

EDUCATION INVESTMENT AND SKILL MISMATCH: UNDERSTANDING THEIR IMPACT ON EUROPEAN LABOUR MARKETS

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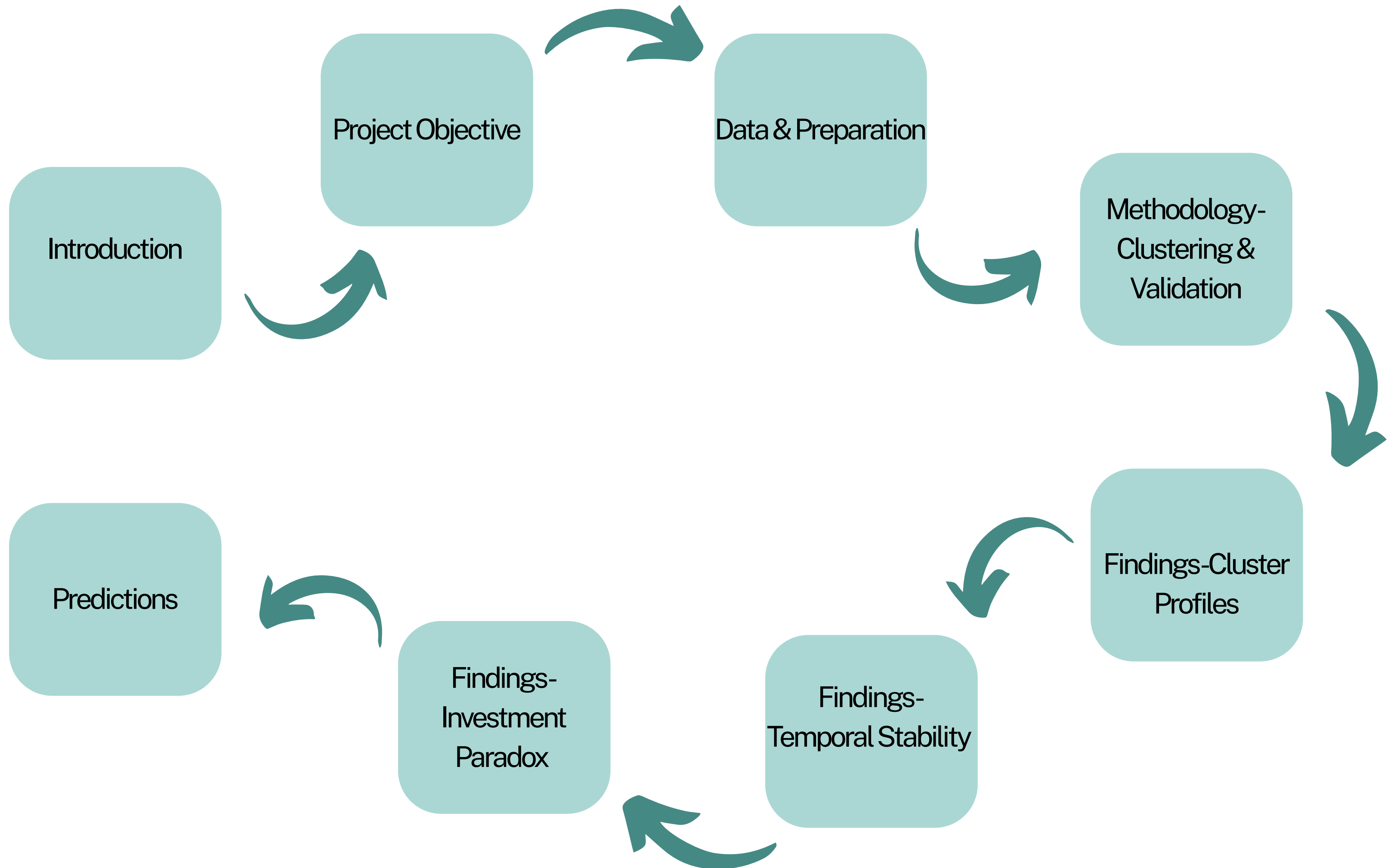
Introduction

Key Points:

- A disconnect often exists between the skills graduates possess and the demands of the labour market.
- This "mismatch" is a significant issue across Europe

○ Consequences:

- For Individuals: Unemployment, overqualification, lower wages.
- For the Economy: Productivity loss, subdued growth potential, social costs.



Goals and Scope of Our Analysis

Scope:

- European Countries (Specify the general set of countries included in the analysis)
- Period: 2012 - 2022
- Data Source: Eurostat
- To conduct an in-depth analysis of education-skill mismatch patterns among European countries.

Specific Goals:

Group countries based on similar mismatch profiles

Examine the evolution of these mismatch patterns over time

Understand the potential impact of policy interventions.

Derive policy implications from the analysis.

Our Data and Preparation Process

- Eurostat as the data source.
- List of key datasets used (e.g., GDP per capita, Employment by education, Overqualification, Youth unemployment, Education expenditure, Educational attainment).

Our Data and Preparation Process

Preprocessing Steps:

Engineering new features for analysis (e.g., Youth Education Paradox Score, Employment Efficiency).

Merging different datasets.

Handling missing values (e.g., Imputation with median for some countries/years).

Quantifying Skill and Labor Market Mismatches

- **Education ROI Indicator:** Measures the return on investment in tertiary education in terms of employment.

- **Youth Education Paradox Score:** Highlights the challenge of youth unemployment even among highly educated individuals.

