



Evidence-Based HR Management

*Earning A Seat In
The C-Suite*

**2018 NHRMA
Annual Conference**

**In God we trust...
everyone else
must present
evidence**

Evidence-Based Management

- **Origin in evidence-based medicine**
- **AOM & SIOP focus: SHRM?**
- **Workforce management needs a major transformation**
 - **Decisions based on credible evidence**
 - **Decisions made using scientific method**
 - **Decisions that support the business**

Key To The “C-Suite”?

- Executive management wants more than “I think” as support for decisions
- Finance & Marketing have evolved
 - Use of data analytics
 - Adopting scientific method
 - Supported strategy recommendations with tangible evidence, increasing credibility
- HR has lagged but catching up

A Hypothesis

**“If you pay peanuts...
you get monkeys”**

Gunther Klaus

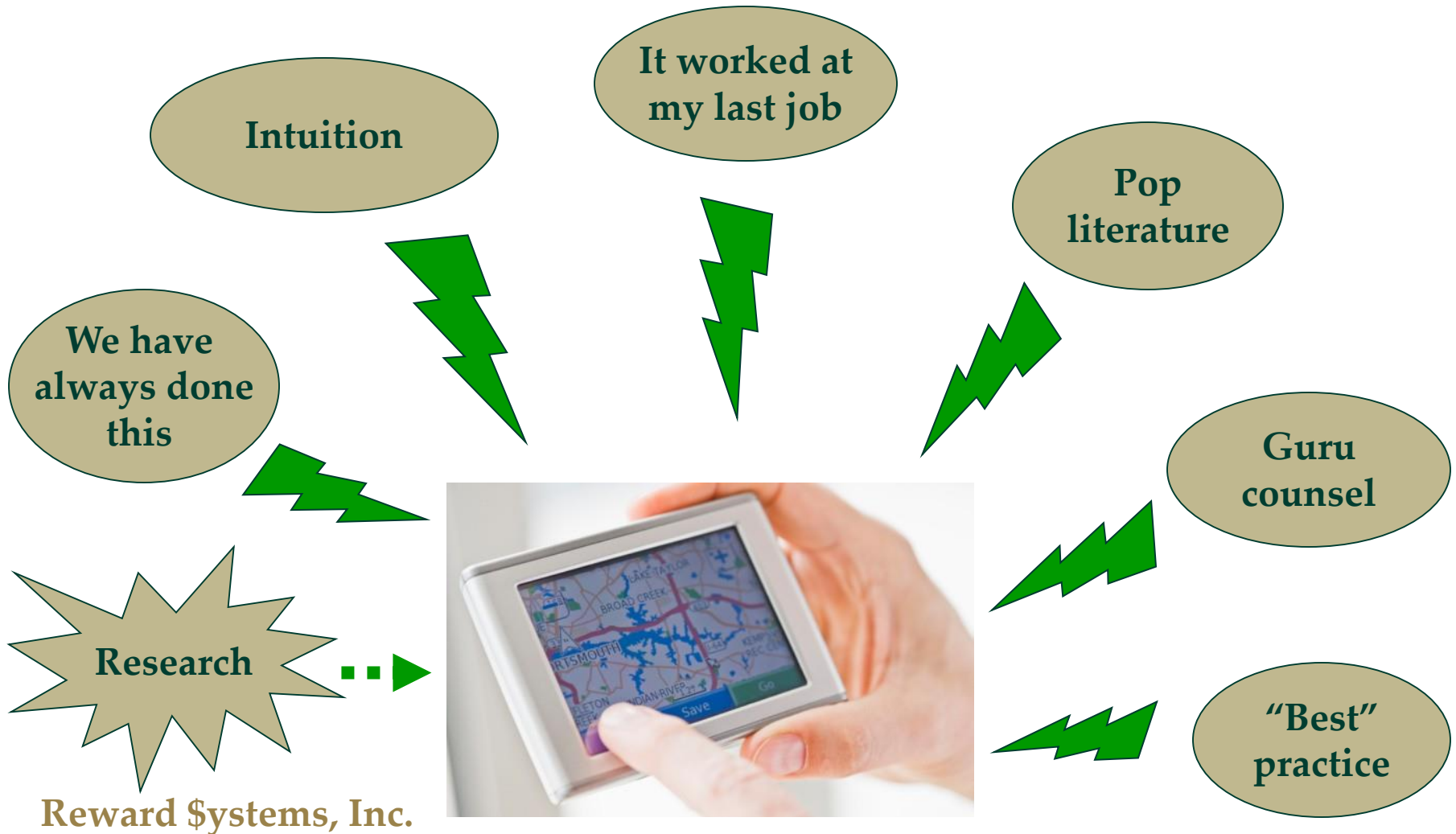
What Are The Underlying Assumptions?

- Pay is important to people
- Paying more is preferable to paying less
- Top contributors are attracted to high pay
- Top contributors are repelled by low pay
- Top contributors have a choice of employers

**Question: That may be what you believe...
but how do you support your assumptions?**

The Practitioner's World

What "evidence" is used in decisions?



What Impedes Use Of Evidence In Decision-making?

**Don't know
it exists**

**Don't
understand
it**

**Don't see its
relevance to
own issues**

**Don't know
how to
apply it**

Let's Test Our Knowledge: True or False?

