

# Bridging Skills Demand and Vocational Qualifications: Linking Online Job Postings with National Skills and Qualification Frameworks

HARNESSING WEB DATA FOR NEXT-GENERATION SKILLS INTELLIGENCE  
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# What we will cover

01

## Motivation and research question

Why a semantic bridge between OJPs and VET?

02

## Italian institutional landscape

VET pathways anchored on the Atlante del Lavoro

03

## Data sources

11.8 M postings · 5,160 courses · 43.5 K skill links

04

## Methodology

SBERT + LLM matching, notation, and the VAI

05

## Results · skill level

Common and Specialised mismatches

06

## Results · sectoral level

VAI ranking and the four-quadrant matrix

07

## Policy implications

A continuous-monitoring policy cycle

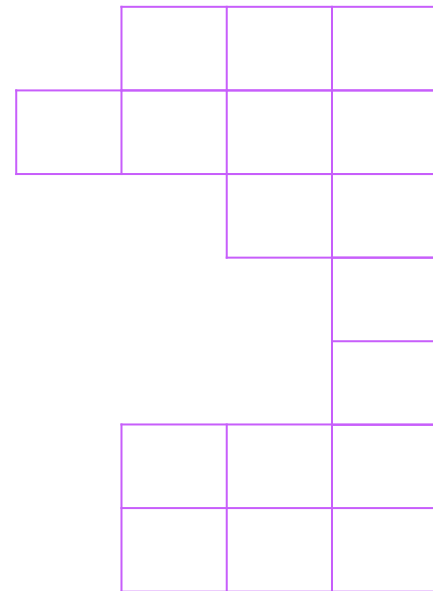
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## Next steps and research agenda

From offer to effective skill provision

# Why a semantic bridge between OJPs and VET?

- OJP-based skills intelligence is organised around occupations (ISCO\ESCO, national qualification).
- VET systems are organised around qualifications and learning outcomes.
- The two universes do not share a common semantic layer.
- **Result: web evidence circulates without informing curriculum updates.**



How aligned is what we teach  
with what employers demand?

And how can the answer be made:

- *measurable at sector level,*
- *interpretable across skill types,*
- *and operational for VET policy?*

# Contribution

## METHODOLOGICAL

### A replicable framework

Two-stage NLP pipeline (multilingual SBERT + LLM filter) anchored on a national qualification repository through the Atlante del Lavoro activities layer.

## ANALYTICAL

### A composite indicator: VET Alignment Index (VAI)

Two-component index combining quantitative (top-K demand-weighted coverage) and qualitative (SBERT cosine similarity) terms, separately by skill type.

# Roadmap

## EMPIRICAL

### First national-level evidence

Skill- and sector-level alignment for the full Italian VET system: 5k courses across leFP / IFTS / ITS pathways and 23+1 Economic-Professional Sectors.

## POLICY

### From description to decision

A continuous-monitoring cycle and a four-quadrant priority matrix that translate the indicator into operational signals for sector boards and qualification authorities.

# Italian VET: three pathways, one semantic infrastructure

**Atlante del Lavoro e delle Qualificazioni:** Reticular description of the world of work · 24+1 Economic-Professional Sectors (SEP) · 840 Areas of Activity (ADA) · indexed to CP2021 occupations (mappable to ISCO-08) and to ATECO 2007 sectoral codes (NACE Rev. 2 compatible)

## 1 leFP

EQF 3-4 · 3-4 years

### GOVERNANCE

Regional providers

### TARGET QUALIFICATION

Vocational Operator Qualification (3 yrs) ·  
Vocational Diploma (4 yrs)

### COVERAGE IN CORPUS

Upper-secondary regional pathway ·  
largest by volume

## 2 IFTS

EQF 4 · ~800-1,000 hours

### GOVERNANCE

Regional partnerships (schools, providers,  
firms, universities)

### TARGET QUALIFICATION

Higher Technical Specialization  
Certificate

### COVERAGE IN CORPUS

Intermediate share · technician-level  
demand

## 3 ITS Academy

EQF 5 · 2 years (1,800-2,000 hrs)

### GOVERNANCE

ITS Foundations (school, university, firm,  
training body)