- **Employment Efficiency Score:** Evaluates how effectively countries translate high educational attainment into employed individuals, while considering over-qualification

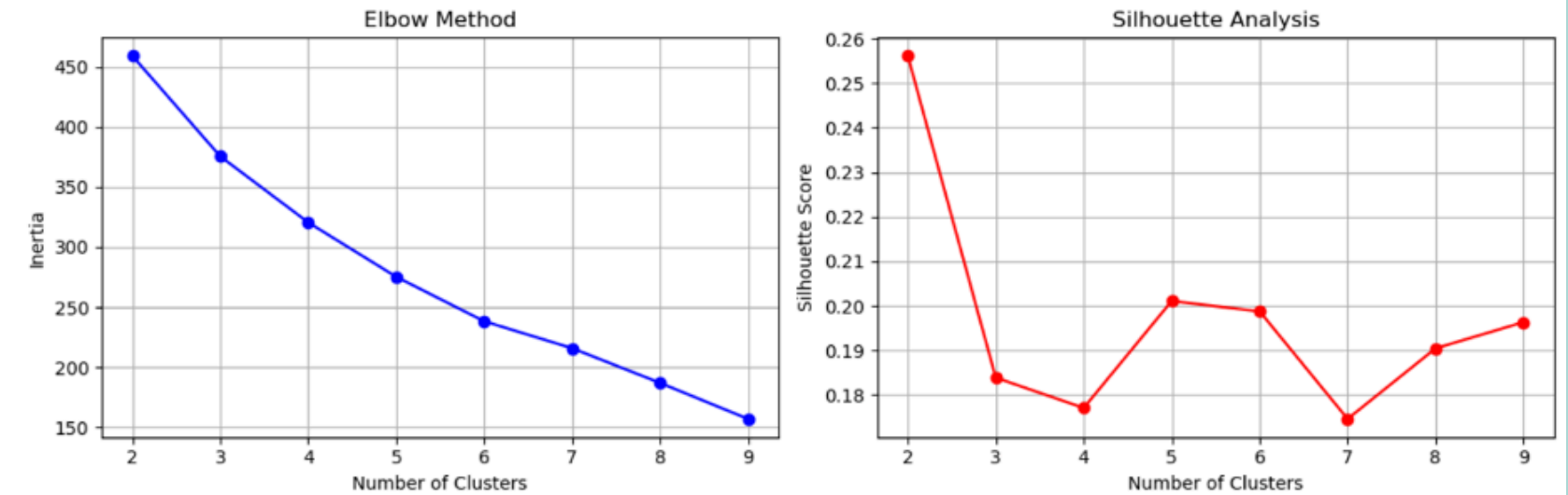
- **Skills Gap Indicator (High-Tech):** Assesses the alignment between the supply of highly educated individuals and the demand in high-tech sectors.

- **Over-qualification Adjusted by Investment:** Combines the rate of over-qualification with the level of tertiary education investment.

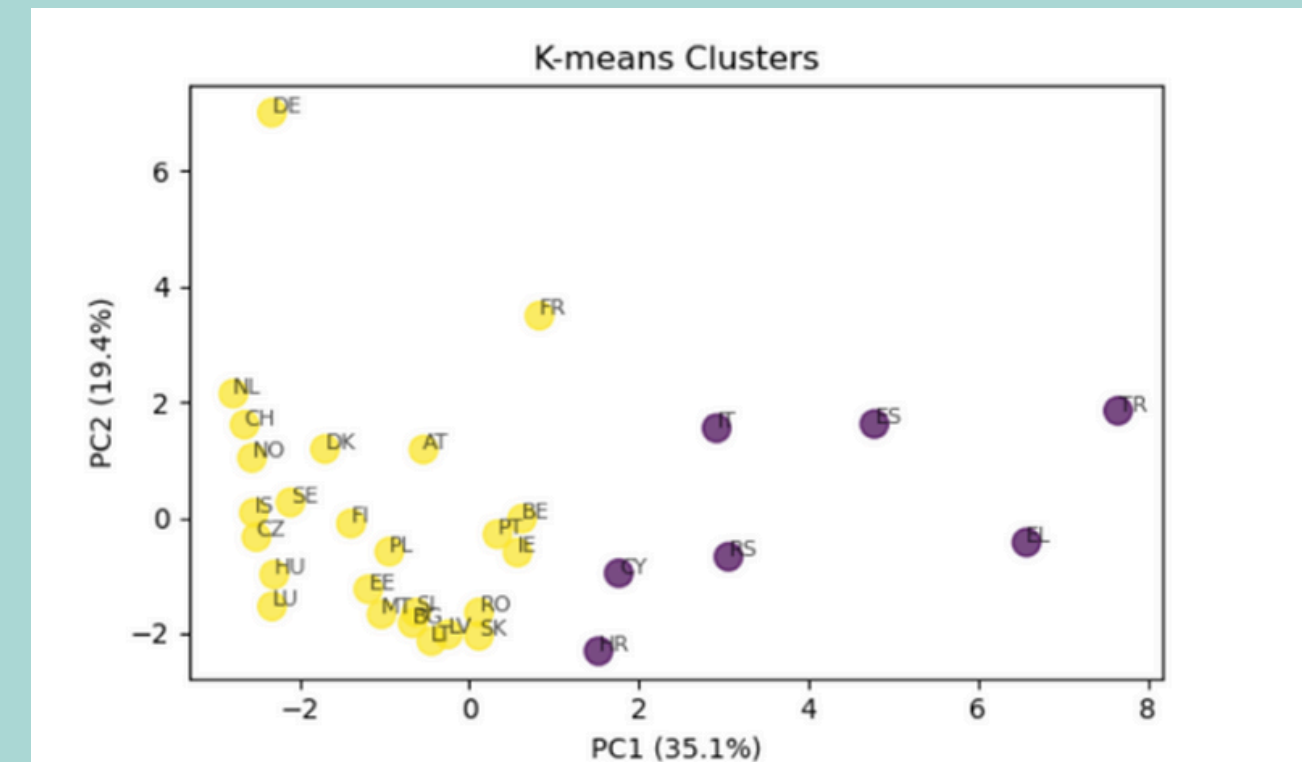
Methodology - Clustering & Validation

- Method Used: K-Means Clustering Algorithm.
- The K-Means clustering algorithm was applied based on 19 input features.
- Various statistical metrics were used, including Silhouette Score, Elbow Method.

Determining optimal number of clusters...



Optimal number of clusters (by silhouette score): 2



Standar scaling

**Dimensionality Reduction
(PCA):**

ENHANCING OUR CLUSTERING STRATEGY:

This process led to the identification of 4 distinct clusters, representing different country archetypes based on their skill and labor market indicators.

Hierarchical Clustering

DBSCAN

Why 4 Clusters?



