



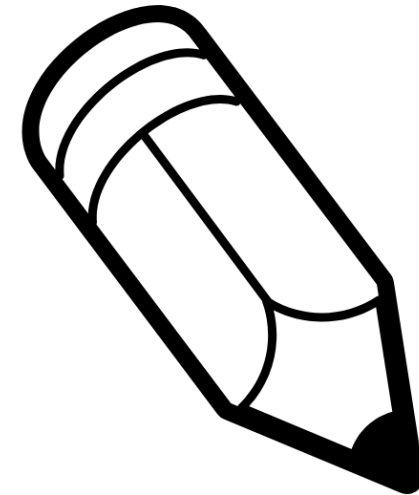
People Analytics & Employment Selection: Opportunities & Concerns

KELLY TRINDEL, FORMERLY CHIEF ANALYST AT EEOC

NEWLY HEAD OF I/O SCIENCE + DIVERSITY ANALYTICS AT PYMETRICS

Outline

- ▶ Introduction & Overview: People Analytics
 - ▶ Promise & Danger
- ▶ Foundational Laws & Regulations
 - ▶ UGESP
 - ▶ Measuring Adverse Impact
 - ▶ Measuring Validity
- ▶ Brainstorming research interests/ideas



People Analytics

- ▶ AKA Workforce Analytics, Talent Analytics, HR Analytics...
- ▶ The application of diverse data sources and machine learning techniques to employment decisions
 - ▶ Employment Selection
 - ▶ e.g. Sourcing, Hiring, Promotion, Discharge
 - ▶ Pay
 - ▶ Succession Planning
 - ▶ Workplace Design
- ▶ Data can be passively compiled or collected directly

What do People Analytics tools look like?

Passive Recruiting tools and screens of passive candidates

Facial expression/tone of voice/language pattern analysis from recorded interviews

Profiling tools that allow employers to select candidates who are similar to a particular profile

Simple or complex 'games' that collect job fitness measurements

Tools designed to track employee movement and communication patterns

????? There seems to be a new tool on the market every week

The Promise

- ▶ Efficiency
 - Automated & scalable
 - Predict rather than describe
 - Improve the candidate & employee experience
- ▶ Effectiveness
 - Demonstrate ROI
- ▶ Job Relatedness
 - Criterion validity is built into the process (cross validation)
- ▶ Fairness
 - Minimize the likelihood of intentional discrimination
 - Remove bias while retaining signal
 - Automate Adverse Impact Analysis
 - Automate the search for less discriminatory alternatives



The Danger



- ▶ Job Relatedness
 - Construct and Content validity evidence often missing
 - Traditional job analysis often missing
- ▶ Fairness
 - Algorithms replicate previous decisions
 - If training data is homogeneous, algorithm results will tend to perpetuate that homogeneity in race, gender, age, etc.
- ▶ Data and computer scientists tend not to be trained in issues of fairness or job-relatedness
 - Employment decisions are much more high-stakes and better-regulated than marketing decisions
 - Optimizing on **accuracy** and not **fairness**.
 - **Predictive** versus **explanatory** analytics

Foundational Laws and Regulations Enforced by EEOC

- ▶ Title VII of the Civil Rights Act – protections on the basis of race, sex, religion & national origin.
 - ▶ Uniform Guidelines on Employee Selection Procedures (UGESP)
- ▶ Title I of the Americans with Disabilities Act - makes it illegal to discriminate against a person with a disability
- ▶ Age Discrimination in Employment Act - protects people who are age 40 or older from discrimination because of age.
- ▶ Genetic Information Non-Discrimination Act – protections for genetic information.
 - ▶ Including information about family members, as well as information about any disease, disorder or condition of an individual's family members (family medical history).



Theories of Discrimination

- ▶ Employment tests and screens can be very effective, but their use must be lawful
 - ▶ Disparate Treatment: Cannot be used to *intentionally* screen out people of a certain race, sex, national origin, religion, disability, or age (40 or older).
 - ▶ Disparate Impact: **Even if the discrimination is not intentional**, these measures cannot screen on protected characteristics *unless the Employer can properly justify their use*
- ▶ Landmark Supreme Court Case: Griggs v. Duke Power (1971)

