

# Understanding the Pay Gap

Insights from Industry/Practice

- Measuring Pay Gap
- Evaluating Systems and Functional Infrastructure

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# Presentation Outline

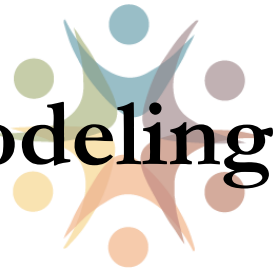


- Insights from Data Modeling
- Proactive Analysis—Practical Considerations
- Defending Against Pay Discrimination Charges—  
An Expert's Insight



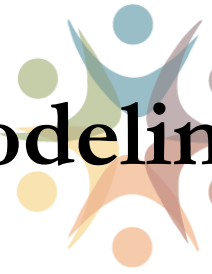
# **INSIGHTS FROM DATA MODELING**

# Pay Disparity Insights from Data Modeling

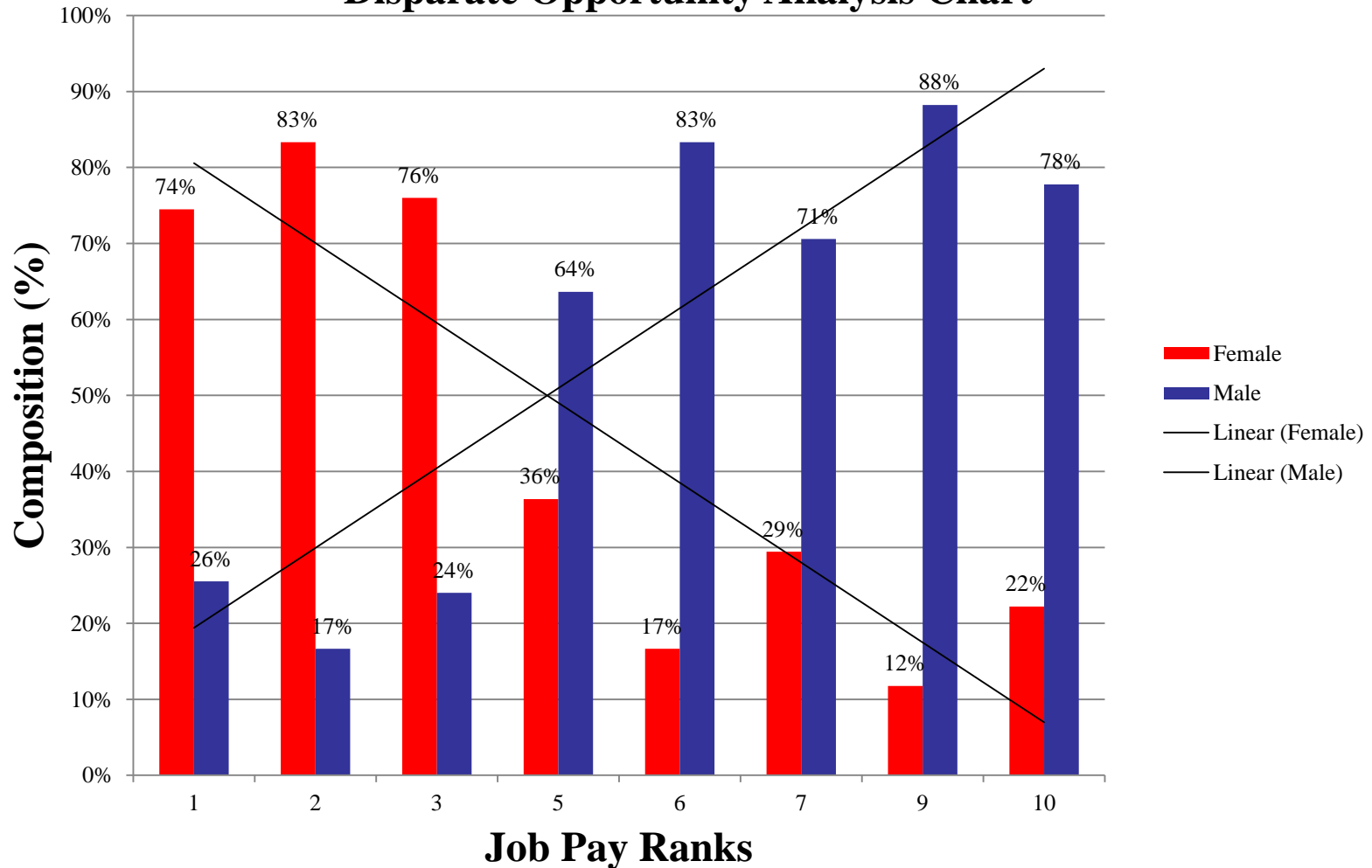


- Pay Gap—Findings from Recent Studies
- Pay Gap is due to placement differences—  
Between-Job Difference
  - CA State Auditor—County Pay Practices