- “Increasing employee satisfaction will increase productivity”
- “Conscientiousness is a better predictor of performance than intelligence”
- “There will be no difference between rating distributions whether or not the manager must share the rating with the employee being rated”
- “Encouraging participation in decision –making impacts performance more than setting performance goals”
- CLUE: 3 are False; 1 may be True

Two Approaches To Finding & Applying Evidence

- **Deductive (theory > findings)**
 - Identify testable theory in the form of a hypothesis (more pay > less turnover)
 - Identify the assumptions made
 - Test hypothesis
 - Assess, refine/confirm/reject theory
- **Inductive (evidence > findings)**
 - Accumulate evidence
 - Analyze patterns
 - Develop theory

Deductive Is Hard Work

- **Formulating feasible hypotheses**
 - “Above market pay gets/keeps top talent”
- **Defining all the underlying assumptions**
 - “People value pay”
 - “They believe their pay is above market”
 - Other...
- **Constructing a test of the assumptions**
 - *Running tests on impact of pay is risky*

And Inductive Is *“In The News”*

- Literature is full of “big data/analytics”
- All the stories are about successes: surprise? How many people write about their failures?
- Plus, trolling massive data sources is easy
- And, the chaotic environment makes theory formulation difficult
- And, above all, analytics is the “new best thing”

But Danger Lurks

- *If sufficiently tortured data will confess to anything*
- **We are subject to cognitive bias**
 - We more readily accept information that agrees with what we believe/ want to be so
 - We see things in clouds when there are only clouds/random patterns
 - We develop conclusions from inadequate samples
 - We often assume correlation = causation

Common *Serious* Mistakes

- Using measures just because they are quantitative
- Assuming something that can be counted is important, whether it is or not
- Shying away from qualitative measures requiring subjective judgments
- Q: Aren't many of the most critical decisions made based on judgments?

Single Focus On Data Analytics Can Drive Out “Soft Stuff”

- Book/movie *“Moneyball”* was example of the value of using metrics to make better decisions (induction)
- But movie *“Trouble With The Curve”* illustrated that experience and knowledge (deduction) can be a valuable partner with data and analytics

The “Soft Stuff” Counts

- Ryan Leaf was picked ahead of Peyton Manning in the NFL draft (*Ryan who?*)
- Leaf failed not on athletic ability, that can be measured objectively
- Leaf failed on personality, that can only be measured subjectively

But Relying *Only* On Soft Stuff May Not Be Wise

- The *most widely* used selection tool is the one-to-one unstructured interview
- The *least valid* predictor is the one-to-one unstructured interview
- Using some hard data may prompt use of more valid process

Data Analytics

- **Determining what you need**
 - Historical data
 - Trend data
 - Predictive data
 - Prescriptive data
- **Knowing what type of relationship you need**
 - Correlation
 - Causation

Evaluating Data

- **From relevant contexts?**
- **Source of data valid?**
- **Age of data impact usefulness?**
- **Accuracy of data questionable?**
- **Quantitative or qualitative data?**
- **Evaluator have the qualifications and the neutrality to analyze impartially?**

But What If There Is No Data?

- **Data can be found if an organization wants evidence about**
 - What has happened
 - What is happening
- **Innovation/invention may involve conditions that have never been**
- **So if the future will not be like the past/present is there useful data?**

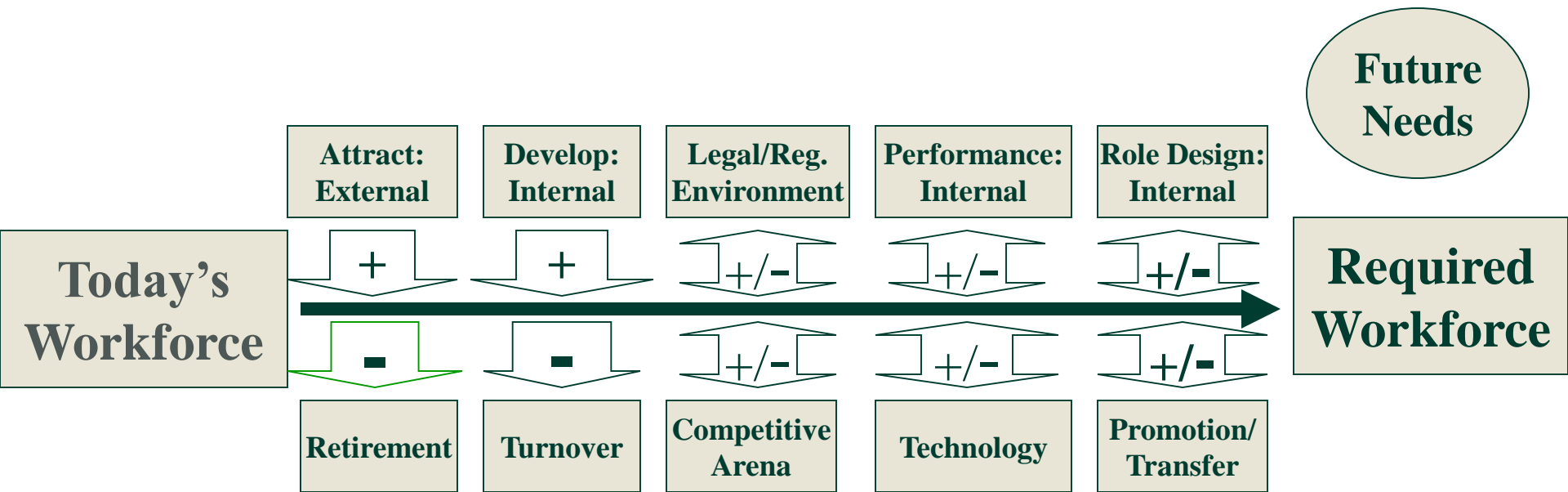


The Tools

Analyzing Data

- **Determine what type of data you need**
 - Historical data
 - Trend data
 - Predictive data
 - Prescriptive data
- **Know what type of relationship you need**
 - Correlation
 - Causation

Example Of Predictive Model: Assessing Workforce Viability Against Future Requirements



Identify sources of supply and losses, project net gains or losses and then determine whether human capital will be adequate

Predictive & Prescriptive Model: Flow Analysis: Control Room Operators

	Entry Level	Journey Level	Senior/Lead
Current Staff	4	6	10
Current Demand	2	12	6
Current Gaps	-2	6	-4
Demand: 1 year out	3 (+1)	12 (-)	6 (-)
Losses projected: next year	1	-1	4
Gaps: 1 year out	0	7	0
Demand: 3 years out	3 (-)	14 (+2)	8 (+2)
Losses projected: next 3 years	4	5	10
Gaps: 3 years out	4	14	12

Progression through levels must be projected as well to determine staffing needs

Reward Systems, Inc.

