

# **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:

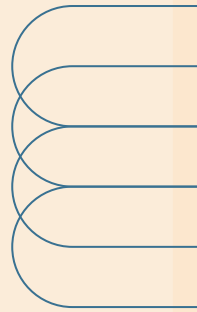
- Artificial intelligence is transforming labour markets faster than education systems can adapt.
- Most national frameworks lack clear metrics for AI-readiness or skill alignment.
- Emerging evidence shows widening digital and data literacy divides across institutions.

# LSE Change Makers Pilot study

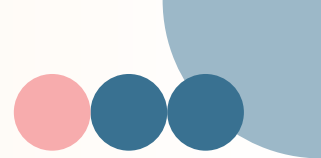
- This study was conducted as part of the LSE Change Makers research initiative (2024-2025).
- Our aim was to understand how experiential learning prepares students for AI-augmented work environments.
- This study focused on alumni of the Department of Statistics at LSE.

## Approach:

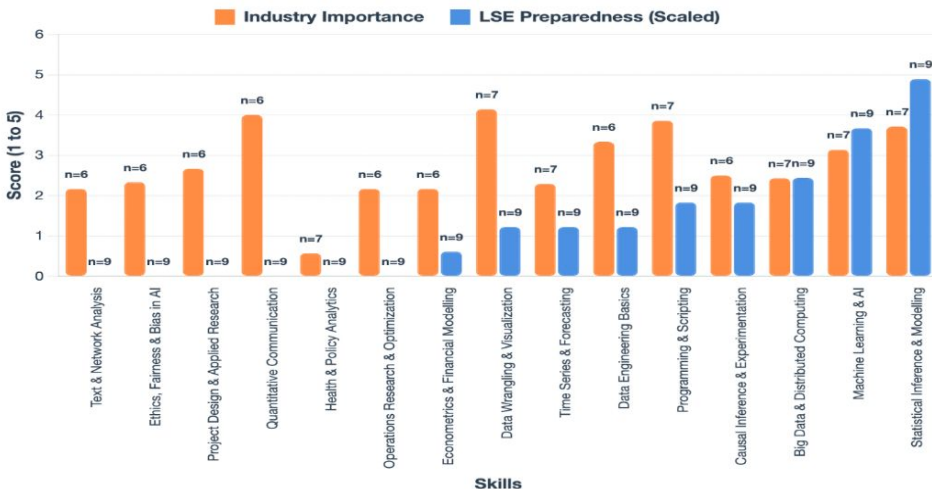
- Combined a curriculum–industry skill mapping with a student perception survey.
- Developed two indicative measures:
  - Skill Coverage Gap Score (SCGS)
  - Perceived Readiness Gap Score (PRGS)



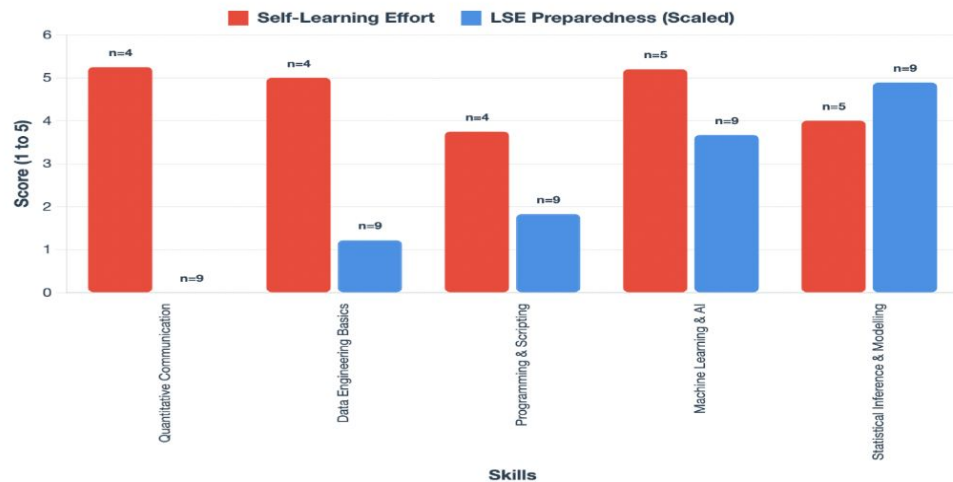
# Key takeaways from Change Makers



## 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 → +1.58 (High = more exposed)*
- BLS Employment Projections (2024–2034):  
*1,113 occupations · Growth range: -36 % → +50 %*

## 2. Analytical Method

- Jaccard Similarity Coefficient:

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

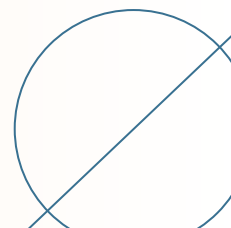
*If Job A & Job B share 6 of 10 skills →  $J = 0.60$*

- 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

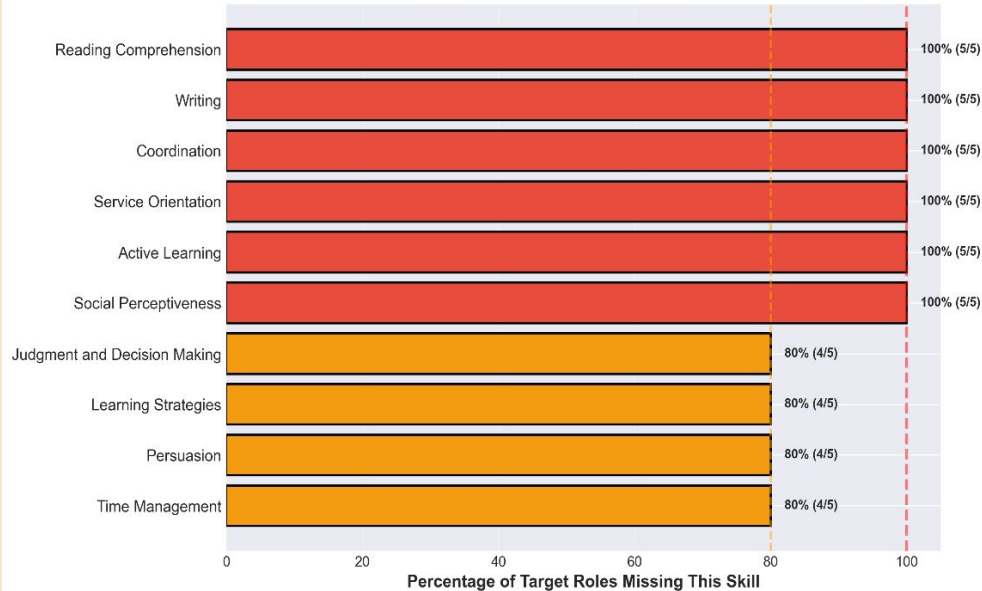
Q2: 25 career transitions (5 origins × 5 destinations)



