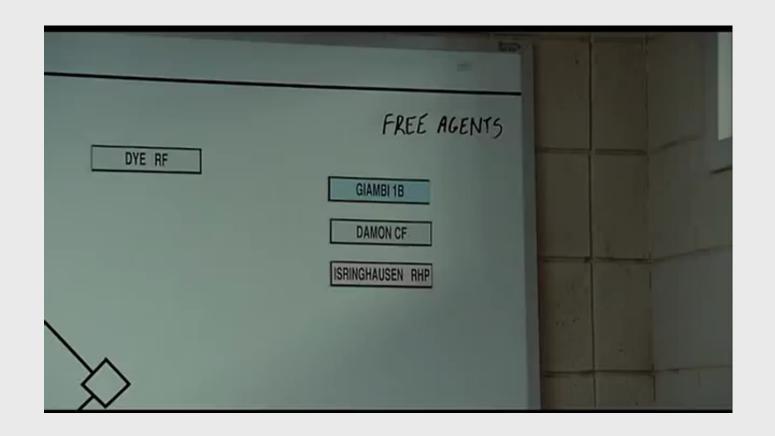
### Moneyball, Anecdotes, and Hypotheses







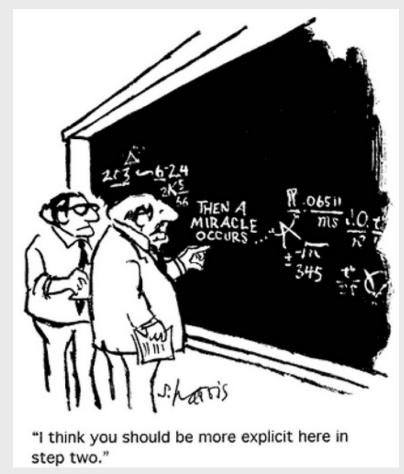
### Moneyball, Anecdotes, and Hypotheses







### Living Too Much in the Data

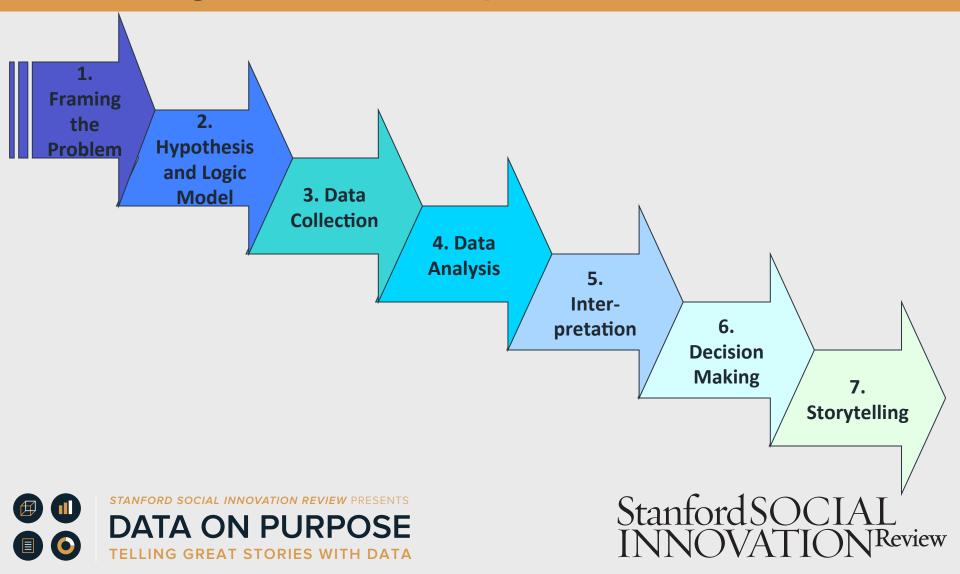




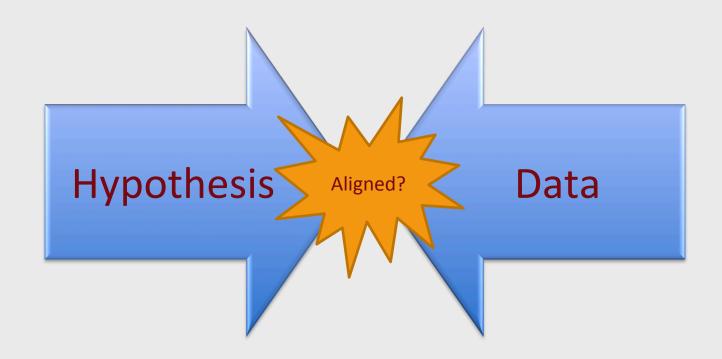




### 7 Stages of Data-Driven Decision Making: Where Nonprofits Fall Short



### Data-Driven Decision Making is About Alignment



Board/managements team's choice: Improve the results (data) or change the hypothesis





### Our Experience at Measuring Success Suggests

- 1. Hypotheses to Test?
- 2. Set up Data Systems
- 3. Analytics and Visualization
- 4. Culture Change Coaching
  - & Data Storytelling





### Principles

- A. Who is the client, what are stakes?
- B. Adaptive change: how maintain disequilibrium?
- C. Since anecdote rules data, how do you frame data findings with anecdote?
- D. Use classic storytelling arcs
- E. Hypothesis testing up front takes power out of anecdotes
- F. Don't shortchange steps in data-driven decision making





















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#### **Audience Questions**



Sacha Litman Managing Director, **Measuring Success** 