Making Sense Of Data

- When data points are numerous they need to be clustered into like “chunks”
- Often using an average (or median) disguises the true patterns
- Using histograms (frequency distributions) can enlighten the analyst
“Getting into the weeds” can be key to understanding what the data says

Employee Attitude Survey

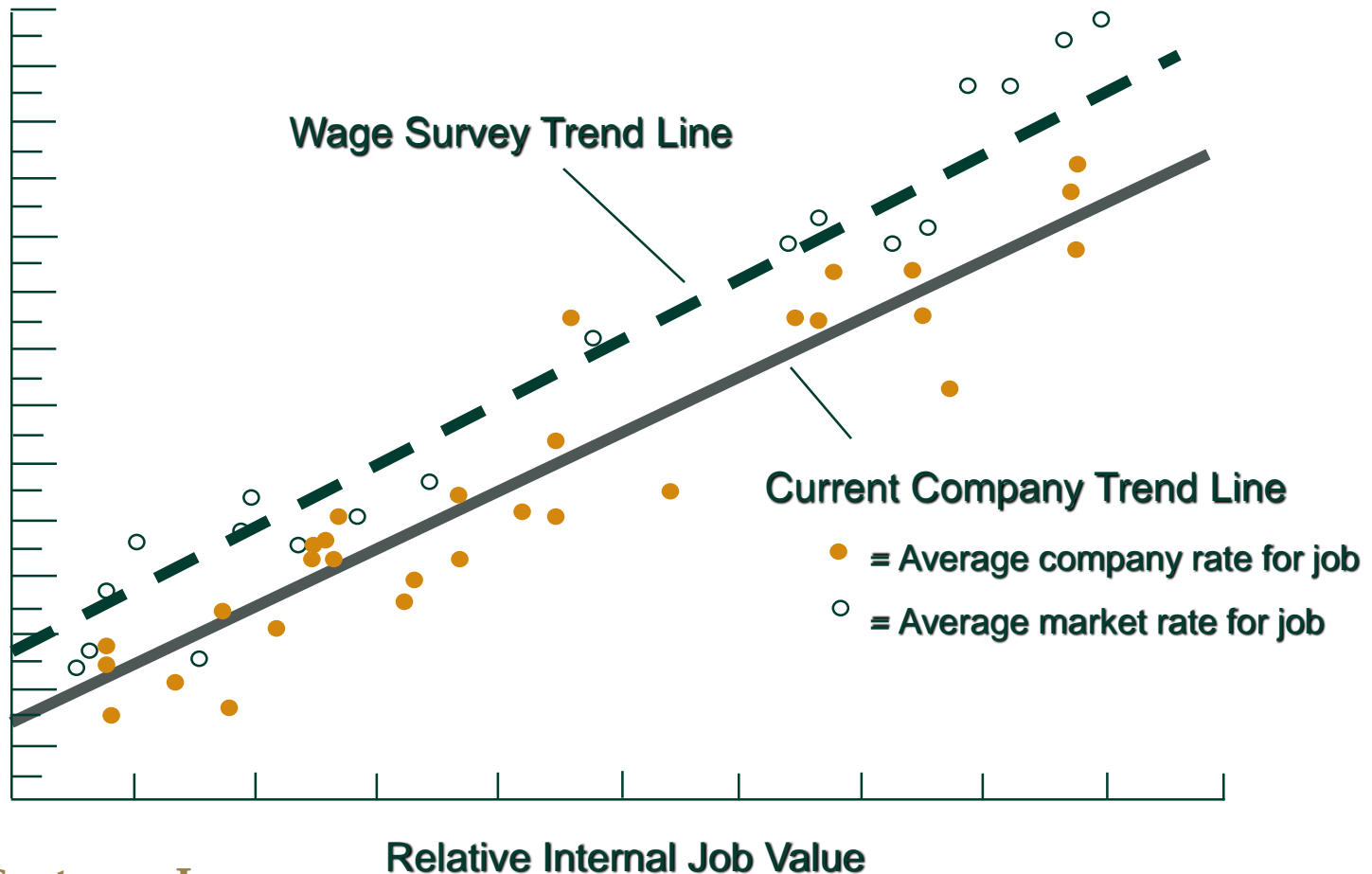
- Common error: report “averages” rather than frequency distributions: what does the average tell you in data below?

										A								X		
										V								X	X	
					X					E						X	X	X		
					X					R					X	X	X	X	X	
					X					A				X	X	X	X	X	X	
				X	X	X		X		G			X	X	X	X	X	X	X	
			X	X	X	X	X	X		E			X	X	X	X	X	X	X	
	X	X	X	X	X	X	X	X	X	A		X	X	X	X	X	X	X	X	
X	X	X	X	X	X	X	X	X	X	V	X	X	X	X	X	X	X	X	X	X

Regression Analysis: Exploring Relationships

Monetary Value

Example: Single Factor Linear Regression



Multiple Factor Regression: Powerful Tool For HR

- **Statistically testing factors impacting selection of new hires**
- **Identifying potential departures**
- **Explaining pay rates... what factors are influencing pay (e.g., performance ratings, longevity in job, grade)**
- **Widely used in testing statistically for discrimination**

An Example Of An Application

- Analysis done using multiple regression, finding what factors impact pay rates. The following are found to have statistically significant positive correlations with pay:
 - Education level
 - Years of experience
 - Job grade
 - Gender
 - Ethnicity
- What would your reaction be? What additional information would you want before you called your in-house/out-house counsel?