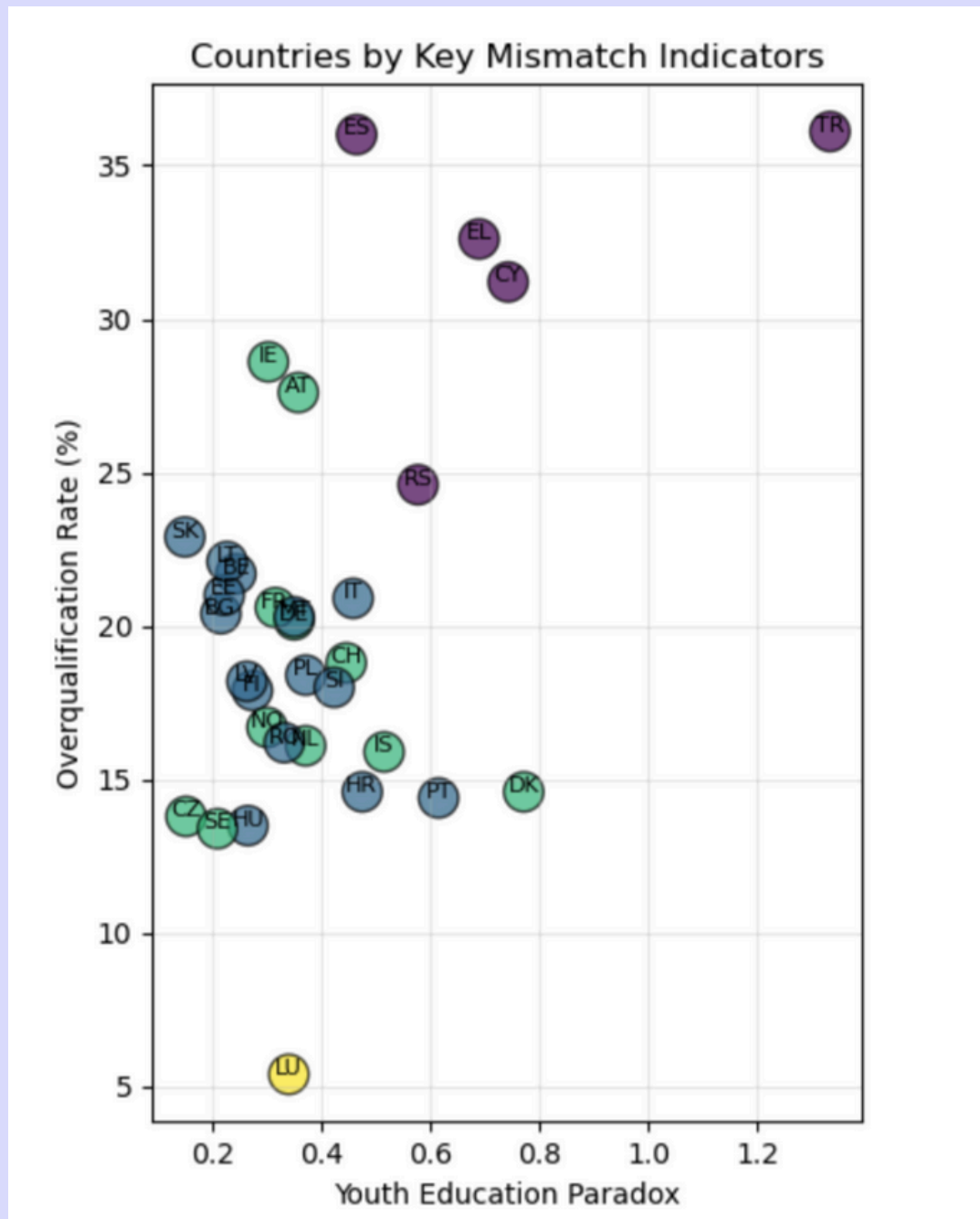
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CLUSTER VALIDATION SUMMARY
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k	Silhouette	Calinski-Harabasz	Davies-Bouldin	Inertia	Composite	Score
3	0.172	8.2	1.578	183.6		0.038
4	0.190	8.2	1.251	153.3		0.220
5	0.170	8.0	1.217	132.2		0.140

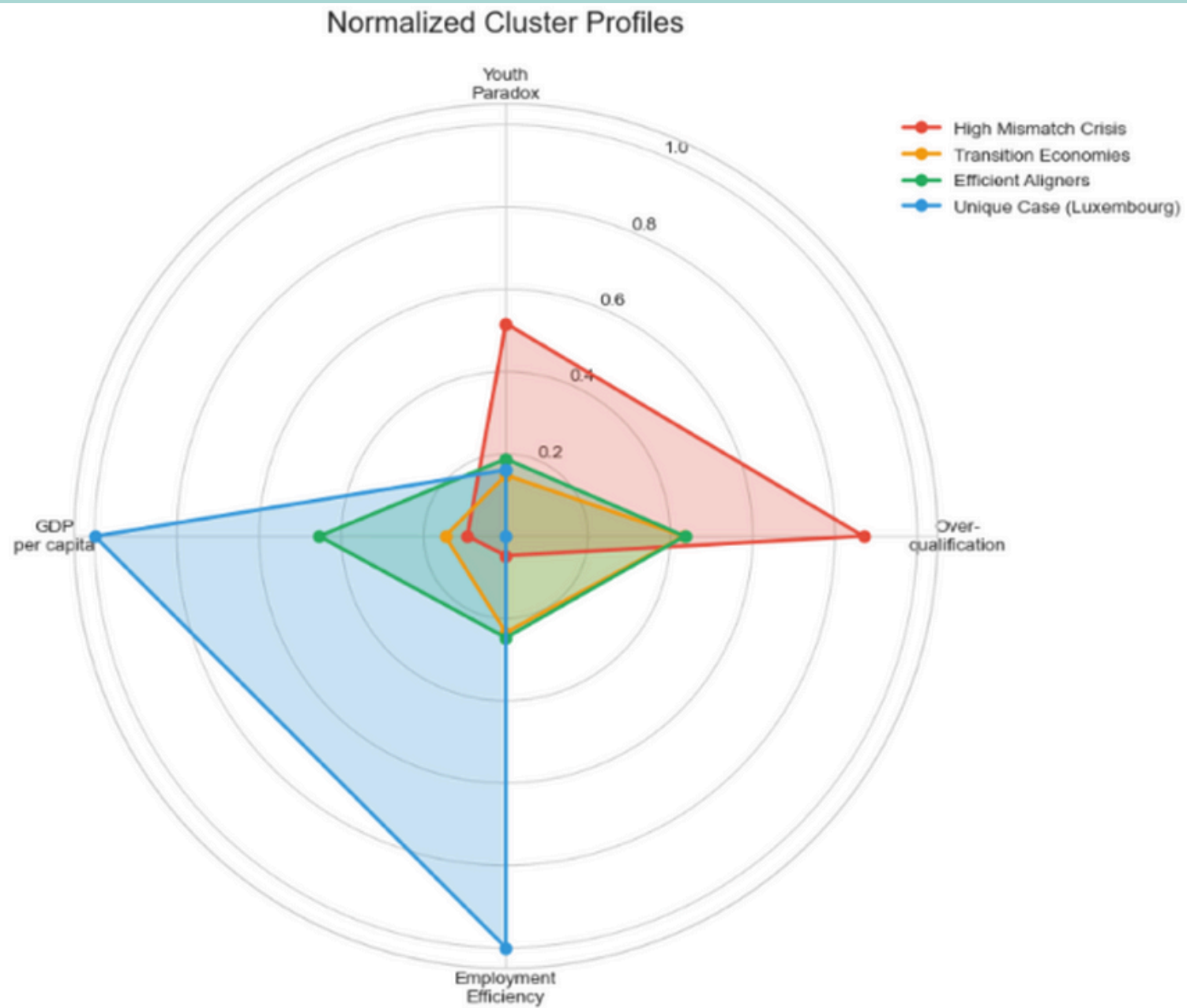
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CONCLUSION: k=4 is optimal because:
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1. Elbow point in inertia curve (mathematical detection)
2. Local maximum in silhouette score
3. Good balance between cluster separation and cohesion
4. Highest composite score combining all metrics
5. Meaningful interpretation (4 distinct economic patterns)
6. Stable cluster assignments across different initializations