    - <http://www.auditor.ca.gov/pdfs/reports/2015-132.pdf>
  - Glassdoor Study
    - <https://www.glassdoor.com/research/studies/gender-pay-gap/>
    - <http://fortune.com/2016/03/23/unexplained-wage-gap/>

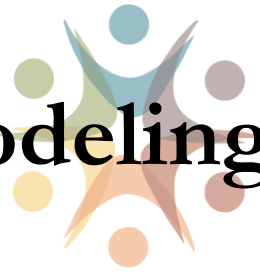
# Pay Disparity Insights from Data Modeling



## Disparate Opportunity Analysis Chart

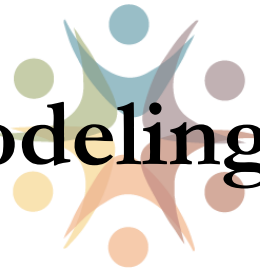


# Pay Disparity Insights from Data Modeling



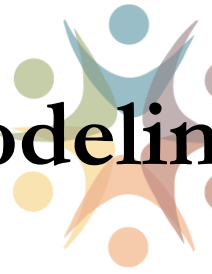
- It is not that simple.
- Experts agree that the Pay Gap is significantly more complex.
- Pay Disparity can be largely accounted for by *Starting Salary*
- It's a complicated mess.
  - Pay Gap is due to differences in Starting Salary WITHIN jobs and Placement BETWEEN jobs?