Reverse Regression

- **Harris Trust case on gender-based pay discrimination**
 - Organization found pay for women averaged about the same as for men: therefore, conclusion was that there was no problem
 - Statistician showed that women were more qualified as a group, therefore should have been paid more than men on average
 - Lesson: look at everything; from every angle... and then evaluate patterns

Correlation & Causation

- If there is a high *correlation* between A and B it does not establish *causation*
 - A and B may both be caused by C
 - The correlation may be accidental
 - Causation unclear... taller people tend to weigh more, but does one cause the other (or do they co-vary)?
 - B may cause A or A cause B
- Causal path analysis enables relationships to be identified and implications understood

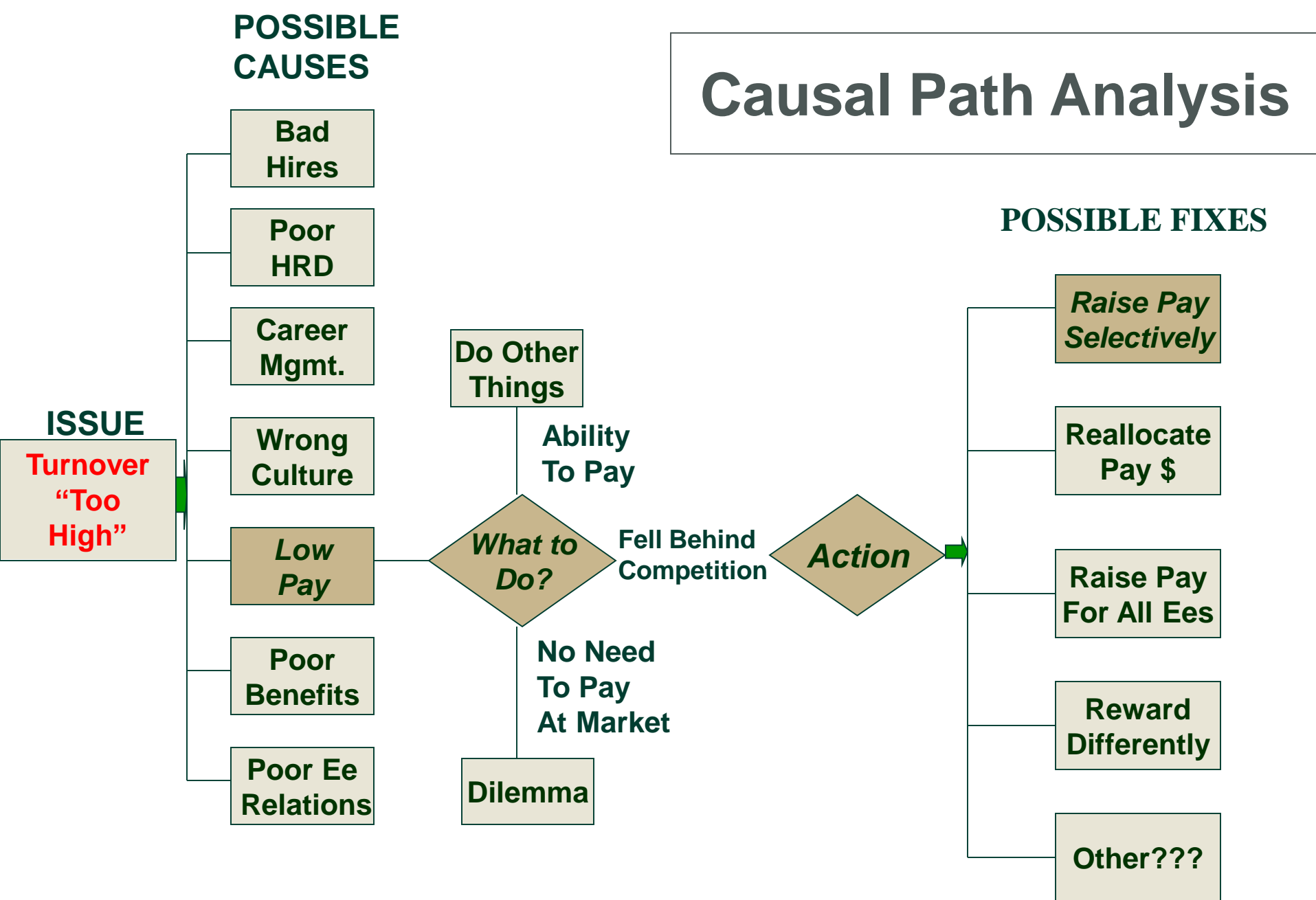
What Causes What?

- **Taco Bell found:**
 - Stores in top quartile in customer satisfaction had better financial performance
 - Stores in top quartile in employee turnover (*it was low*) had double the sales and 55% higher profits than stores in lowest quartile
- **What can be concluded from this?**
- **What was the direction of causation?**

Causation: The Direction Counts

- Does satisfaction impact performance?
- Does performance impact satisfaction?
- What do you think research shows?