Findings - Cluster Profiles

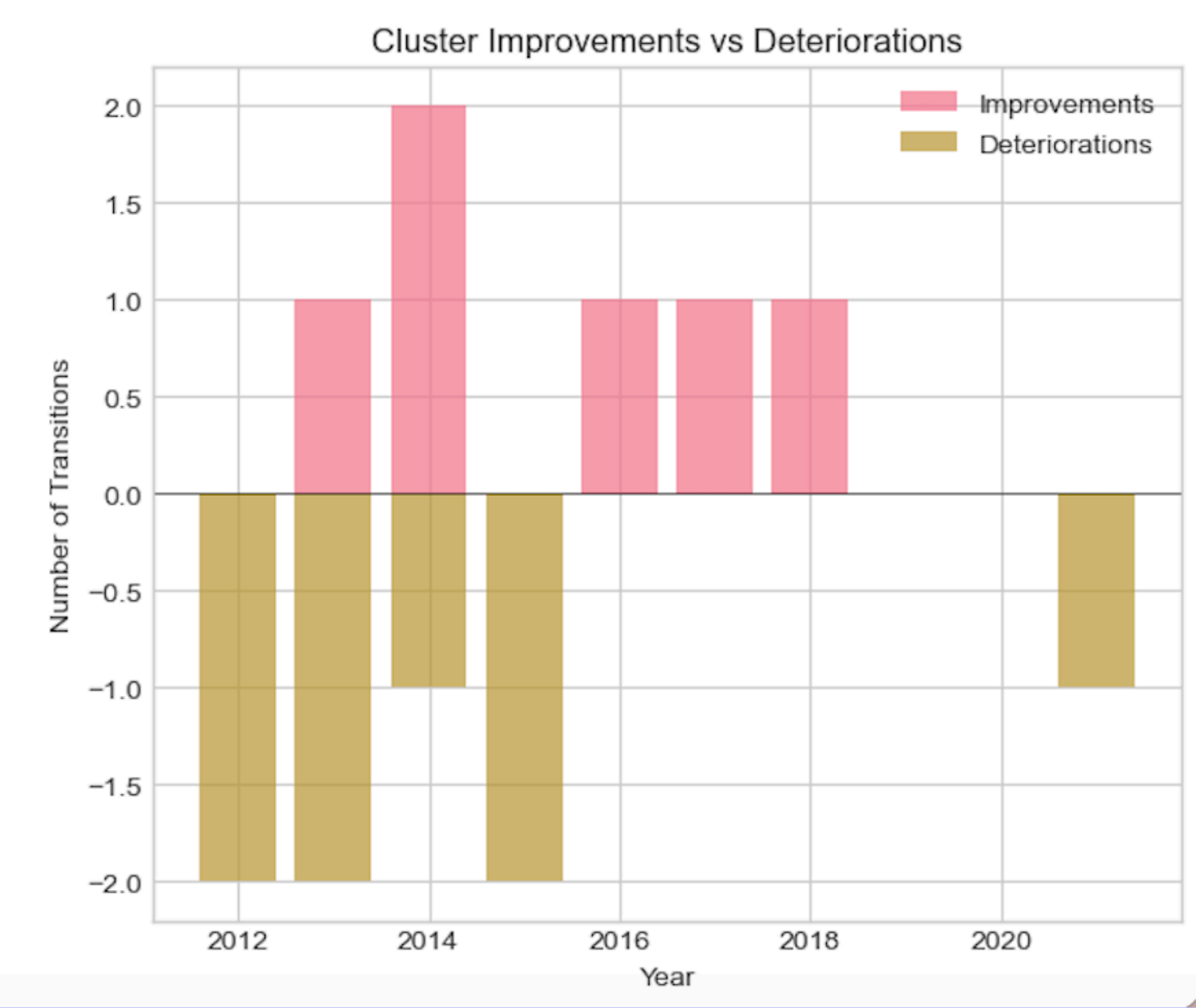


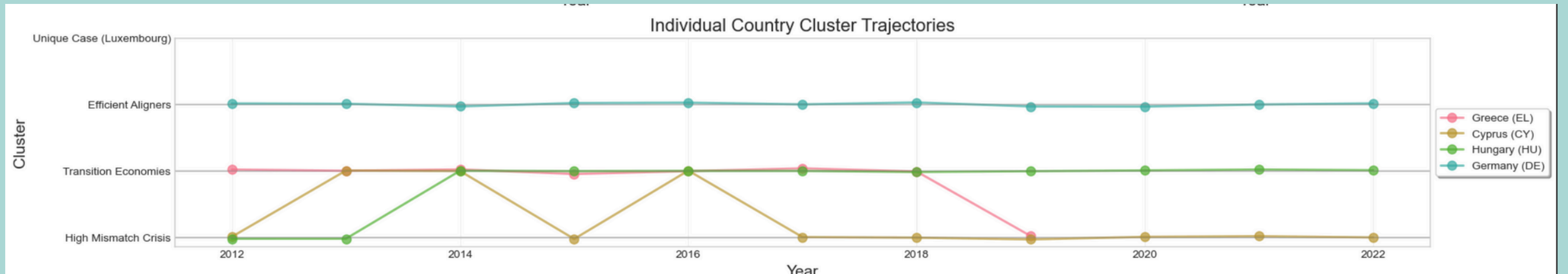
- Brief Profile of Each Cluster:
 - **Cluster 0:** High Mismatch Crisis: High overqualification, pronounced youth paradox. (Example countries: ES, EL, TR, CY)
 - **Cluster 1:** Transition Economies: Moderate mismatch, lower education investment. (Example countries: PL, RO, HU, HR)
 - **Cluster 2:** Efficient Aligners: Strong economies, good education-labour market matching, low mismatch. (Example countries: DE, FR, NL, SE)
 - **Cluster 3:** Unique Case (Luxembourg): Very high GDP, distinct profile.