# Pay Disparity Insights from Data Modeling

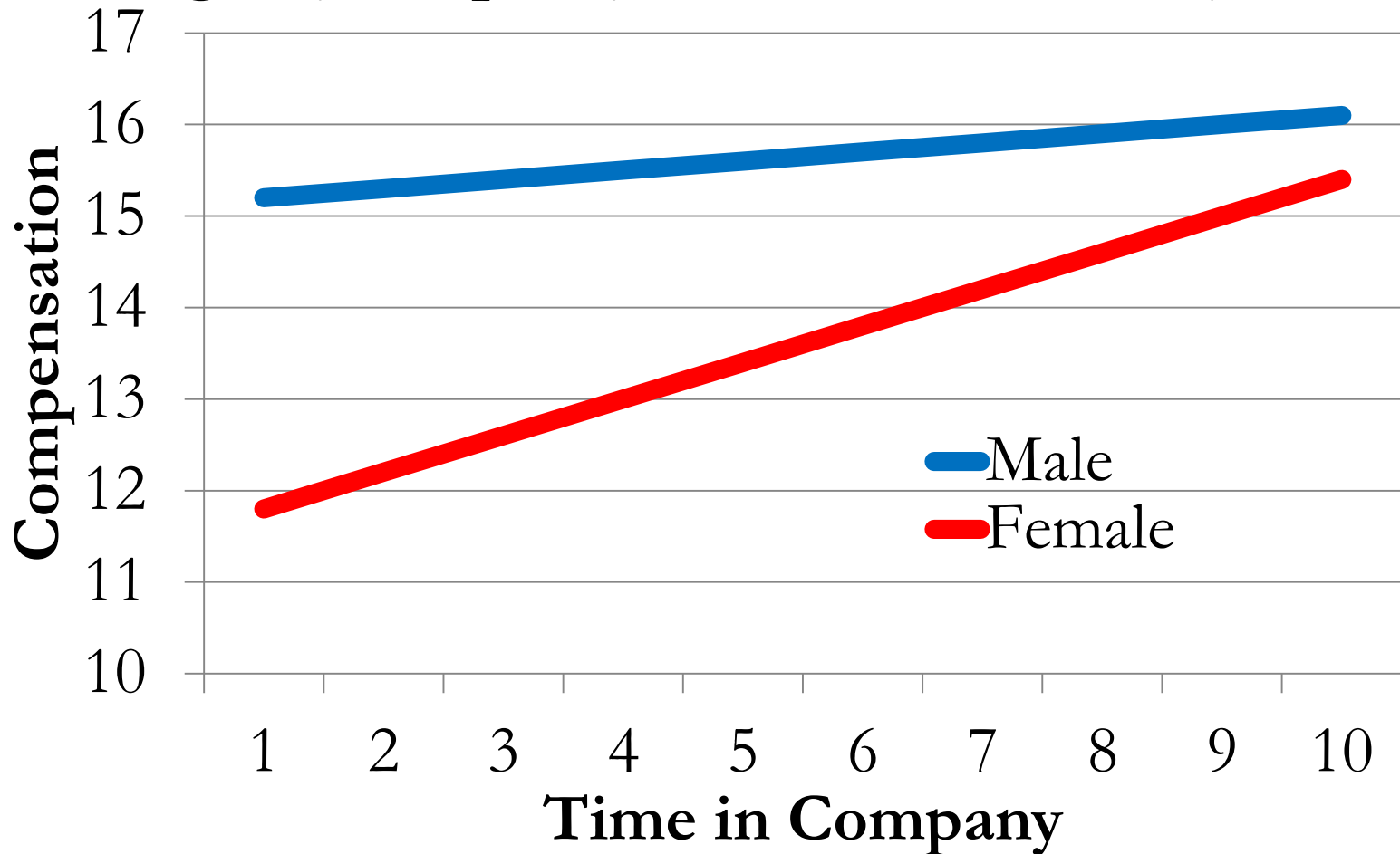


- Pay Gap: double-whammy problem.
  - Women tend to have lower Starting Salary compared to men **WITHIN** the same job.
  - Women tend to have lower Starting Salary compared to men **BETWEEN** jobs (enter into lower paid positions).

# Pay Disparity Insights from Data Modeling



## Modeling Pay Disparity: Visual Data Analytics







# **PROACTIVE ANALYSIS— PRACTICAL CONSIDERATIONS**

# Proactive Analysis--Facts to Consider



- Laws work: Many companies conduct proactive compensation analyses to minimize legal exposure.
- Many proactive analyses are done incorrectly.
  - 10+ Compensation *Common* Mistakes
- Data is rarely “complete” and available.
  - Best data are often not in electronic form.
- Data is often unreliable and not 100% valid.
- Operational/Functioning reality clashes with analytical methods

# Proactive Analysis—*Rarely* Ideal



## Statistical Analyses Is Not (Too) Common

- 75%-90% of S-SSEGs cannot be statistically analyzed due to small **sample size**.
- Most statistical analyses capture 35%-45% of workforce.
- Most of the workforce are not captured in typical compensation analyses.

# Proactive Analysis—*Rarely* Ideal



When things are messy, keep it simple and *focused*.

- Compensation Data (4×):
  - Pay (Base/Total Comp)
  - Group Identity (Gender/Race)
  - Explanatory Factors (TIC, TIJ, PA)
  - (Substantially) Similarly Situated Employee Groups (S-SSEG)
- Simple Models are *best models*, e.g.:
  - $\text{Pay} = \text{Gender} + \text{TIC} + \text{TIJ} + \text{PA}$
  - $\text{Pay} = \text{Gender} + \text{TIC} + \text{TIJ} + \text{PA} + \text{GeoDiff}$

# Proactive Analysis—*Rarely* Ideal



## Statistical Analysis Example.

### *Descriptive*

Data		Male	Female
Gender	Count	15	15
Salary2	Average	\$80,958	\$68,142
Tenure	Average	2.7	2.1
Difference	Raw (\$)		\$12,816
	Pct (%)		16%