Causal Path Analysis



Trend Analysis

- **Be aware of trends in data over time**
 - Cyclical/seasonality
 - Upward or downward pattern
 - Extreme values: timing measurements of improvement (regression to mean)
- **Particularly important in HRM for:**
 - Scenario planning; forecasting future
 - Understanding how significant changes are
 - Incentive plan target setting

Example: Setting Targets For An Incentive Plan

- **Need to set “baseline” to determine the size of the incentive fund, based on the performance level**
- **Need to determine if implementing the plan made a positive difference**

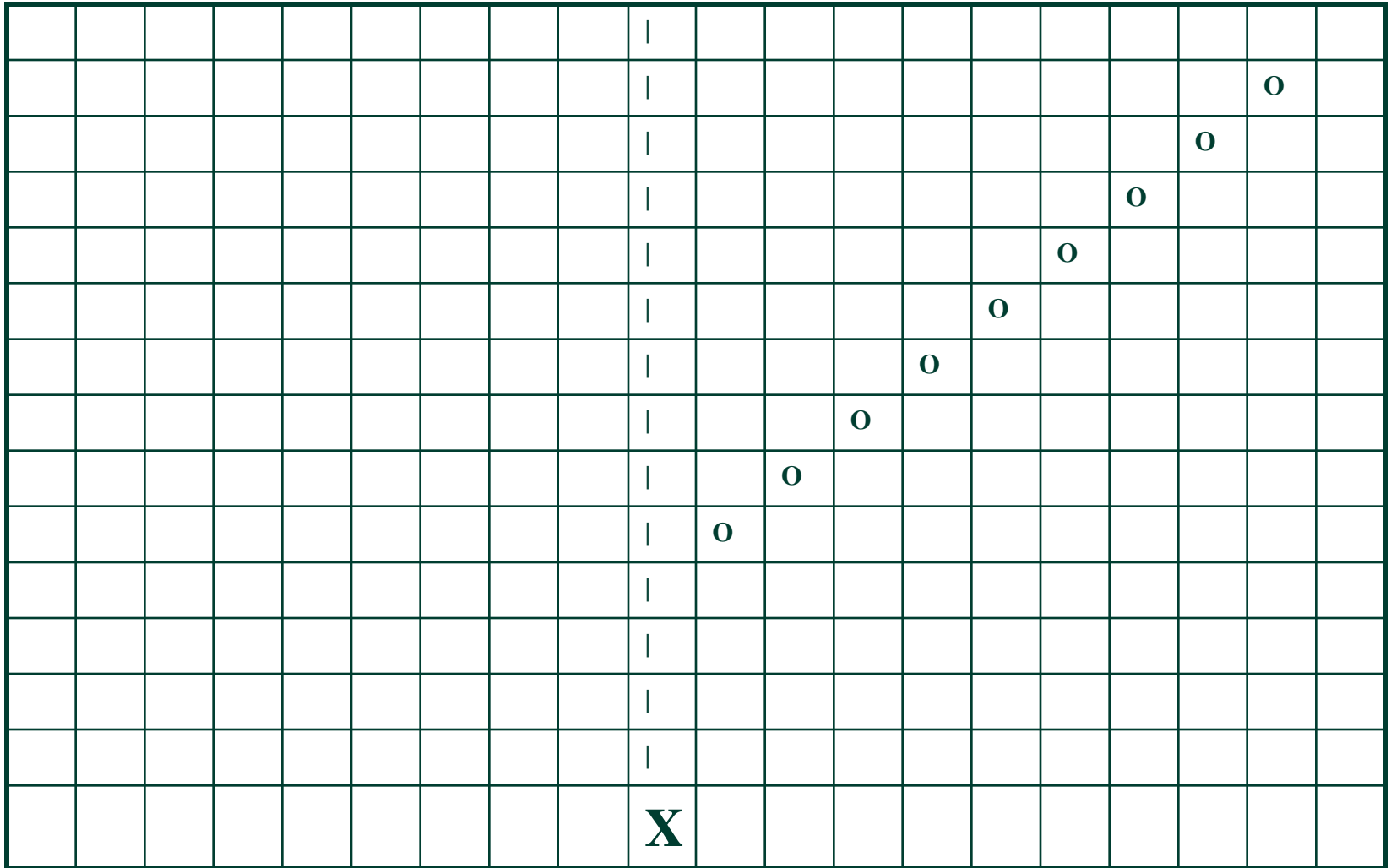
Last 20 Periods: Where To Set Baseline?