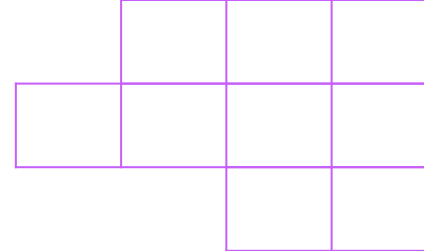
### TARGET QUALIFICATION

Higher Technician Diploma

### COVERAGE IN CORPUS

Smaller volume · strategic technological  
areas (D.M. 203/2023)

# ADA as the semantic pivot between demand and supply



## DEMAND SIDE

### Online job postings

Each posting carries an occupational code (CP2021 / ISCO-08) and an ATECO sectoral code attached by the Lightcast pipeline.



CP2021 codes index one or more ADA

## SEMANTIC BRIDGE

### Areas of Activity (ADA)

# 840

#### 4-LEVEL HIERARCHY

Sectors → Processes → ADA → Qualifications

## SUPPLY SIDE

### Vocational qualifications

Each qualification in the INAPP National Repository is linked to one or more ADA in a logic of potential employment outlets.



Qualifications index one or more ADA

*Each ADA is doubly referenced — via occupational codes (reachable from postings) and via qualifications (reachable from courses).*

*Routing demand and supply through a shared ADA layer enables a like-for-like comparison: this is the bridge the rest of the study exploits.*

# Three sources, one integrated dataset

## CLASSIFICATIONS USED

### SKILLS

Lightcast Open Skills taxonomy · Specialised (technical) vs Common (transversal) types

### OCCUPATIONS

CP2021 (Italian) at posting level · mappable to ISCO-08 for international comparability

### SECTORS

ATECO 2007 at posting level · compatible with NACE Rev. 2 · projected onto 25 SEP via the Atlante

### QUALIFICATIONS

INAPP National Repository · 840 ADA · leFP / IFTS / ITS pathways

### EMBEDDINGS

Multilingual SBERT (sentence-transformers, 384-dimensional) for the qualitative component of the VAI

## ITALIAN ONLINE JOB POSTINGS

Lightcast Italian database (11M job postings) · collected since January 2023 · skills extracted by the Lightcast Open Skills NLP pipeline.

## VOCATIONAL COURSES

leFP · IFTS · ITS Academy · National Repository of Qualifications maintained by INAPP (~5k qualifications).

## COURSE-SKILL LINKS

After matching · 8.4 Lightcast skills per course on average · 23+1 SEP sectors covered (~43k skills extracted).

# The course corpus by pathway

PATHWAY	EQF	DURATION	GOVERNANCE	COURSES (N)	SHARE
leFP — Initial vocational pathways	3–4	3–4 years	Regional (and IP)	≈ 3,600	≈ 70%
IFTS — Higher technical specialisation	4	≈ 800–1,000 h	Regional partnerships	≈ 1,200	≈ 23%
ITS Academy — Tertiary technical	5	2 years (1,800–2,000 h)	ITS Foundations	≈ 360	≈ 7%
<b>ALL PATHWAYS</b>	<b>3–5</b>	<b>—</b>	<b>—</b>	<b>5,160</b>	<b>100%</b>

## Pathway weighting

leFP dominates by volume (≈70% of courses), which shapes how the integrated dataset represents Italian VET.

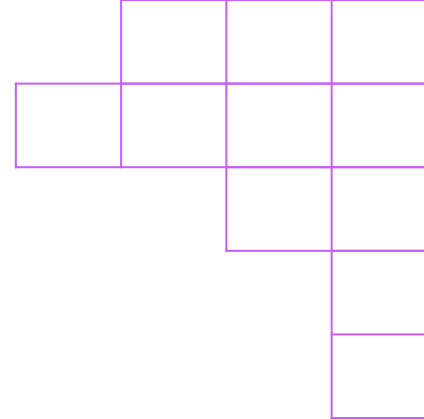
## Granularity

ITS Academy courses, though fewer, contribute disproportionately to the technical and digital skill coverage.

