

# Topic Analysis and Synthesis on “Don’t Just Evaluate Candidates on Skills”

Akshit Rameshkumar Desai<sup>1</sup>

Concordia University, Montreal, QC, Canada.  
`akshitrameshkumar.desai@mail.concordia.ca`

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**Abstract.** This Topic Analysis and Synthesis report explores engineering management, drawing insights for the topic “Don’t Just Evaluate Candidates on Skills”. It introduces a novel approach to evaluating engineering candidates, emphasizing the importance of balancing technical skills with personal values and abilities. This methodology aims to create well-rounded teams where diverse skills and personalities complement each other. Traditionally, hiring practices have overly focused on technical expertise, which can lead to challenges in team dynamics and effectiveness. The report advocates for considering a broader range of attributes, including shared values and soft skills, to form stronger, more cohesive teams. Additionally, the analysis will reference and examine other studies and resources to enrich the understanding of effective

team building in engineering. By blending practical advice with theoretical foundations and real-world examples, this TAS aims to provide a comprehensive view of successful engineering team management in the modern context.

**Keywords:** Candidate Evaluation · Skill Assessments · hiring smart

## 1 Introduction

### 1.1 Motivation

While skill assessment forms the cornerstone of the hiring process, it is increasingly evident that a sole focus on technical abilities is insufficient for assembling high-performing teams. The motivation for this research stems from the understanding that teamwork and productivity are significantly influenced by factors beyond mere skill proficiency. These include interpersonal dynamics, value alignment, adaptability, and emotional intelligence. In today's collaborative and agile work environments, these attributes play a pivotal role in ensuring that teams can effectively communicate, innovate, and respond to changing circumstances. This section aims to delve into why it's crucial to adopt a more holistic evaluation approach that transcends traditional skill assessments.

### 1.2 Problem Statement

The central question this research seeks to address is: "Is it important to evaluate candidates on factors other than just skills, and if so, what are these factors and how do they affect a team's productivity?" This inquiry is significant in the context of evolving workplace dynamics where team synergy and cultural fit are becoming as crucial as technical expertise. The problem statement will explore the potential risks and limitations of a narrowly focused skill-based hiring model, and investigate how other attributes like values, work ethic, and interpersonal skills contribute to the overall effectiveness and productivity of a team.

### 1.3 Objectives

- **Identifying Benefits:** To outline the advantages of a comprehensive candidate evaluation approach, emphasizing how it enhances team cohesion, innovation, and overall productivity.
- **AI in Assessment:** To explore the feasibility and effectiveness of using Artificial Intelligence (AI) tools in assessing non-technical attributes of candidates, such as values alignment and soft skills.
- **Cost Analysis:** To evaluate the additional costs involved in a more comprehensive hiring process, including potential increases in hiring time, human hours, and resources. This will assess whether the long-term benefits justify these additional costs.
- **Role of Additional Interviews:** To examine if incorporating more in-depth interviews focused on values, behavior, and soft skills can significantly improve hiring outcomes.

## 2 Background Material

### 2.1 Evolution of Hiring Practices

- **Early Hiring Models:** Traditionally, hiring models have predominantly focused on assessing candidates' technical skills and academic credentials. This approach, grounded in a straightforward assessment of quantifiable skills, often involved evaluating candidates based on their education, experience, and demonstrated expertise in specific skill areas relevant to the job. This method, while effective in identifying technically proficient candidates, often overlooks other critical aspects such as interpersonal skills, adaptability, and cultural fit.
- **Shift Towards Soft Skills and Cultural Fit:** In recent years, there has been a significant shift in hiring practices towards valuing soft skills and cultural fit. This change stems from the growing recognition that effective communication, teamwork, problem-solving, and adaptability are vital for success in most modern workplaces. As businesses increasingly value innovation and collaborative work cultures, the ability of employees to integrate into teams and contribute positively to the company culture has become crucial.
- **Impact of Technology and Globalization:** The advent of technology and the rise of globalization have further transformed hiring practices. With remote work and global teams becoming more common, there is an increased emphasis on hiring individuals who can work effectively in diverse, dispersed teams. Technology has also enabled more sophisticated assessment tools, including AI-driven analytics, to evaluate candidates' soft skills and cultural fit, broadening the scope and accuracy of the hiring process.