Cluster Summary Statistics

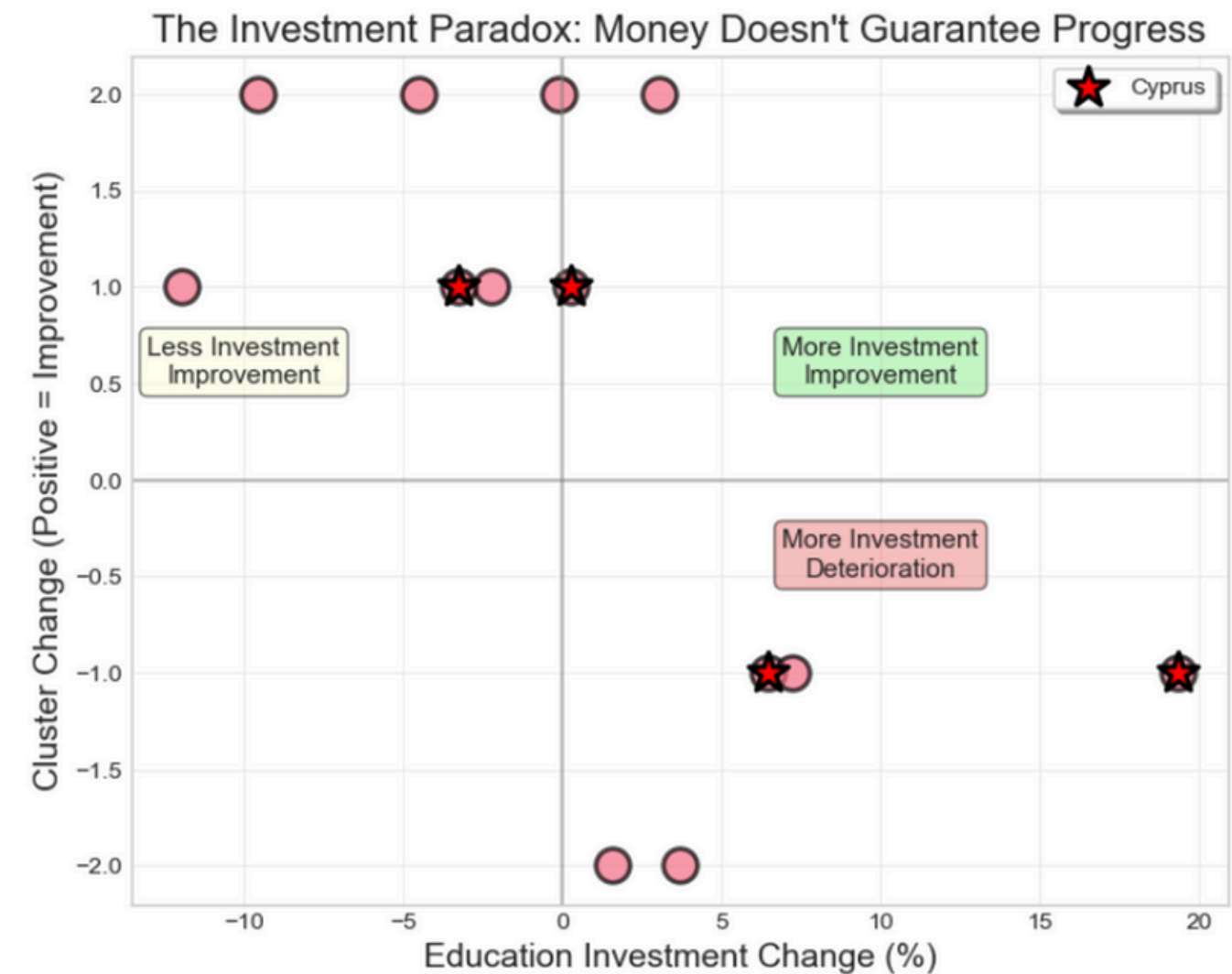
Cluster	N	Youth Paradox	Overqual Rate	Emp. Efficiency	Countries
High Mismatch Crisis	5	0.76	32.1%	2.4	CY, EL, ES, RS, TR
Transition Economies	15	0.32	18.7%	4.5	BE, BG, EE, FI, HR, HU, IT, LT, LV, MT, PL
Efficient Aligners	11	0.37	18.8%	4.7	AT, CH, CZ, DE, DK, FR, IE, IS, NL, NO, S
Unique Case (Luxembourg)	1	0.34	5.4%	13.1	LU





The Investment Paradox

- **Finding:** Our data shows increasing education investment doesn't always lead to better skill matching.
- You can invest more money, but the mismatch situation might not improve, or could even get worse.



