

1. A Practical Guide for Technical Recruiting

Let's get started.

Introduction

This guide will help you navigate the complex world of data engineering, AI engineering, automation, and no-code development when sourcing candidates. The tech landscape changes fast, and finding the right talent requires understanding the nuances of each role and how to spot true expertise.

This framework isn't about rigid rules - it's a flexible tool to help you recognize quality candidates and understand what makes someone junior, mid-level, or senior in these specialized fields.

Why does this matter?

Data engineering and AI roles are some of the most misunderstood positions in tech recruiting. Here's why getting this right is crucial:

- **High business impact:** These roles directly influence an organization's ability to leverage data for competitive advantage
- **Costly mis-hires:** The wrong hire can derail projects and cost hundreds of thousands in lost productivity
- **Technical complexity:** These fields combine multiple disciplines that require nuanced understanding
- **Evolving landscape:** Tools and approaches change rapidly, making it hard to know what's current
- **Candidate shortages:** The best candidates get multiple offers, so we need to move quickly and confidently

Education requirements assessment.

Minimum Requirements vs. Practical Equivalents

In technical fields like data engineering and AI, formal education requirements have evolved significantly. Here's a practical guide:

Traditional Path (Degree-Based)

- Bachelor's in Computer Science, Informatics, Data Science, Mathematics, or related field
- For AI roles, often prefer Master's or PhD in AI, Machine Learning, or Statistics

Practical Equivalents (Experience-Based)

- Bootcamp certification + 1-2 years practical experience
- Self-taught + substantial portfolio of projects (GitHub, etc.)
- Adjacent technical degree + significant upskilling
- Industry certifications + practical implementation experience

Why this matters: The best candidates don't always follow traditional paths. Focus on demonstrated ability rather than credentials alone. Someone with a psychology degree who self-taught Python and built several data pipelines may be more effective than a CS graduate with no practical experience.

 Pro Move

When evaluating non-traditional candidates, look for:

- Evidence of continuous learning
- Substantial project portfolio
- Contributions to open source or community
- Technical writing or speaking that demonstrates expertise
- Strong problem-solving capabilities demonstrated through work examples

Thoughts from a CTO.

Things To Consider

Remember, the best candidates aren't always the ones with the most impressive-sounding titles or the longest list of technologies on their profile. Look for people who can articulate their specific contributions, demonstrate problem-solving abilities, and show a passion for continuous learning.

A great data or AI engineer doesn't just *know* the tools—they understand how to apply them to solve real business problems. They can communicate complex technical concepts clearly and collaborate effectively with diverse teams.

By focusing on demonstrated abilities rather than just keywords, you'll identify candidates who can truly drive value for your organization, not just those who are good at crafting their LinkedIn profiles.