

Curriculum gaps and Transition Barriers: Evidence for Equitable Workforce Policy in the AI era

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What do we mean by “equity” in the AI era?

- Ensuring that all graduates, regardless of institution or background, have equal opportunity to participate in AI-augmented work.
- Curriculum design must reflect inclusion in emerging AI driven sectors, not just employability in traditional ones.
- Policy and institutional alignment should promote fair pathways for upskilling and reskilling across all universities.
- The goal : equal readiness, not identical curricula

Why this matters: The policy context

The changing landscape:

- Artificial intelligence is transforming labour markets faster than education systems can adapt.
- Most national frameworks lack clear metrics for AI-readiness or skill alignment.
- Emerging evidence shows widening digital and data literacy divides across institutions.

LSE Change Makers Pilot study

- This study was conducted as part of the LSE Change Makers research initiative (2024-2025).
- Our aim was to understand how experiential learning prepares students for AI-augmented work environments.
- This study focused on alumni of the Department of Statistics at LSE.

Approach:

- Combined a curriculum–industry skill mapping with a student perception survey.



Developed two indicative measures:

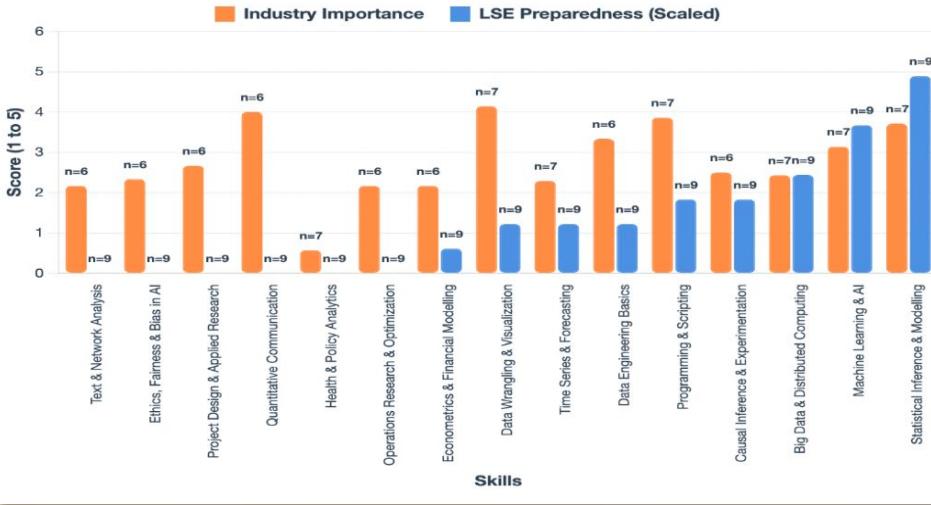
- Skill Coverage Gap Score (SCGS)
- Perceived Readiness Gap Score (PRGS)



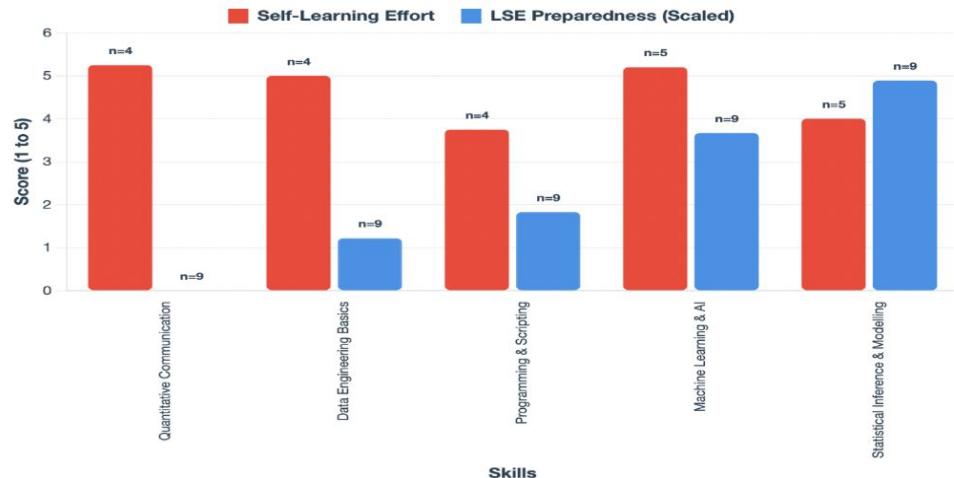
Key takeaways from Change Makers



SCGS From MSc Statistics



PRGS From MSc Statistics



Expansion - The research questions

*If even elite institutions show measurable skill gaps,
how can national policy ensure equitable
AI transitions for all?*

Our research questions:

- RQ1: Which curriculum-to-occupation skill gaps create the greatest barriers to entering high demand AI augmented roles?
- RQ2: What skill overlap patterns enable feasible transitions from declining high AI exposure occupations into growing roles?