### 2.2 Importance of Team Dynamics in Productivity

- **Team Synergy:** It refers to the way individual team members' skills and attributes complement each other, leading to enhanced collective performance. When team members have a mix of technical skills and soft skills, they tend to collaborate more effectively, leading to increased creativity, problem-solving, and productivity. Effective team synergy relies not just on the individual competencies of team members but also on their ability to communicate, adapt, and work together towards common goals.
- **Challenges of Diverse Teams:** While diverse teams can bring a range of perspectives and skills, they also present unique challenges. Differences in cultural backgrounds, communication styles, and work ethics can lead to misunderstandings and conflicts. Therefore, it's crucial for hiring processes to focus on finding candidates who not only bring diversity in skills and perspectives but are also adept at working in diverse environments. This includes the ability to respect differences, communicate effectively across cultural barriers, and adapt to various working styles[3][1]

### 3 Methods and Methodology

#### 3.1 Case study-1: Assessments of Organisational Fit

The paper, "Matching Candidates to Culture: How Assessments of Organisational Fit Shape the Hiring Process," delves into how employers and recruitment consultants define and apply the concept of organizational fit in professional labor markets like engineering, marketing, and finance. It explores how this focus on fit can lead to social bias, affecting the inclusivity of the hiring process.[4]

**3.1.1 How study was approached?** The research approaches the problem by examining the role of organizational fit in hiring processes and its potential to create and justify social biases. This exploration is contextualized within professional labor markets, where organizational fit is often a critical, yet vaguely defined, criterion. The study specifically aims to understand how organizational fit is perceived and implemented by external recruitment consultants, revealing the underlying mechanisms that may lead to the exclusion of certain groups or individuals.[4]

- **Semi-Structured Interviews:** The study utilizes semi-structured interviews with 47 external recruitment consultants who assist employers in sectors like engineering, marketing, and finance. This approach offers direct insights into the practices and perceptions of those who play a key role in shaping hiring processes.[4]
- **Theoretical Framework:** The study employs Relational Inequality Theory (RIT) to analyze how organizational fit assessments contribute to inequalities in the hiring process. This framework helps in understanding how exclusionary practices are rationalized and legitimized.[4]

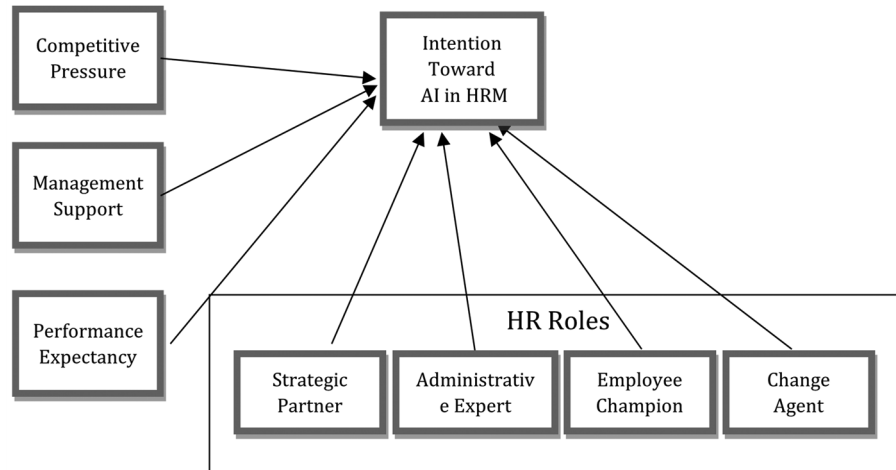
**3.1.2 Techniques used for surveys and analyzing results** The methodology primarily involves qualitative analysis techniques to interpret and synthesize the data obtained from the interviews.