## Why no enrolment weights

Enrolment counts are inconsistently reported across pathways and regions; **we measure the offer of training, not exposure.**

# Three sources, one integrated dataset



## DEMAND SIGNAL

Online job posting

### Electrical installer

for residential buildings

Source

Lightcast Italian OJPs

Region

Lombardia · 2024



## OCCUPATIONAL CODE

CP2021 (mappable to ISCO-08)

### 6.1.3.7.0

Electricians and installers of electrical systems in civil constructions

ISCO-08

7411 · Electrical fitters

EQF level

3-4



## ATLANTE PIVOT

Sector + Area of Activity

### SEP 10

Meccanica · ADA.10.04.05

ADA label

Installation of electrical systems

Process

Plant engineering — residential



## QUALIFICATION

INAPP National Repository

### Operatore elettrico

leFP · EQF 3 · 3-year pathway

Channel

leFP

Indexed to ADA

via the Atlante



## SUPPLY SIGNAL

Course in the corpus

### Course

INSTALLATORE E  
MANUTENTORE ELETTRICO  
ED ELETTRONICO  
(Lombardia)

Linked to ADA

via the qualification

# Linking courses to skills through SBERT + LLM verification

## WHY A TWO-STAGE DESIGN

Recall and precision trade off across thresholds. SBERT retrieval maximises recall on a wide candidate set; the LLM step recovers precision by rejecting semantically-plausible but contextually-irrelevant pairs.

**01**

### Input

Free-text course descriptions and competence units in Italian.  
Sources: National Repository, leFP profiles, ITS reference profiles (D.M. 203/2023).

**02**

### Retrieval

A multilingual SBERT model (paraphrase-multilingual-m-pnet-base-v2) encodes both course passages and Lightcast skill descriptions into a 384-dimensional space. Top-K candidate skills are retrieved per course.

**03**

### LLM verification

A large language model classifies each candidate (skill, course) pair as relevant or non-relevant, using the course passage and the canonical skill name + description as input. Acts as a precision filter.

**04**

### Thresholding

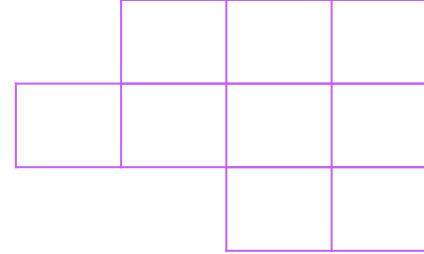
A cosine threshold ( $\geq 0.55$  in baseline) is retained after LLM filtering. Validated against a manually-coded subsample of courses across pathways.

**05**

### Output

43,545 validated course-skill links · 8.4 skills per course on average · attached to the ADA layer and to the SEP sector of the course.

# Linking courses to skills through SBERT + LLM verification



## WHY A TWO-STAGE DESIGN

SBERT alone would keep all eight skills above 0.55; the LLM filter rejects Power Distribution, Solar Panel Installation and Electrical Drafting as out of scope for an EQF-3 residential course.

### COURSE INPUT

**Impianti elettrici civili (IeFP — Operatore elettrico, EQF 3)**

### ADA

09.05.07

### SEP

09 · Construction

## STAGE 1 + 2 · SBERT TOP-K WITH LLM VERIFICATION

CANDIDATE SKILL	COS	VERDICT	LLM RATIONALE
Electrical Installation	0.91	KEEP	Core practical skill for residential wiring.
Wiring	0.88	KEEP	Directly taught in module 2 of the course.
Electrical Safety	0.84	KEEP	Explicit competence unit on safety norms.
Circuit Testing	0.79	KEEP	Listed under verification activities.
Conduit Bending	0.71	KEEP	Practical activity in the workshop block.
Power Distribution	0.69	DROP	Industrial scope; course is residential only.
Solar Panel Installation	0.66	DROP	Out of programme scope at EQF 3.
Electrical Drafting	0.58	DROP	Not in learning outcomes; auxiliary skill.