The Methodology



1. Data Integration (All Public Sources)

- O*NET 30.0: 35 standardized skills, 668 occupations
(e.g., *Critical Thinking, Programming, Coordination, Service Orientation*)
- Felten AIOE Index: AI automation exposure by occupation
Range: $-2.67 \rightarrow +1.58$ (High = more exposed)
- BLS Employment Projections (2024–2034):
1,113 occupations · Growth range: -36% $\rightarrow +50\%$

2. Analytical Method

- Jaccard Similarity Coefficient:

$$J(A, B) = \frac{|S_A \cap S_B|}{|S_A \cup S_B|}$$

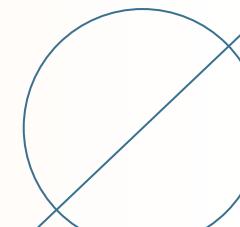
If Job A & Job B share 6 of 10 skills $\rightarrow J = 0.60$

- Built a binary skill matrix (668×35):
1 = required · 0 = not required
- Published, standard method in labour-market research

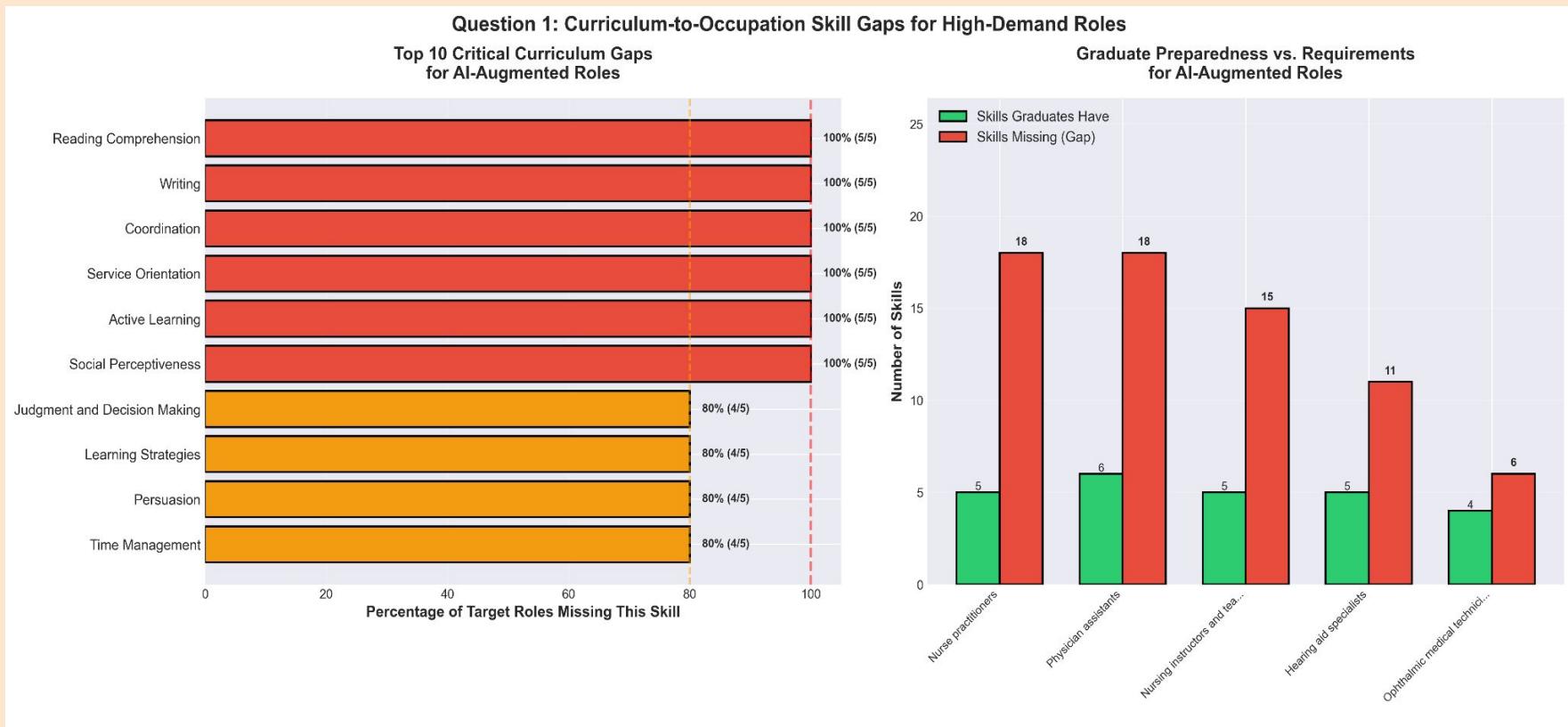
3. Scope

Q1: 5 AI-augmented target roles

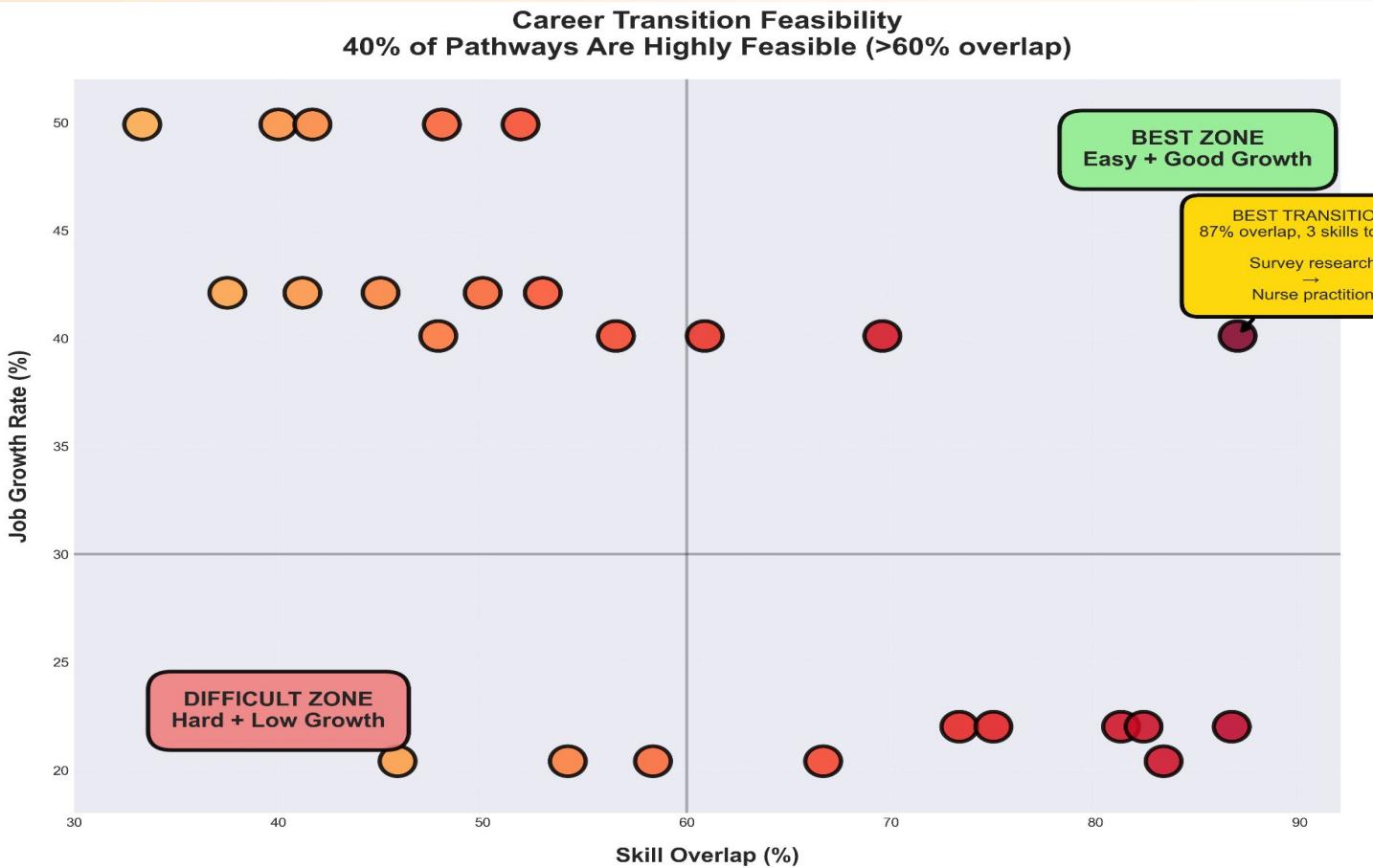
Q2: 25 career transitions (5 origins \times 5 destinations)