																			X
														X					
										X									
												X							
	X					X												X	
			X				X			X									
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		X			X			X											
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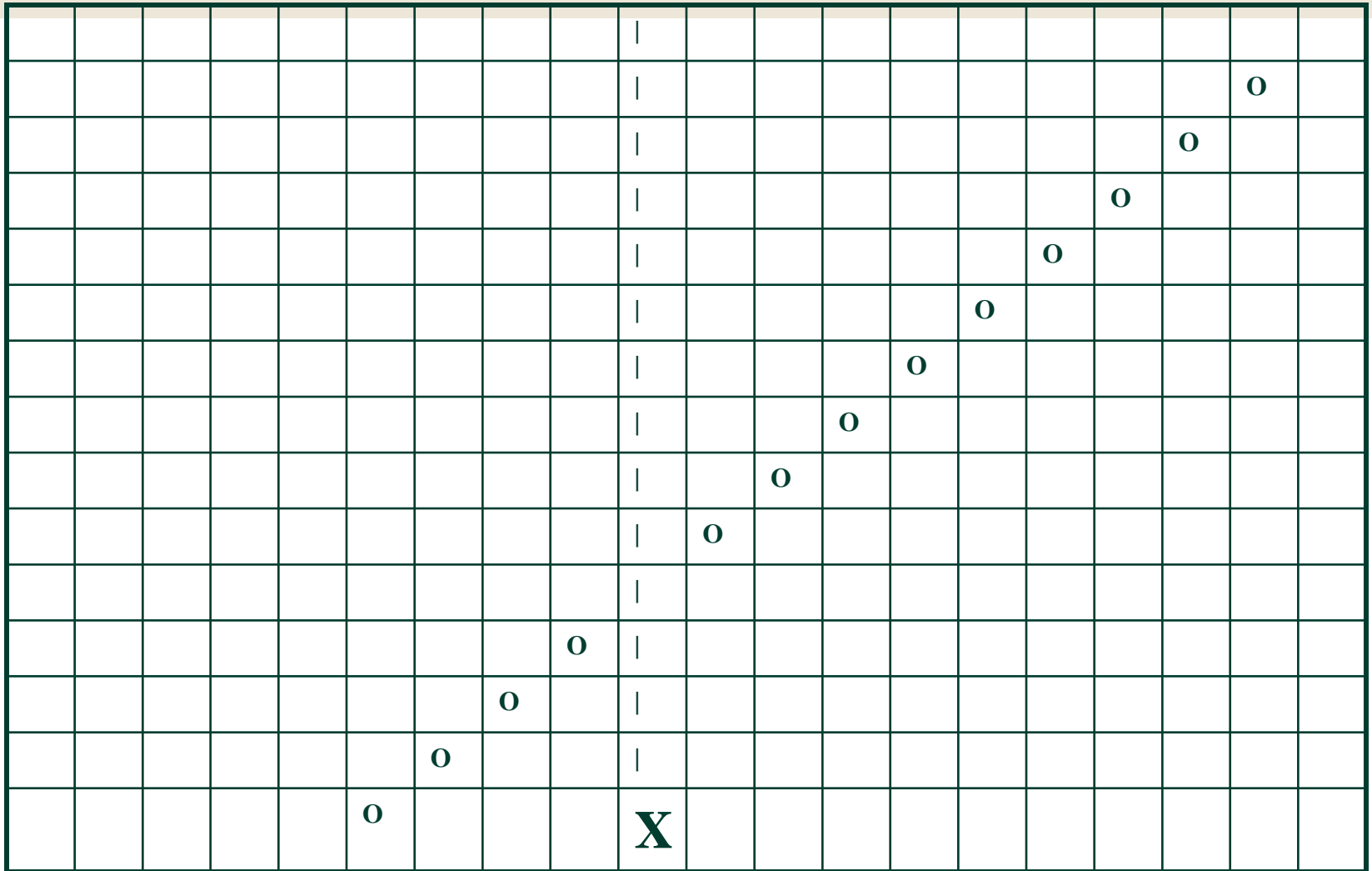
Evaluating Interventions

- **We attempt to intervene**
 - To fix problems
 - To improve performance
 - To change what has been to what we want
- **Evaluating an interventions requires**
 - Understanding what has been
 - Determining if the desired change has occurred

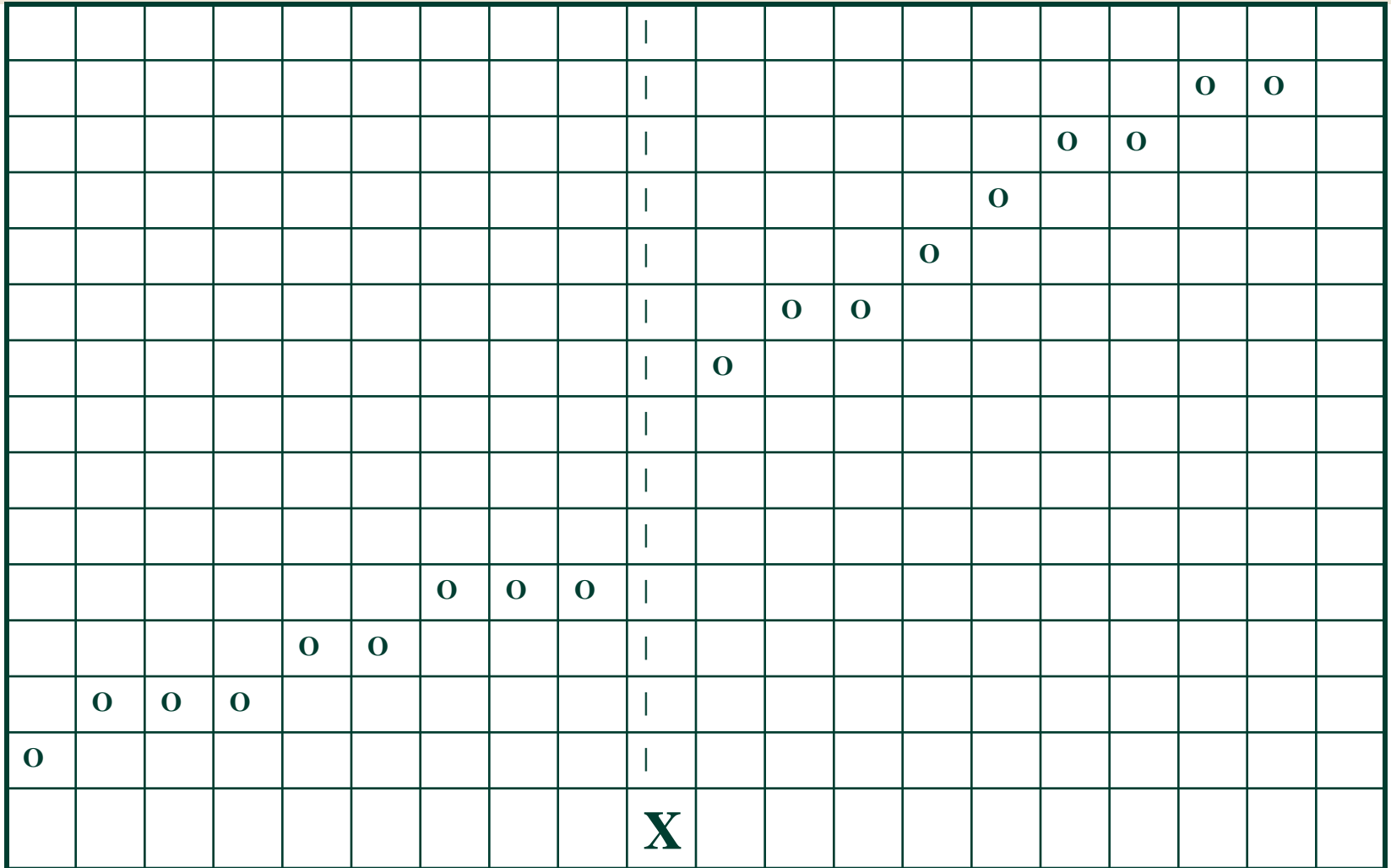
Did "X" Make A Difference?