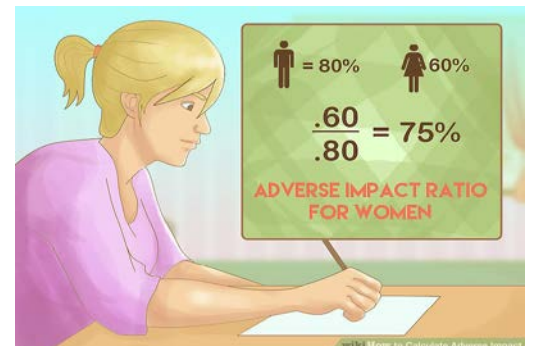
UGESP

- ▶ The Uniform Guidelines (EEOC et al., 1978)
 - ▶ If there is statistical evidence of adverse (disparate) impact the employer must be able to demonstrate:
 - ▶ The validity of the procedure
 - ▶ Job-relatedness
 - ▶ (Test prep) Fairness
 - ▶ Applicants/employees had equal access to any available preparation materials
 - ▶ Attempts to identify equally-valid alternative selection devices with less impact
 - ▶ Additional attempts to reduce adverse impact

Measuring Adverse Impact

§ 1607.4 (D) *Adverse impact and the 'four-fifths rule.'* A selection rate for any race, sex, or ethnic group which is less than four-fifths (4/5) (or eighty percent) of the rate for the group with the highest rate will generally be regarded by the Federal enforcement agencies as evidence of adverse impact

- ▶ Smaller differences in selection rate may nevertheless constitute adverse impact, where they are significant in both statistical ($p < .05$) and practical terms or where a user's actions have discouraged applicants



Measuring Validity

- ▶ **Criterion-Related Validity** - The extent to which test scores are systematically related to a relevant criterion
 - ▶ Criterion usually defined as some measure of job performance
 - ▶ Measures of job performance themselves may be biased (e.g. absenteeism)
 - ▶ Reverse-engineering to demonstrate criterion-related validity, providing built-in defense
- ▶ **Content Validity** - The extent to which the items on a test are representative of the construct the test measures
 - ▶ In employment, the construct the test measures is the ability to do the job
 - ▶ Requires a qualitative/quantitative study of the job itself, identification of its essential functions, KSAs
- ▶ **Construct Validity** - Involves accumulating evidence that a test is based on sound psychological theory
 - ▶ Convergent & divergent evidence that the construct is what you think it is

§ Sec. 1607.15 Documentation of Impact and Validity Evidence

- ▶ **Users** [with more than 100 employees] of selection procedures . . . should maintain and have available for each job, records or other information showing whether the total selection process for that job has an adverse impact . . . Adverse impact determinations should be made at least annually for each such group which constitutes at least 2 percent of the labor force...
- ▶ Where a total selection process for a job has an adverse impact, the user should maintain and have available records or other information showing *which* components have an adverse impact.
- ▶ Where there is evidence of adverse impact, the employer should have evidence of:
 - ▶ Validity of the selection device
 - ▶ Attempts to reduce AI

Summary



- ▶ People Analytics has to do with the application of diverse data sources and machine learning techniques to employment decisions
- ▶ Foundational laws to protect people from unfair decisions based on protected characteristics
- ▶ UGESP is a set of guidelines for using employment selection tools without violating Title VII of the Civil Right Act
 - ▶ It establishes the concept of disparate (adverse) impact, which **need not be intentional**
 - ▶ Gives a general outline: Device should be fair, and job-related. Should optimize on fairness
 - ▶ UGESP is now 40 years old. It was not written with machine learning or people analytics approaches in mind
 - ▶ Some say it's not equipped to handle more contemporary techniques
 - ▶ Its likely that, at some point, regulatory agencies will pass guidance to address
 - ▶ Is likely that accumulating case law will address

Brainstorming Research Interests



▶ Fairness

- ▶ Investigating relationships between passive data and protected characteristics
 - ▶ Or intermediary variables related to both?
 - ▶ Including age, genetic information, disabilities

▶ Validity

- ▶ Investigating *reliability* of passively-collected data
 - ▶ Can it lead to false assumptions about people?
- ▶ Construct: Is personal history really related to career, work, job performance?
 - ▶ Biodata in I/O psychology
 - ▶ Example: credit history and job performance...?

▶ Explanatory versus predictive analytics; non-biased comparisons

- ▶ Help regulators understand this
- ▶ What's the model for properly combining the two?

Questions :: Comments :: Ideas

- ▶ Kelly Trindel, PhD
- ▶ Kelly@pymetrics.com