RQ1 : Findings

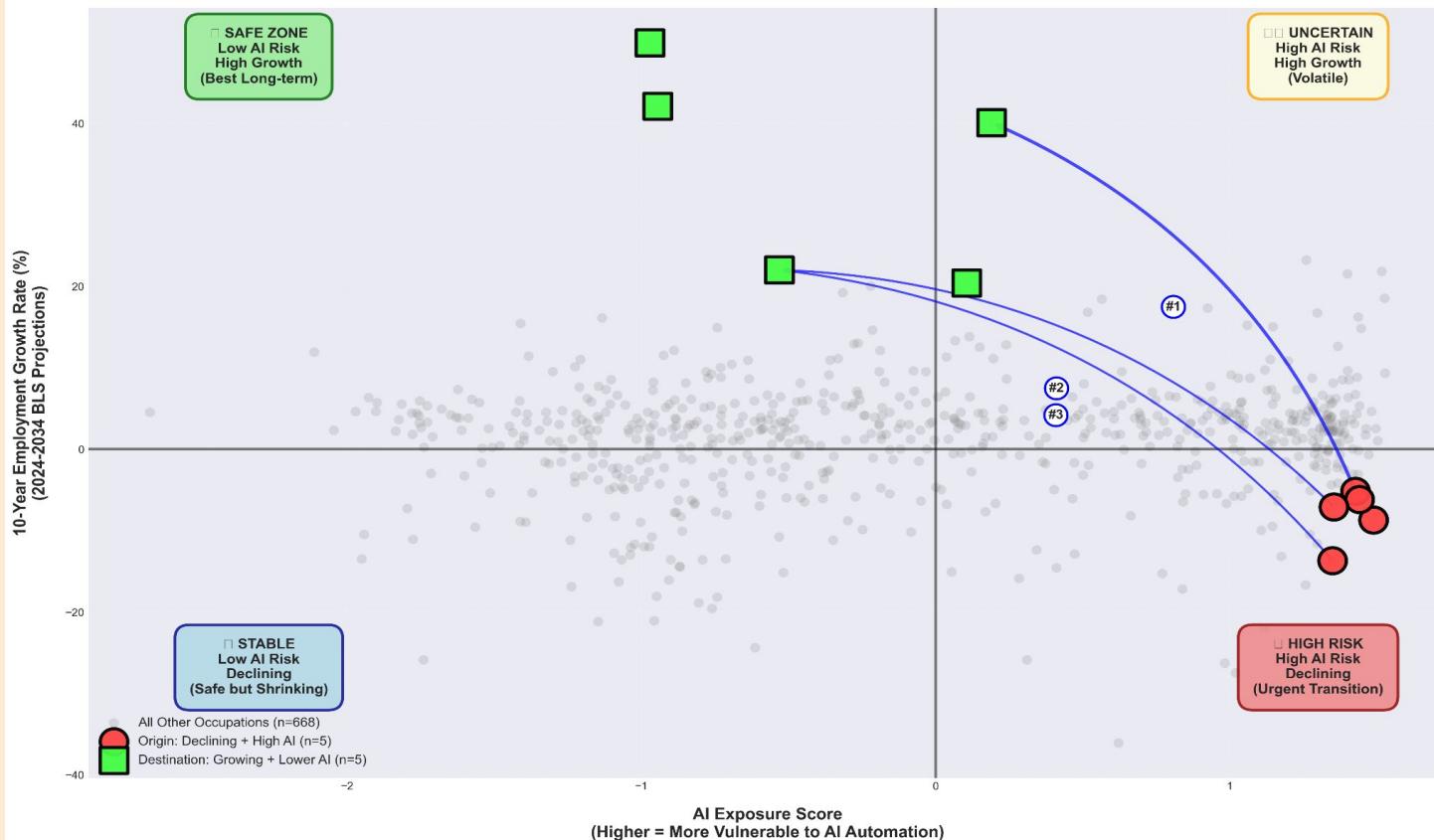


RQ2: Findings



Full Occupational Landscape

Occupational Vulnerability Landscape in the AI Era
AI Exposure vs. Employment Growth with Transition Pathways