## OUTPUT · ATTACHED TO THE COURSE

# 5

validated skill links  
(5 KEEP / 3 DROP)

### VALIDATED SKILL SET

- Electrical Installation
- Wiring
- Electrical Safety
- Circuit Testing
- Conduit Bending

# The VET Alignment Index: the elementary indicators

## PRIMITIVES

<b>P</b>	Universe of Italian online job postings	$ P  = 11,760,514$
<b>C</b>	Universe of vocational courses	$ C  = 5,160$
<b>S</b>	Universe of Lightcast skills (Spec + Common)	-
$\Sigma$	23+1 Economic-Professional Sectors	$ \Sigma  = 24$
<b>T(s)</b>	Skill type: Specialised   Common	-
<b>skills(p)</b>	Skills extracted from posting p	from postings
<b>skills(c)</b>	Skills attached to course c	from courses
<b>s(p), s(c)</b>	SEP sector of posting / course	$\in \Sigma$

## ELEMENTARY INDICATORS

### **Demand share of skill s**

demand\_share(s) =

$$| \{ p \in P : s \in \text{skills}(p) \} | / |P|$$

### **Supply share of skill s**

supply\_share(s) =

$$| \{ c \in C : s \in \text{skills}(c) \} | / |C|$$

### **Skill-level gap**

$$\text{gap}(s) = \text{demand\_share}(s) - \text{supply\_share}(s)$$

*Sectoral versions: restrict numerator / denominator to postings and courses in  $\sigma$ .*

# The VET Alignment Index and its two components

$$VAI(\sigma, T) = \alpha \cdot \min(\text{Quant}(\sigma, T), 1) + (1 - \alpha) \cdot \text{Qual}(\sigma, T)$$

baseline  $\alpha = 0.2$  (qualitative-dominant) ·  $K = 30$  (top-K cutoff, following OECD 2023) · evaluated separately for  $T \in \{\text{Specialised}, \text{Common}\}$

## QUANTITATIVE COMPONENT

### Demand-weighted coverage

$D\_K(\sigma, T)$  = top-K skills of type T in  $\sigma$   
ranked by  $\text{demand\_share}_\sigma$

$\text{taught}(\sigma, T) = \{ s \in S : T(s) = T, \\ \text{supply\_share}_\sigma(s) > 0 \}$

**Quant( $\sigma, T$ ) =**

$\sum_{\{ s \in D\_K(\sigma, T) \cap \text{taught}(\sigma, T) \}} \\ \text{demand\_share}(s)$

## QUALITATIVE COMPONENT

### Semantic similarity

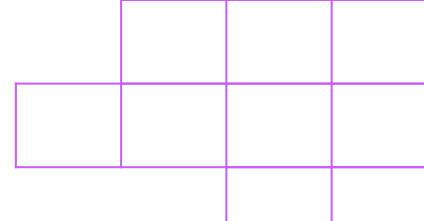
$v(s) \in \mathbb{R}^{384}$  = multilingual SBERT  
embedding of skill s

$\text{sim}(s, s') = \cos( v(s), v(s') )$

**Qual( $\sigma, T$ ) =**

$(1/|D\_K|) \cdot \sum_{\{ s \in D\_K \}} \max_{\{ s' \in \\ \text{taught} \}} \text{sim}(s, s')$

# Calibration choices



## WEIGHTING $\alpha$

$\alpha = 0.2$

### ARGUMENT

Quant has limited discriminatory power once partial coverage is achieved — it cannot distinguish a sector teaching ten of the top demanded skills well from one teaching them poorly. Qual recovers that information.

### ROBUSTNESS

Re-running with  $\alpha \in \{0.0, 0.2, 0.5, 0.8, 1.0\}$  shifts sector ranks by at most two positions; the bottom-quartile priority ranking is preserved at every  $\alpha$ .

## TOP-K CUTOFF

$K = 30$

### ARGUMENT

Follows the OECD (2023) precedent. Captures the demand-relevant tail without diluting the indicator with marginally-demanded skills.

### ROBUSTNESS

Increasing  $K$  from 30  $\rightarrow$  50  $\rightarrow$  100 leaves the identity of the bottom-three and top-three sectors unchanged. Restricting to courses with explicit ADA metadata yields the same substantive ranking.

## SUPPLY SIDE

Offer, not exposure

### ARGUMENT

Each course counts once: VAI measures the structure of the training offer, not the volume of learners actually exposed to it. Enrolment data are not consistently available across pathways and regions.