### *Outcome*

Impact		Male	Female
	Group		Female
	Amount		\$2,057
	SD		0.47

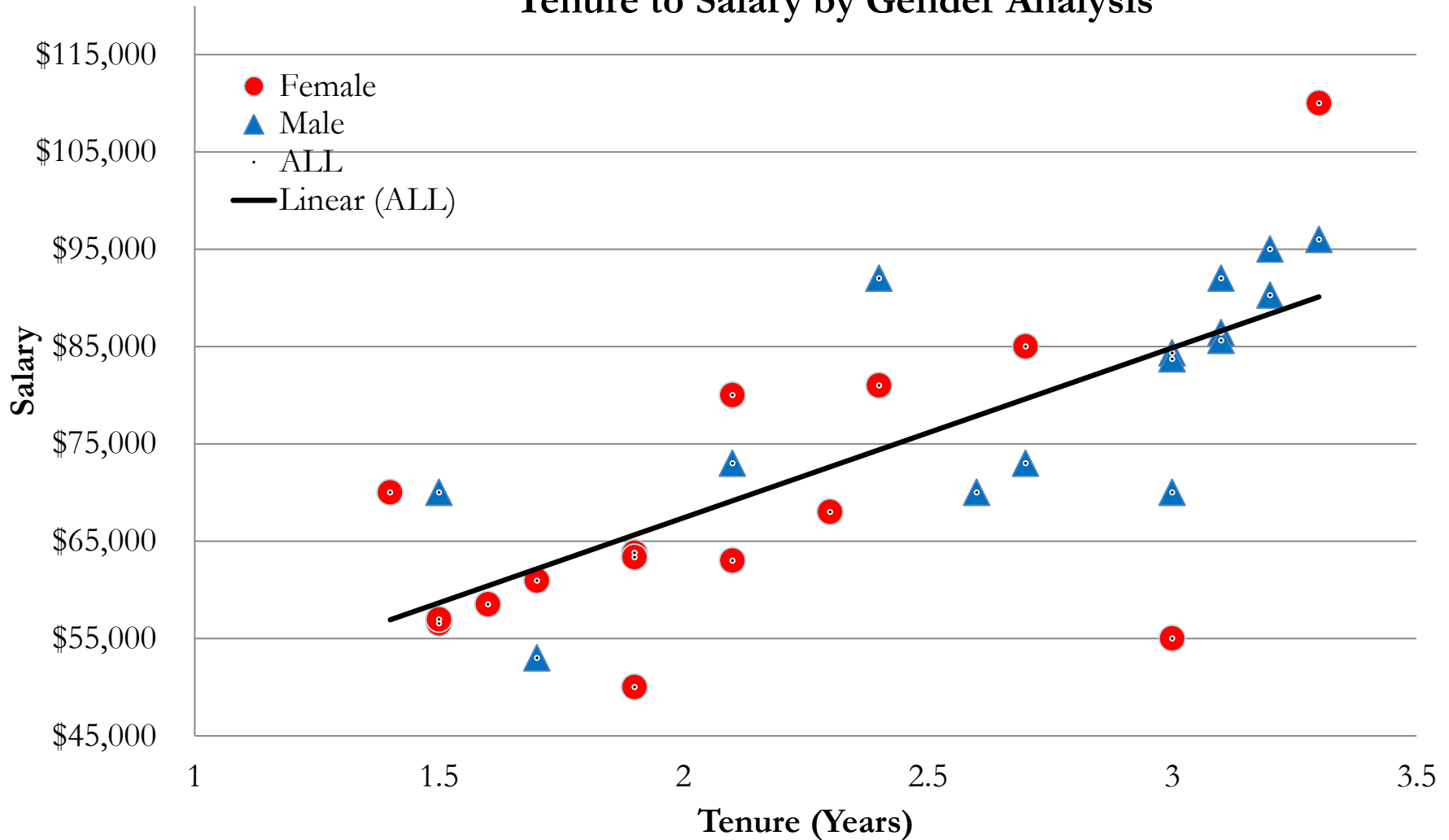
### *Statistical Results*

Regression		Male	Female
Overall	R-Square		0.56
	Sample (n)		30
	p-Value		0.00
Gender	Beta		2057.40
	SD		0.47
	p-Value		0.65
Tenure	Beta		16636.42
	SD		4.78
	p-Value		0.00