Did "X" Make A Difference?



Did "X" Make A Difference?



But Are We Sure It Was “X”?

- Did X cause all of the difference?
- What else might have contributed?
- Would X make the same difference the next time?
Under different conditions?

Validity Of Measures

- **Measures used must avoid**
 - **Contamination:** including factors that are not related to what you intend to measure (e.g., rewarding sales person on \$ volume, which includes price increases over which reps had no control, in addition to units sold)
 - **Deficiency:** not including factors that are a part of what is being measured (e.g., appraising performance on quantity without considering quality)

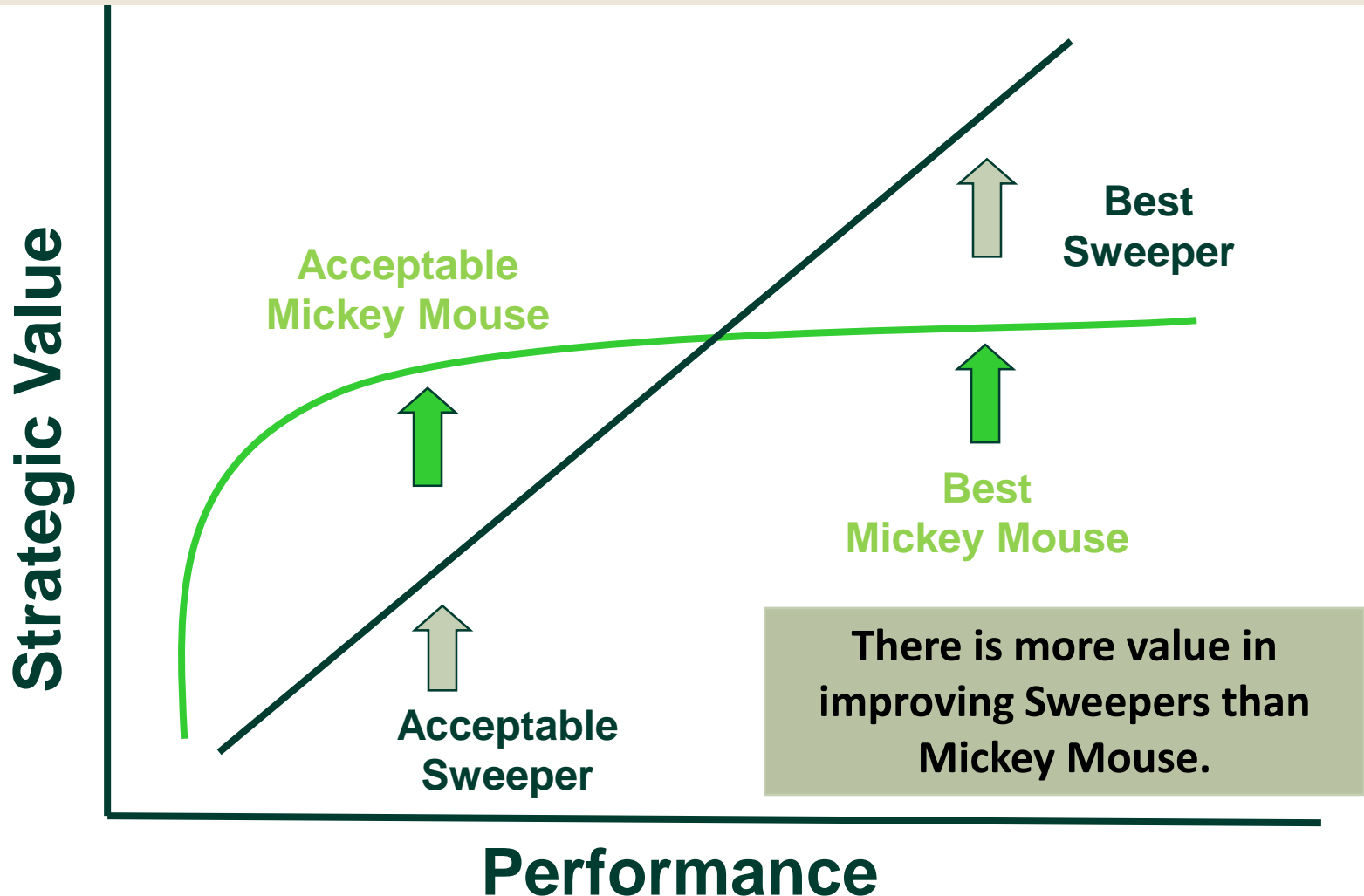
Example Of Poor Measure For Determining Incentives

- Sales reps for manufacturer paid on total sales volume (\$)
- Included were parts orders to repair machines (up to 50% of volume for some of the reps)
- Company needed new machine sales
- Sales rep not involved in parts orders
- So paying for parts volume is _____?