### ROBUSTNESS

Enrolment-weighted re-estimation deferred to future work. Qualitative findings are not expected to change; quantitative magnitudes might.

# The VET Alignment Index: how to read

## FIRST COMPONENT: BREADTH

### Quant

*What share of demand is taught at all?*

*$\sum$  demand\_share over taught top-K skills*

#### WHAT IT MEASURES

The fraction of demand-weighted skills covered by the training offer - coverage of what employers actually ask for.

#### RATIONALE

A sector can teach many skills, but if none are demanded, alignment is illusory. Quant disciplines the indicator with employer-weighted demand and penalises curricula that ignore high-frequency skills.

## SECOND COMPONENT: DEPTH

### Qual

*How close are the skills that ARE taught?*

*mean cosine to closest taught skill, over top-K demanded*

#### WHAT IT MEASURES

Semantic proximity in SBERT space between every demanded skill and the closest skill actually taught - the depth of the match.

#### RATIONALE

Two sectors with identical Quant can differ sharply: one teaches the demanded skill, the other a vague cousin. Qual recovers this distinction that Quant cannot make.

#### READING A VALUE · $VAI \in [0, 1]$

< 0.30 · Severe misalignment

0.30 – 0.45 · Weak

0.45 – 0.55

0.55 – 0.70 · Good

> 0.70 · Strong alignment

Paper 0.276

Chemicals 0.350

Construction 0.584

Common area 0.740

# The VET Alignment Index: decomposition for Construction

## INPUT · TOP-K DEMANDED SPECIALISED SKILLS, $\sigma = 09$ (CONSTRUCTION)

#	SPECIALISED SKILL	DEMAND	TAUGHT?	COS	CONTRIB. Q	CONTRIB. SIM
1	Wiring	8.4%	✓	0.92	8.4	0.92
2	Electrical Installation	6.1%	✓	0.89	6.1	0.89
3	Working Drawings	5.7%	✓	0.81	5.7	0.81
4	Structural Steel	4.2%	—	0.42	0.0	0.42
5	Plumbing	3.9%	✓	0.78	3.9	0.78
6	Welding	3.5%	✓	0.74	3.5	0.74
7	Safety on Site	3.2%	✓	0.88	3.2	0.88
...	(further skills, ranks 8–30)	...	...	...	...	...
TOTAL OVER K = 30					48.7%	0.76

## COMPUTATION · SPECIALISED

### STEP 1 · QUANTITATIVE

$Quant(\sigma, T) = \sum \text{demand\_share over taught top-K skills}$

**Quant = 0.487**

### STEP 2 · QUALITATIVE

$Qual(\sigma, T) = \text{mean cosine to closest taught skill, over top-K demanded}$

**Qual = 0.762**

### STEP 3 · COMPOSITE ( $\alpha = 0.2$ )

$VAI = \alpha \cdot \min(Quant, 1) + (1 - \alpha) \cdot Qual$

$= 0.2 \cdot 0.487 + 0.8 \cdot 0.762$

**VAI(Spec) = 0.609**

# Four findings

## Skill-level evidence and sectoral patterns

21% / 28%

Asymmetric universes

*The supply universe is structurally narrower than the demand universe; only 21% of the Specialised skill set and 28% of the Common skill set are covered by both sides (Jaccard).*

20.4 pp

Qualitative transversal gap

*The transversal skills employers most consistently demand are essentially absent from formal curricula: a sector-invariant, systematic and severe under-supply.*

20k vs 398

Dispersed Specialised mismatch

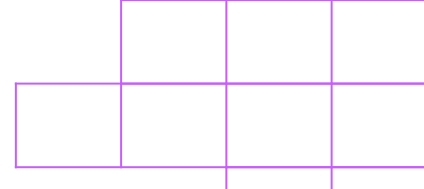
*Demand for technical skills is spread over a much larger universe than for transversal skills; the Specialised mismatch is sector-specific and concentrated on cross-occupational competences.*

18 / 24

Structural type asymmetry

*Manufacturing-intensive sectors codify technical content; care and service sectors codify transversal content. The asymmetry is institutional, not pedagogical.*

# Demand and supply universes, by skill type



QUANTITY	SPECIALISED	COMMON
Distinct skills demanded (postings)	20,791	398
Distinct skills taught (courses)	4,986	143
Skills both demanded and taught	4,476	119
Overlap (% of union)	21%	28%

Size of the demand and supply universes by skill type.  $|P| = 11,760,514$  postings,  $|C| = 5,160$  courses.