Policy Recommendations

1. Curriculum Modernization *(from RQ1)*

Action: Integrate 6 critical skills into *all* university programs

- Reading Comprehension · Writing · Coordination
- Service Orientation · Active Learning · Social Perceptiveness

Goal: Ensure equitable skill access across institutions

Impact: Closes 13.6-skill average gap for AI-augmented roles

2. Targeted Reskilling Programs *(from RQ2)*

Action: Create 6–12 month certificates teaching transition-critical skills

- Operations Monitoring · Instructing · Systems Analysis
- Systems Evaluation · Persuasion

Goal: Enable 10 high-feasibility career pathways

Impact: Median 7-skill gap is achievable through focused training

3. Data-Driven Quality Assurance

Action: Set quantitative skill benchmarks & track graduate outcomes

- Identify under-resourced programs
- Measure progress in AI-readiness

Goal: Continuous improvement & accountability across systems

Validation & Limitations

What This Proof-of-Concept Demonstrates

- Public datasets can be systematically integrated (replicable)
- Skill gaps can be precisely quantified (specific, actionable)
- Transition feasibility can be objectively measured (quantitative)
- Evidence-based workforce policy is achievable (validated approach; Jaccard)

Key Limitations

- RQ1 Proxy: Used *entry-level occupations* as a stand-in for graduate capabilities → *Course-Skill Atlas* needed for direct curriculum mapping (*Phase 2 priority*)
- Sample Size: Proof-of-concept scale → Q1: 5 roles, Q2: 25 pathways → Scalable to 200+ with resources
- Geographic Scope: Based on *U.S. datasets*; International replication required
- Skill ≠ Full Feasibility: High overlap doesn't capture *licensing, credentials, or practical barriers*

Conceptual framework

Department-Level Skill Mapping

- Collect course-specific skill data from each university department
- Identify actual skills students gain across diverse programs

National Integration

- Link skills to domestic labour market data (jobs, sectors, growth rates)
- Apply Jaccard similarity for precise skill-overlap and transition analysis

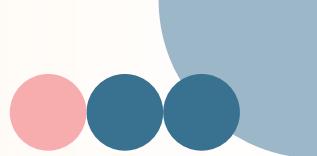
Curriculum-to-Occupation Alignment

- Compare extracted course skills with *national occupation datasets*
- Measure *real curriculum–industry gaps* across institutions

Policy Application

- Use results to design *curricula and reskilling pathways* that:
 - Equip graduates with *future-proof skills*
 - Minimize *automation risk*
 - Promote *equitable workforce readiness*

Next steps & Scaling



Phase 1 — Completed

- Methodology validated across 3 integrated datasets
- 668 occupations analyzed
- 6 curriculum gaps identified (Q1)
- 10 transition pathways quantified (Q2)
- Proof-of-concept: *public data enables precise, evidence-based analysis*

Phase 2 — Planned (6–12 months)

- ➡ Integrate Course-Skill Atlas
 - Analyze *actual university curricula* (not proxies)
 - Compare elite vs. non-elite institutions → answer Q1 fully
- ➡ Scale Occupation Analysis
 - Expand to 200+ occupation pairs
 - Cover *multiple sectors* (tech, business, healthcare, education)
 - Add *international comparisons* (UK, EU, Asia)

Phase 3 — Implementation (12–24 months)

➡ Tool Development

- *Transition Pathway Explorer* for career counselors
- *Curriculum Gap Dashboard* for institutions

➡ Validation & Piloting

- *Employer and educator interviews*
- *Pilot reskilling program* with outcome tracking
- *Policy briefs* for government agencies

Conclusion

Substantial Skill Gaps Exist

- **13.6-skill average gap for AI-augmented roles**
- **Only 29 % graduate preparedness**
- **6 critical skills systematically missing from curricula**

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Feasible Transition Pathways Exist

- **40 % of analyzed transitions are *highly feasible* (> 60 % overlap)**
- **7-skill median reskilling requirement**
- **Specific, quantified career pathways identified**

Evidence-Based Policy Is Achievable

- **Public datasets enable *precise, replicable* analysis**
- **Jaccard similarity quantifies *transition***

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Thank you!