Challenge: Appropriate Scaling

- **5 point performance appraisal scale may not produce useful data**
 - Some jobs are “pass – fail”
- **Do differences make a difference?**
 - Need to know the return on improved performance

How Important Is Performance Variation?

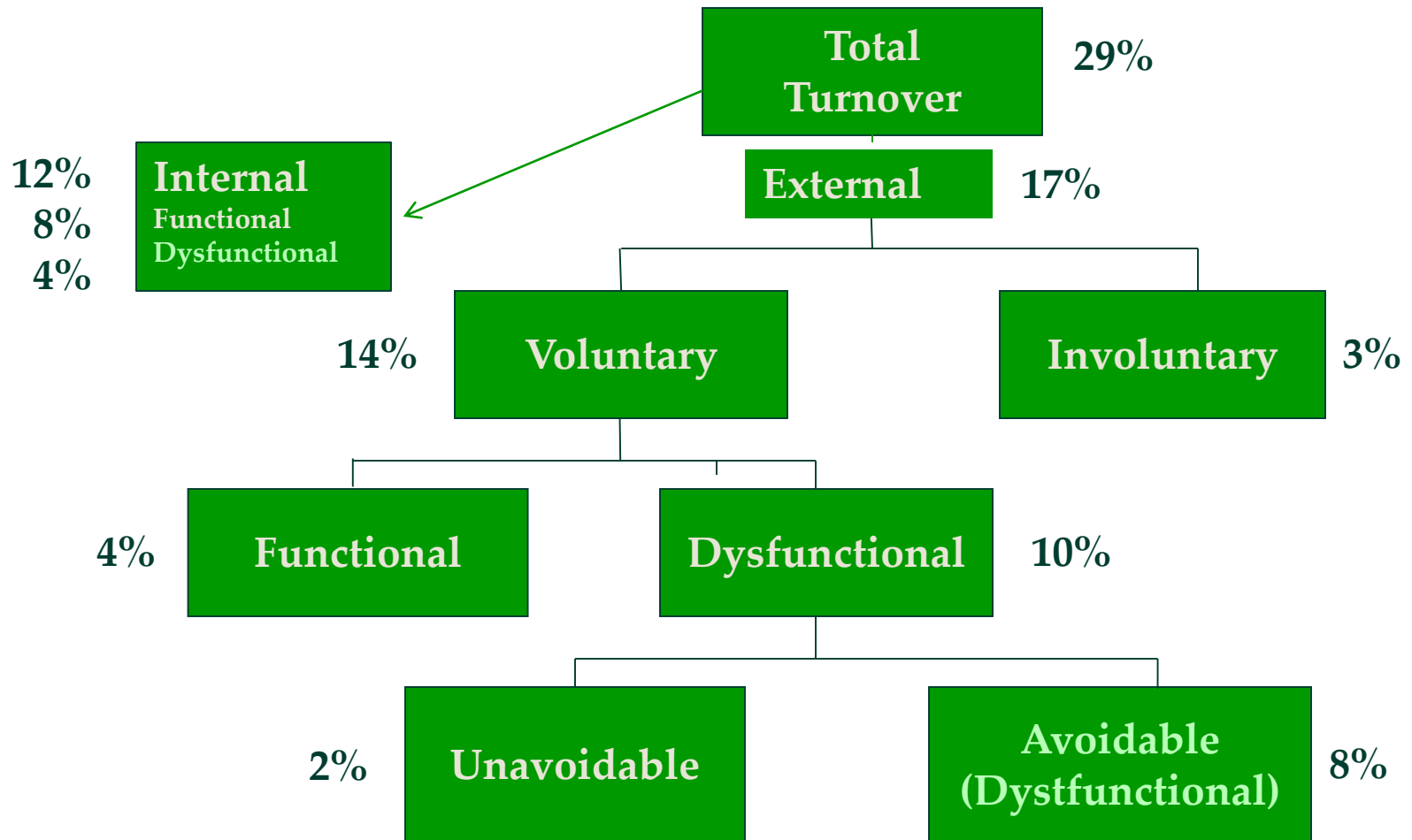


Evaluating Turnover

- **How do you measure turnover?**
- **What is the acceptable range of turnover?**
- **Is it the same for all occupations/roles?**
- **What type of turnover?**
- **What is the impact of the turnover?**
- **What can be done about it?**

Evaluate Turnover

Is It Too High? What Are The Implications?



Is This An Objective Analysis? Not Totally

- **Is internal turnover good or a problem?**
 - Someone has to make an assessment
- **Is turnover to the outside “voluntary”?**
 - Managers can make someone want to leave
- **Is turnover to the outside “functional”?**
 - Was there a celebration when person left?
- **Was turnover avoidable?**
 - Could actions be taken earlier to avoid it?

Bottom Line

- Evidence can be data based or judgment based
- Value of evidence varies depending on the application
- Data can support anything if it is manipulated or selection is poor
- Subjective opinions can be valid and provide unique insights
- We are not going to be replaced by machines!

And A Lot Depends On The Quality Of Your Analysts



Don't be afraid to step out of the box and to use wizards!

Must Haves In Your Library

- **“A Short Introduction To Strategic HRM”**
– Cascio & Boudreau
- **“Becoming The Evidence-Based Manager”** – Latham
- **“Beyond HR”** – Boudreau & Ramstad
- **“Applied Psychology In HRM”**
– Cascio & Aguinis

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