- **Thematic Coding:** The data from the interviews are thematically coded using NVivo 12 software. This involves both inductive and deductive coding to identify key themes related to organizational fit and its role in potential biases and exclusionary practices.
- **Inductive and Deductive Analysis:** The analysis is carried out in two rounds – the first establishes a priori codes related to organizational fit, while the second involves hand-coding to analyze exclusion mechanisms and categorization processes.
- **Claims-Making Analysis:** A critical part of the analysis involves examining claims-making as a process through which exclusion occurs. This aspect focuses on how recruitment consultants and employers use the concept of organizational fit to justify hiring decisions that may be biased.

### 3.2 Case study-2: Adoption of AI in human resource management

The paper primarily investigates the determinants of AI adoption in HRM. It focuses on various factors such as competitive pressure, top management support, performance expectancy, and HR roles within organizations. The study's findings indicate that performance expectancy is a strong predictor of the intention to adopt AI in HRM. It also highlights the positive influence of the 'Change Agent' HR role and the negative association of the 'Employee Champion' role with the intention to adopt AI. The results offer significant insights into the role of AI in transforming HRM practices, emphasizing the strategic importance of AI adoption in enhancing HRM efficiency and quality. This study contributes to the understanding of AI diffusion in HRM, providing valuable information for HR leaders and policymakers to stay informed about the impact of AI on HRM and organizational effectiveness.

**3.2.1 How study was approached?** The study adopted a survey methodology, reaching out to senior HR practitioners who have a significant influence on their organization's decision-making, particularly in the context of technological adoption. The survey aimed to gather empirical data on the hypothesized relationships concerning AI adoption within HRM. A total of 192 members were initially contacted, and after validation for completion and accuracy, 186 responses were considered for final analysis. The survey, conducted in English, ensured anonymity and confidentiality, thus encouraging genuine and unbiased responses from the participants. The below image 1 illustrates the conceptual framework to study constructs and their hypothesised relationships. [1]



**Fig. 1.** Framework for studying AI adoption in HRM

**3.2.2 Techniques used for surveys and analyzing results** The research paper employs a methodical and rigorous approach to examining the adoption of AI in HRM, using advanced statistical techniques to ensure the validity and reliability of its findings. The paper's comprehensive methodology underscores the importance of a multifaceted analytical approach in understanding the complexities of technological adoption in the field of human resources.

- **Claims-Making Analysis:** The research utilized CB-SEM, a robust statistical technique that is particularly effective for path-analytic modeling and evaluating complex relationships between multiple constructs. This technique is advantageous in its capacity to handle multidimensional constructs while concurrently minimizing errors and eliminating weaker measurements. CB-SEM's effectiveness lies in its holistic approach to understanding the interplay between various factors, making it an ideal choice for this study's complex and multifaceted nature.
- **Confirmatory Factor Analysis (CFA):** The first level of statistical analysis involved CFA to test the validity and reliability of the measurement scale used in the survey. CFA is critical in understanding the underlying structure of the data and ensuring that the observed variables accurately reflect the constructs they are intended to measure. By examining factor loadings of each indicator, the study ensured that each measurement adhered to established thresholds for validity. This step was vital in confirming the internal consistency and reliability of the data, ensuring the study's findings were based on a solid empirical foundation.
- **Evaluation of SEM and Hypothesized Relationships:** The second level of statistical analysis focused on evaluating the SEM and testing the hypothesized relationships between the study constructs. This involved a comprehensive assessment of the relationships posited in the study, ensuring that they were not only statistically significant but also meaningful in the context of AI adoption in HRM. The researchers meticulously evaluated the measurement scale's reliability and convergent validity by examining Cronbach's  $\alpha$ , Average Variance Extracted (AVE), and Composite Reliability (CR) Coefficient for the constructs. This thorough examination of the data solidified the study's foundation, providing a robust platform for the ensuing analysis and conclusions.