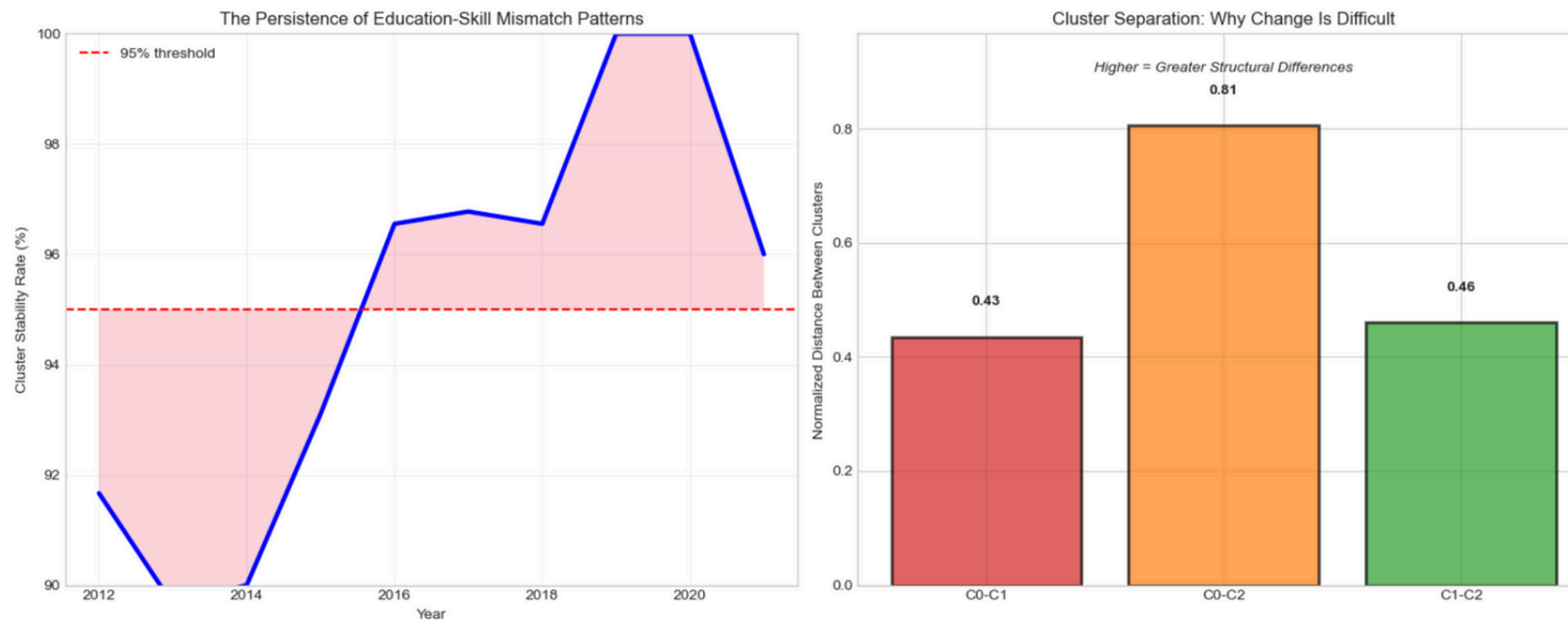
Forecasting the Future and Simulating Policy Impact

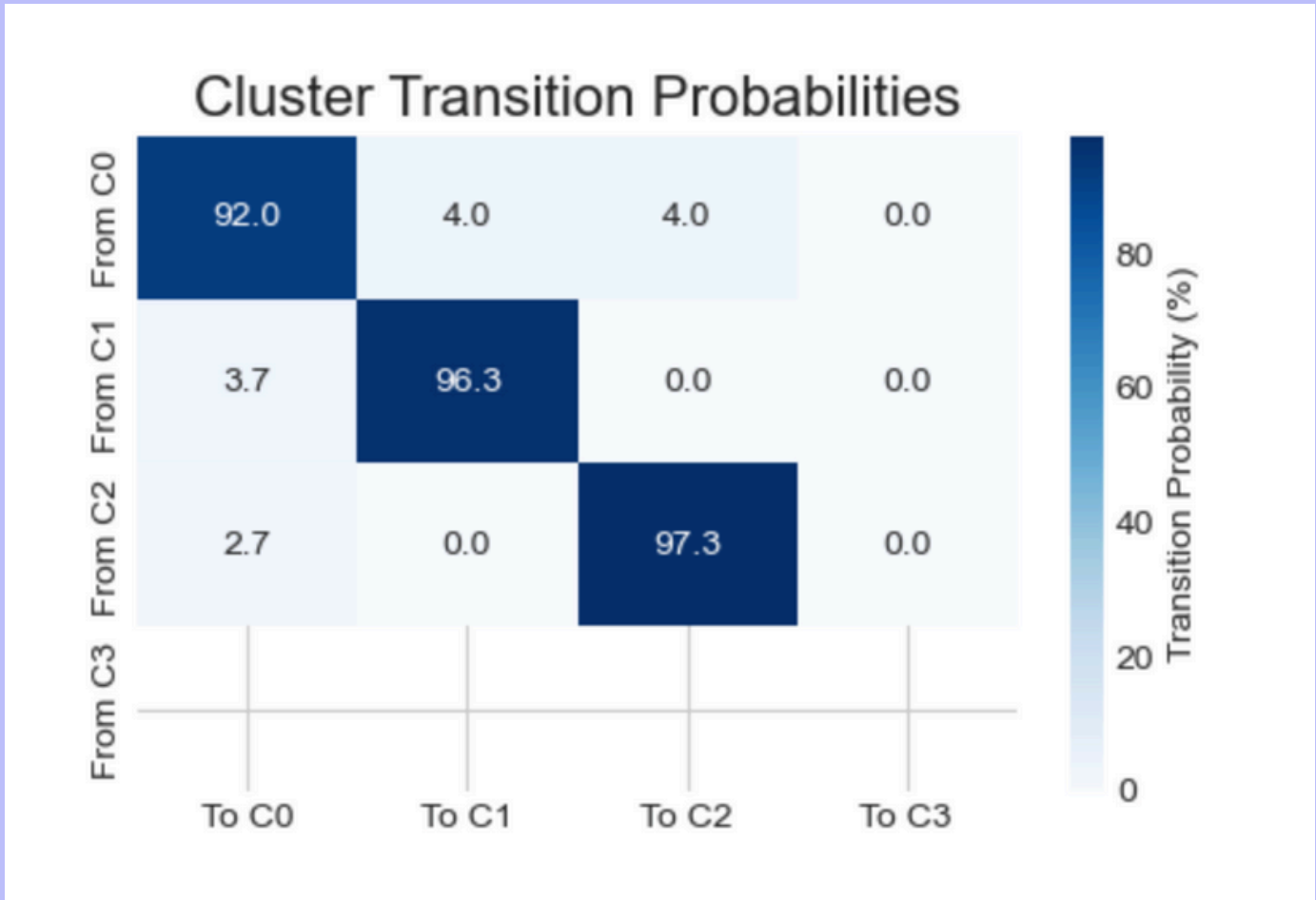
A model (**Random Forest**) was trained on historical data to predict future cluster membership based on a country's latest profile.