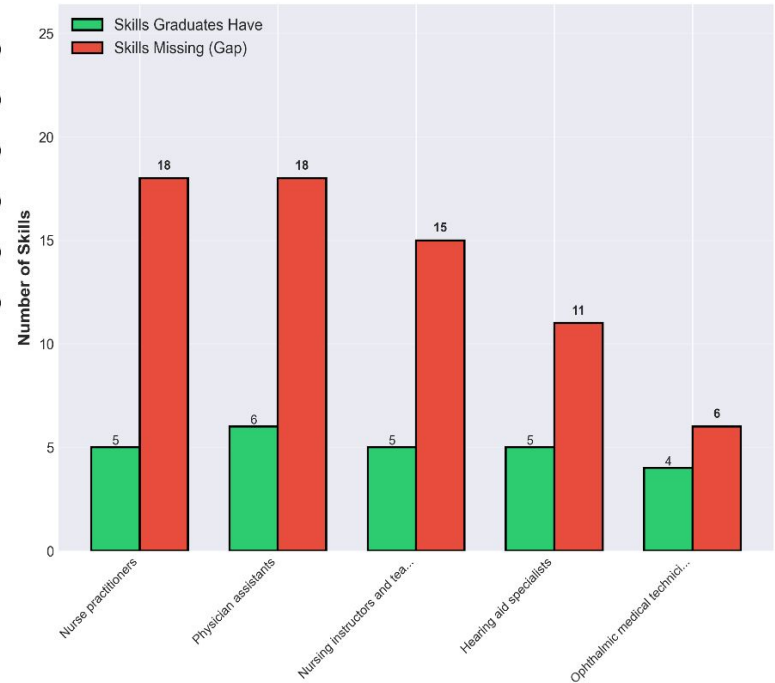
# RQ1 : Findings

## Question 1: Curriculum-to-Occupation Skill Gaps for High-Demand Roles

### Top 10 Critical Curriculum Gaps for AI-Augmented Roles

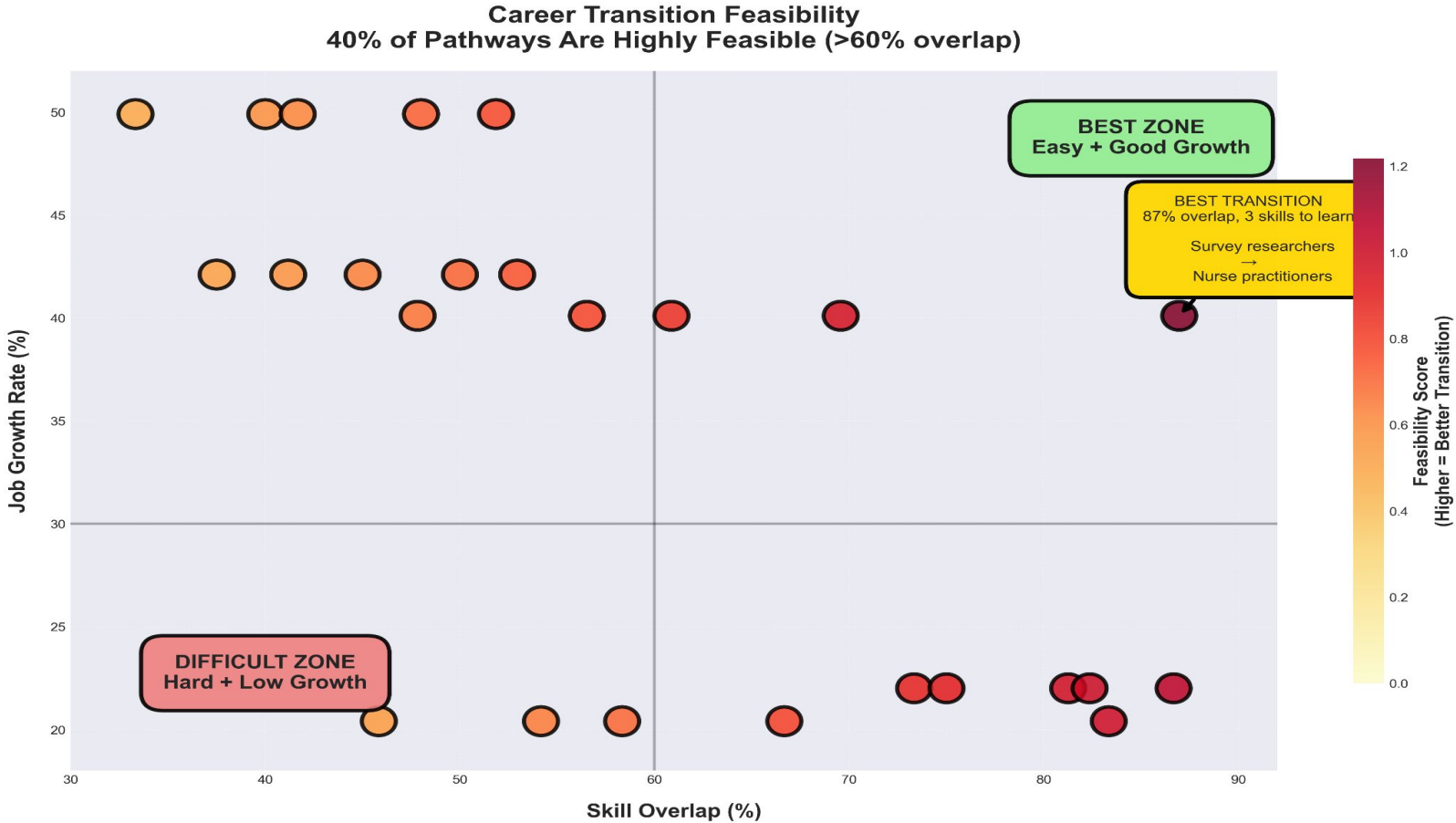


### Graduate Preparedness vs. Requirements for AI-Augmented Roles



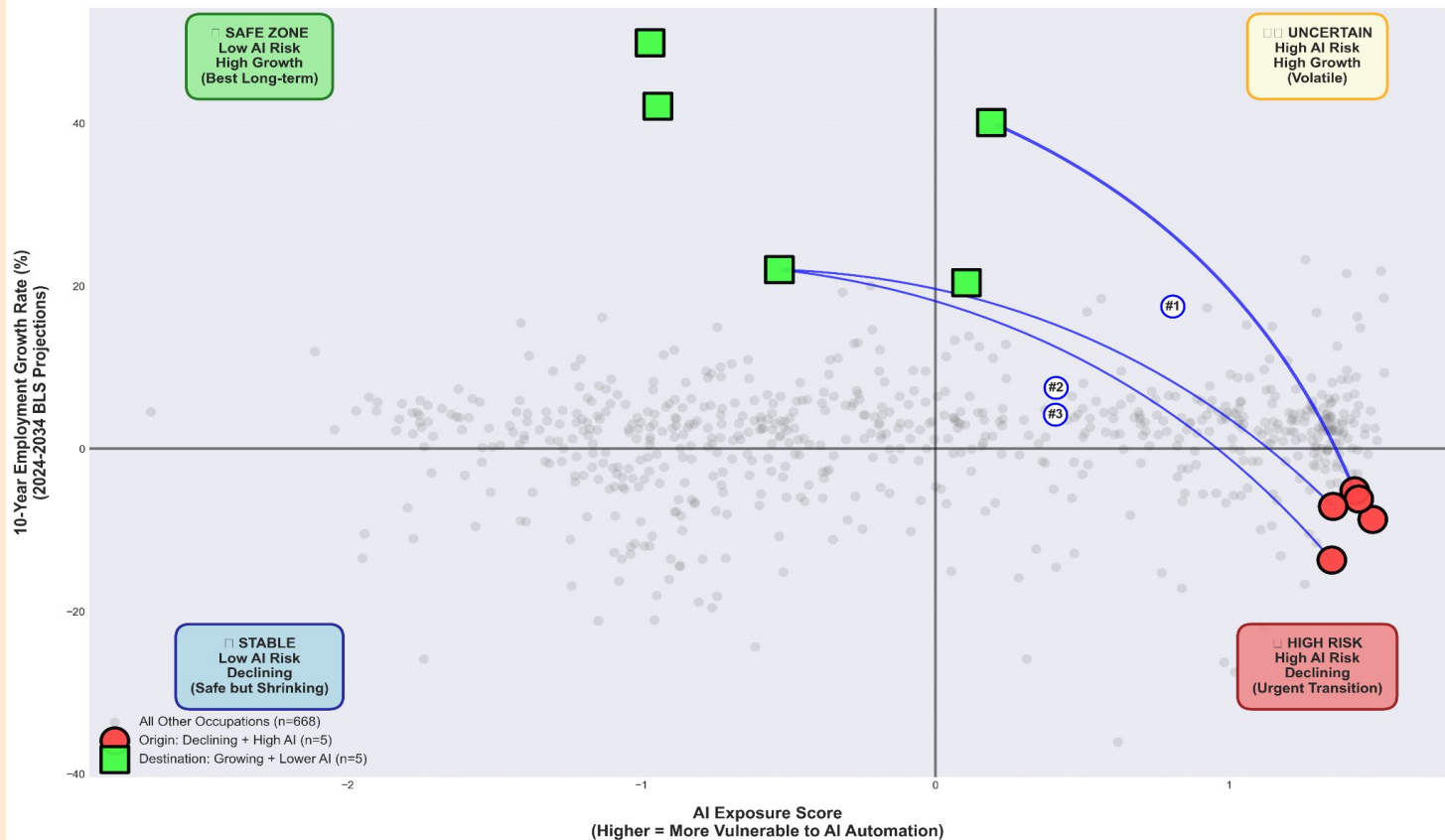


# RQ2: Findings

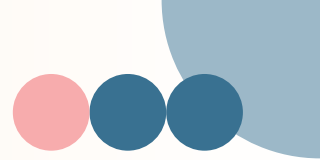


# Full Occupational Landscape

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

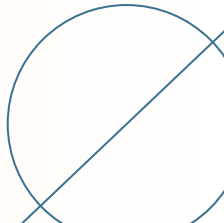
Impact: Median 7-skill gap is achievable through focused training

## 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)

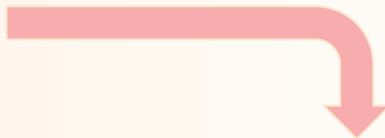
## Key Limitations

- 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

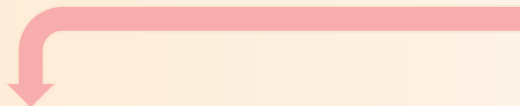
## Department-Level Skill Mapping

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



## Curriculum-to-Occupation Alignment

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



## National Integration

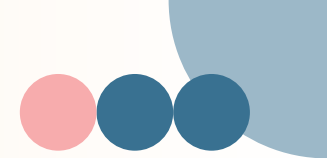
- Link skills to *domestic labour market data* (jobs, sectors, growth rates)
- Apply Jaccard similarity for precise *skill-overlap and transition analysis*



## 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!**