Moreover, the methodology also encompasses the role of AI in evaluating non-technical aspects of candidates. AI can enhance the hiring process by assessing candidates' values, abilities, and alignment with organizational culture, beyond just technical skills. The study's methodology underscores the potential of AI in providing a more holistic view of candidates, aiding in creating well-rounded teams that align with the organization's values and goals.

### 3.3 Case study-3: Employability based on grit, the big five, and other factors

The paper by Butz et al. (2019) explores how employers weigh various factors when making hiring decisions. The primary focus is on the concept of 'grit,' which includes a job seeker's passion and perseverance for long-term goals, and how it is valued in comparison to other attributes. [2]

**3.3.1 How study was approached?** The study engaged a specific group of central Wisconsin employers, all members of the Portage Wood and Marathon County Business Councils. A total of 983 businesses were invited via email to participate in an online survey, which lasted 14 weeks from February to May 2018. The survey included 54 items and took an average of 17 minutes to complete. Out of 115 employers who started the survey, 100 completed it. The respondents represented a diverse range of job titles and organizational sizes, with the oldest organization founded in 1817 and the newest in 2017. The median size of the surveyed organizations was 150 employees, with a range from 1 to 500,000 employees. The respondents were classified into various industries and sectors based on their self-identified NAICS classification. [2]

**3.3.2 Techniques used for surveys and analyzing results** The study employed several techniques for preparing and analyzing results,

- **Measuring Grit:** The study measured how much hiring managers valued grit in job seekers using a 12-item scale adapted from Duckworth et al. (2007). This scale was divided into two subscales: grit interest and grit effort, which were then combined into a composite index.
- **Assessing the Big Five Personality Traits:** The Big Five personality traits were measured using a 10-item measure (BFI-10). Participants evaluated key phrases related to well-qualified applicants for each personality trait. The phrases varied for each subscale, measuring traits like openness, neuroticism, conscientiousness, agreeableness (reverse coded), and extraversion (reverse coded). The reliability of this measure was confirmed using the expanded form of the Spearman-Brown formula.
- **University-Provided Skills and Experiences:** The survey asked employers about the importance of various university-provided skills and experiences for job applicants. These included GPA, written and oral communication skills, problem-solving, technology skills, teamwork, leadership, volunteer experience, internship experience, and previous work experience. Employers rated these on a 5-point scale from 'not at all important' to 'extremely important' as shown in below table 1.



**Table 1.** Adapted grit scale - A well-qualified applicant for an entry-level position at your organisation.

Item content	Sub-scale	Item number
...has overcome setbacks to conquer an important challenge.	Effort	1
...is not discouraged by setbacks.	Effort	4
...is a hard worker.	Effort	6
...finishes whatever he/she begins.	Effort	9
...has achieved a goal that took years of work.	Effort	10
...is diligent.	Effort	12
...have new ideas and projects that sometimes distract him/her from previous ones. R	Interest	2
...have interests change from year to year. R	Interest	3
...has been obsessed with a certain idea or project for a short time but later lost interest. R	Interest	5
...often sets a goal but later chooses to pursue a different one. R	Interest	7
...has difficulty maintaining focus on projects that take more than a few months to complete. R	Interest	8
...becomes interested in new pursuits every few months. R	Interest	11
Notes: Response options were: 1. not like a qualified job candidate 2. not much like a qualified job candidate 3. somewhat like a qualified job candidate 4. mostly like a qualified job candidate 5. very much like a qualified job candidate Statements ending in an italic 'R' signify negatively worded items, that were reverse coded before creating the composite scales.		

## 4 Results obtained

### 4.1 UNDER WHAT CONDITIONS

### 4.2 CONSTRAINTS

### 4.3 Quality -adequate? Subpar? And why?

## 5 Conclusion and future works

### 5.1 Suggested improvements

### 5.2 Limitations to solution

### 5.3 Applications in real world

### 5.4 Conclusion

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