# Proactive Analysis—*Rarely* Ideal



## Tenure to Salary by Gender Analysis



# Proactive Analysis—*Rarely* Ideal



## Non-Statistical *Econometric* Data Modeling

- Pay Gap cannot be closed with Statistics alone.
- Non-Statistical Econometric Data Modeling methods are essential to achieving true pay equity.



**DEFENDING AGAINST PAY  
DISCRIMINATION CHARGES—  
AN EXPERT'S INSIGHT**



# Defending Against Claims



The “General” course

1. Claims made
2. Data Analyzed—Statistical
  - a) Plaintiff expert analyze data
  - b) Defense expert analyze data
  - c) Repeat as many iterations as necessary
3. Data Analyzed—Cohort (non-statistical)
  1. Plaintiff identifies a handful of impacted individuals.
  2. Defense explains pay difference for each identified individual with file-pulling.
4. Settlement

# Defending Against Claims



Applying conventional statistical methods, plaintiffs (in general) have a difficult (~impossible) burden.

- Getting a significant model is difficult, e.g.:
  - Small sample size
  - Data challenges
  - Complex models (e.g. aggregated) validity and interpretation
- Pay differences almost always come down to a handful of individuals.
- Most compensation matters fade into small count cohort arguments.

# Defending Against Claims



Defense Experts have an easier job.

- Undermining a model's validity is not difficult.
  - Picking apart and undermining plaintiff expert's models is #1 most effective defense strategy of choice (if defense experts is good).
- Differences can be explained away
  - Cohort analysis
  - Statistical gymnastics
- Defense (generally) has access to more data.

# Defending Against Claims



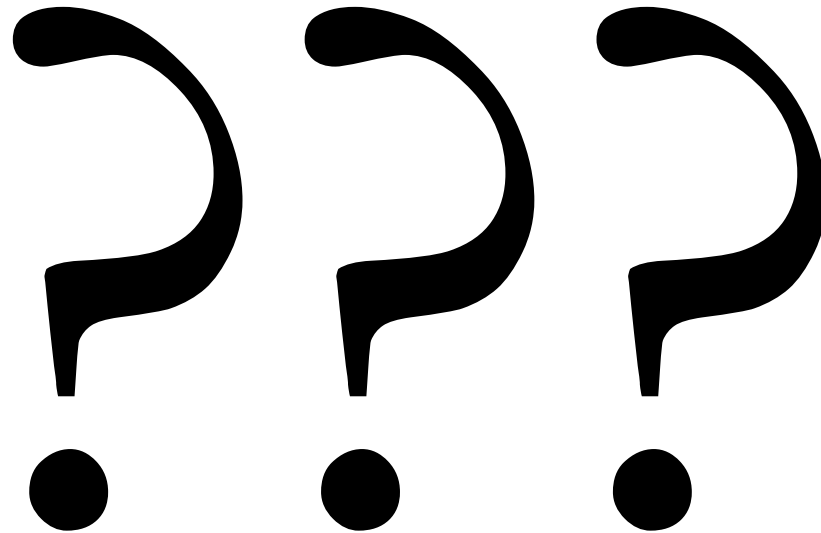
- Recommendations from a defense expert to plaintiffs
- Get a good statistician trained in the social sciences.
  - Avoid a duel of statistical experts.
    - KISS-rule applies: Simple and powerful models are more difficult to deconstruct
  - Develop a deep understanding of context—Depo.
  - Apply unconventional methods and strategies
    - Develop strong non-statistical econometric models earlier than later.
    - Aggregate analyses unconventionally

# Disclaimer



- This is far from comprehensive.
- Given the time constraints and many limitations, this is only a sample of some of the more obvious and important pieces from a practitioner's perspective.

# Questions





**THANK YOU**