- **Finding:** The model predicted that zero countries (out of the 27 analysed) would change their cluster in the next 10 years (based on the latest data).

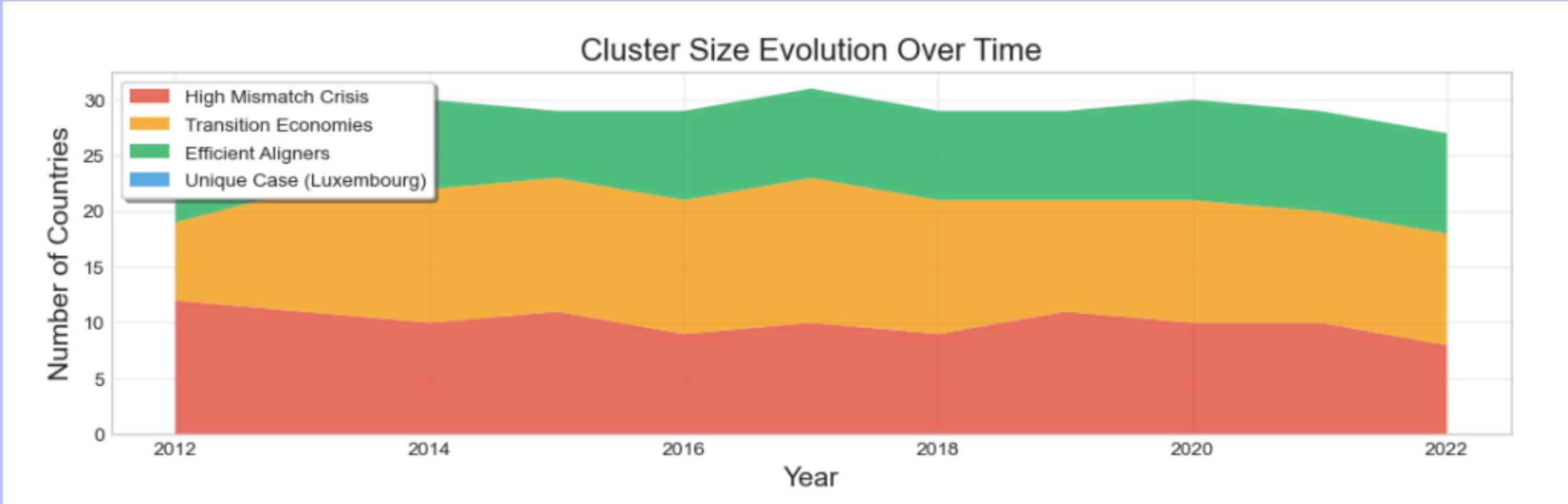
Why Countries Tend to Stay Trapped in Their Clusters

Different mismatch clusters are clearly separated in the space defined by key mismatch metrics (e.g., Overqualification, Youth Paradox)





Finding: Moving between clusters is very rare.

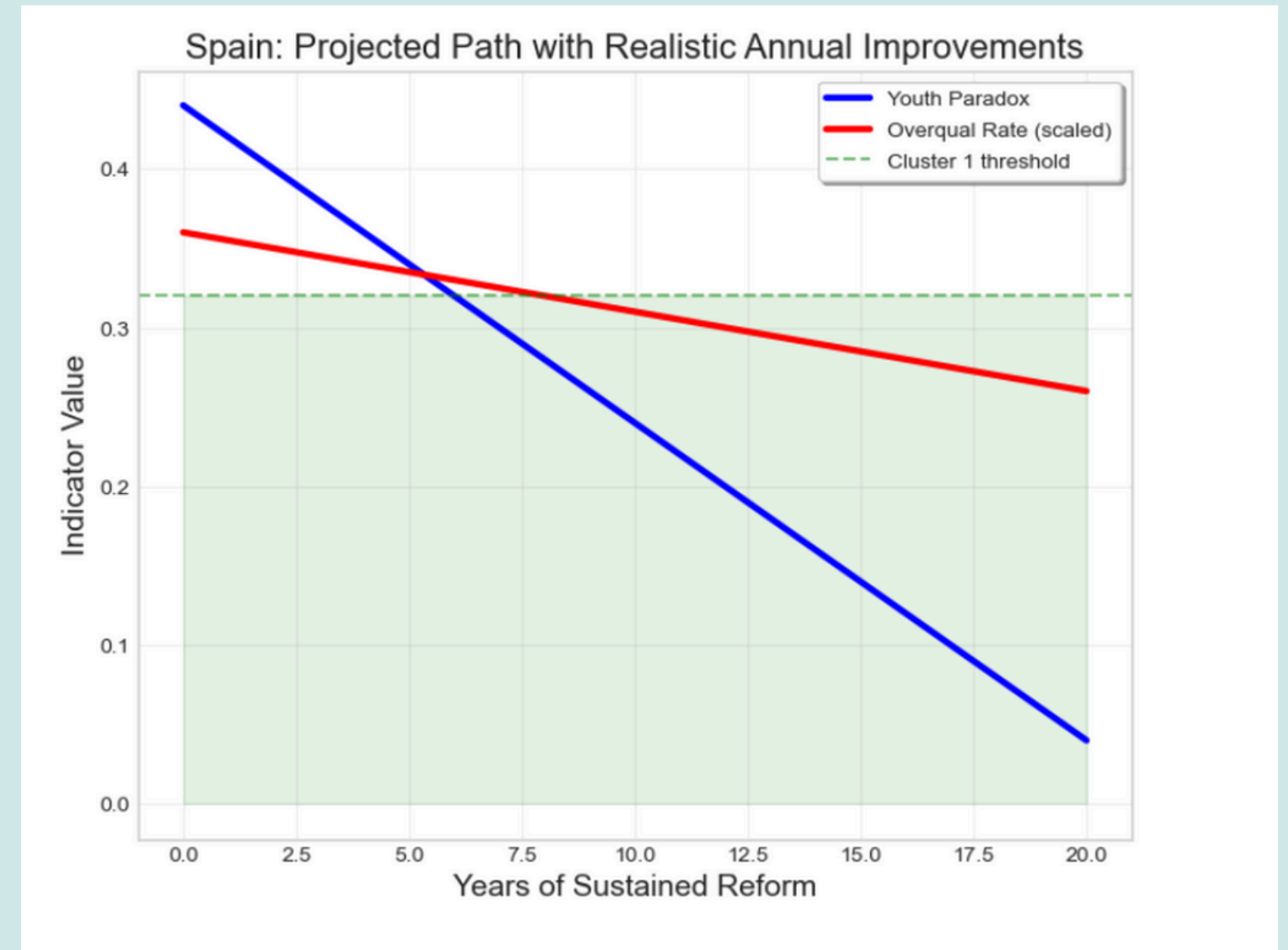


Finding: Mismatch clusters show remarkable stability in size over the decade.

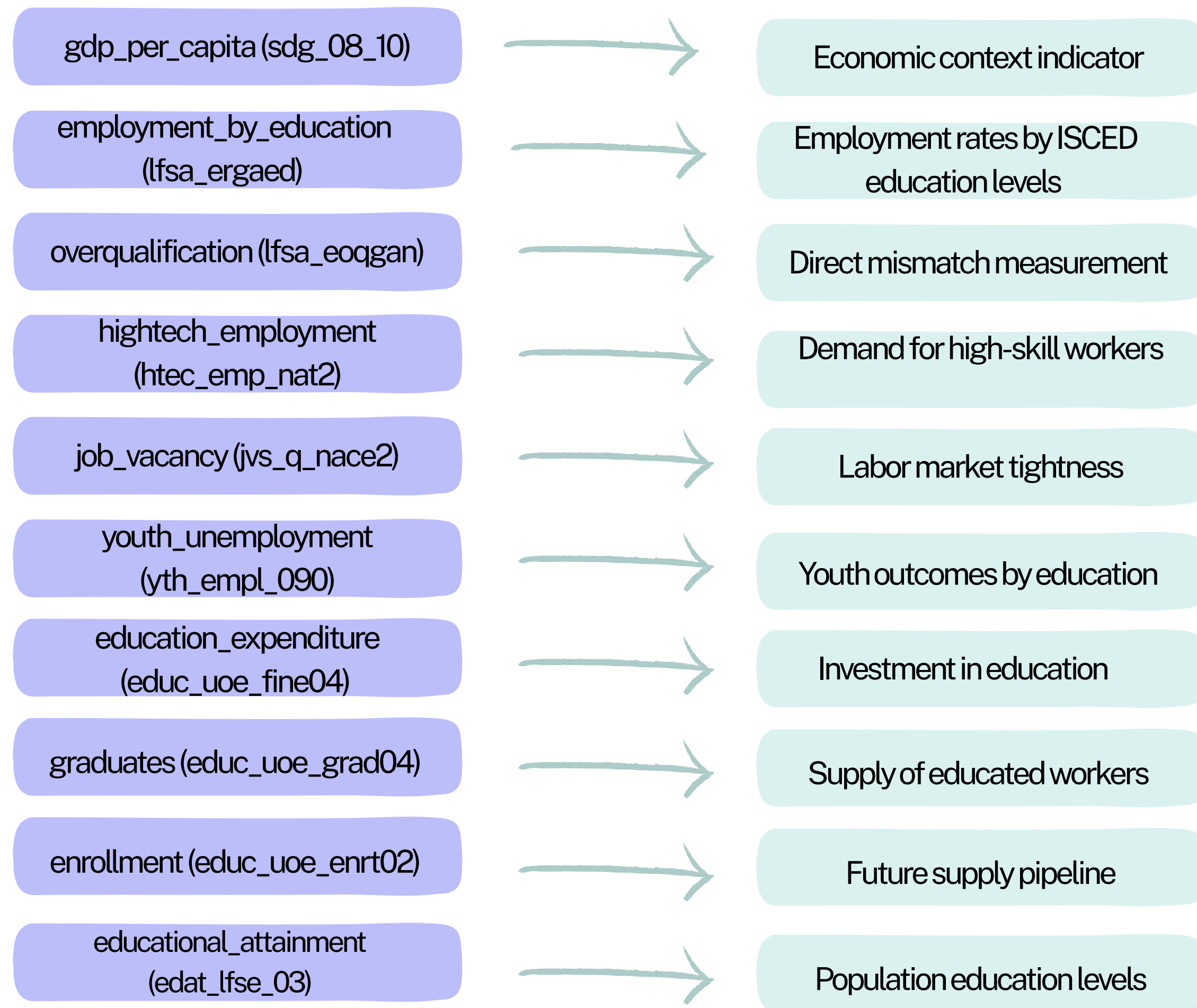
The Core Finding of Our Analysis

A what-if scenario: How would the trajectory of a struggling nation change if everything worked in its favor?

Education-skill mismatches are **STRUCTURAL TRAPS** requiring sustained, multi-dimensional interventions over 10+ years to escape.



Thank You



Feature importance for predicting transitions:

edu_investment: 0.292
employment_efficiency: 0.214
overqualification_rate: 0.166
gdp_per_capita: 0.127
youth_edu_paradox: 0.105
gdp_growth: 0.097

Identifying countries at risk of transition...

No countries have >30% transition probability.

Top 10 countries by transition probability:

FI: 5.2% (Cluster 2)
CZ: 1.2% (Cluster 2)
EE: 0.8% (Cluster 0)
RO: 0.8% (Cluster 1)
CH: 0.7% (Cluster 2)
NO: 0.7% (Cluster 2)
MT: 0.7% (Cluster 1)
TR: 0.7% (Cluster 1)
LT: 0.7% (Cluster 1)
HU: 0.6% (Cluster 1)

Transition probability statistics:

Max: 5.2%
Mean: 0.6%
Median: 0.4%