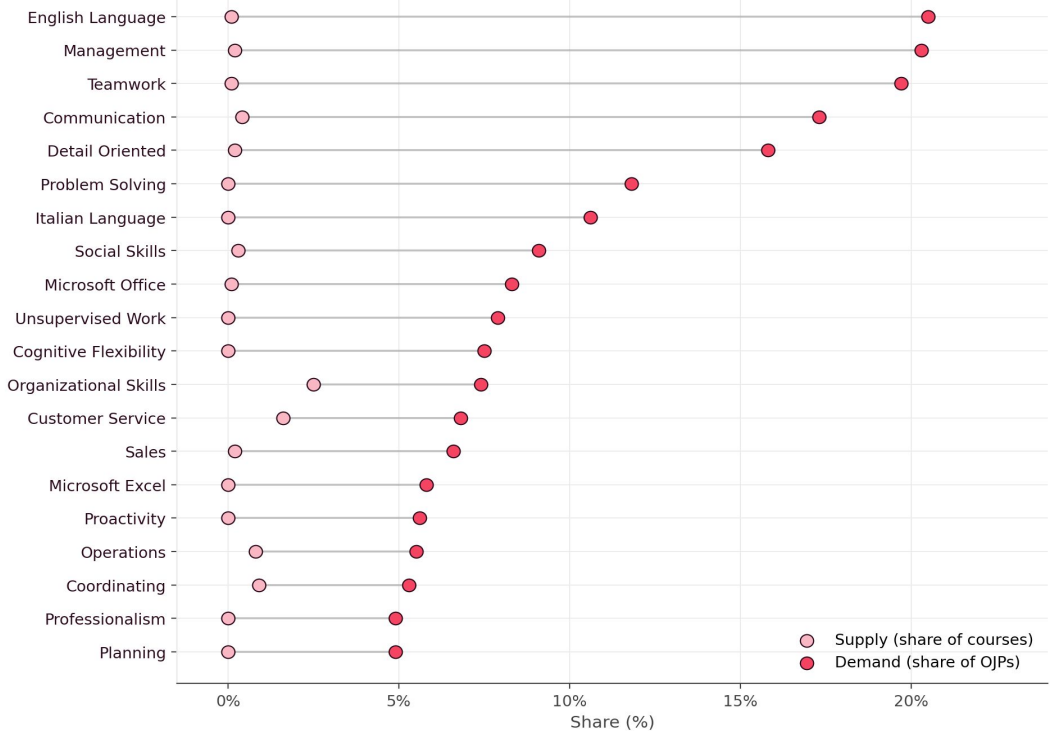
**STRUCTURAL ASYMMETRY**  
Specialised demand spans 20,791 distinct skills; the supply universe covers 4,986 — narrower by design, since VET cannot teach the full long tail.

**TYPE COMPRESSION**  
Common demand spans only 398 skills, of which 143 are taught — but with a far more concentrated demand distribution (top ten skills > 50%).

# Common skills: a systematic and uniform under-supply

Five of the top fifteen Common skills are taught in 0% of the 5k vocational courses analysed

Common (transversal) skills — top 20 demanded vs training supply

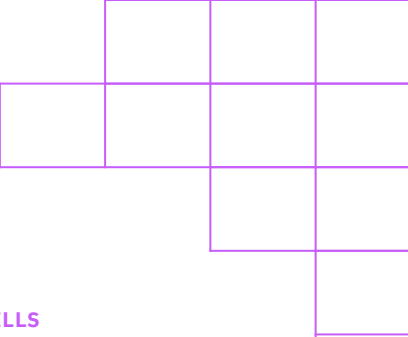


TOP 10 COMMON-SKILL GAPS (PERCENTAGE POINTS)

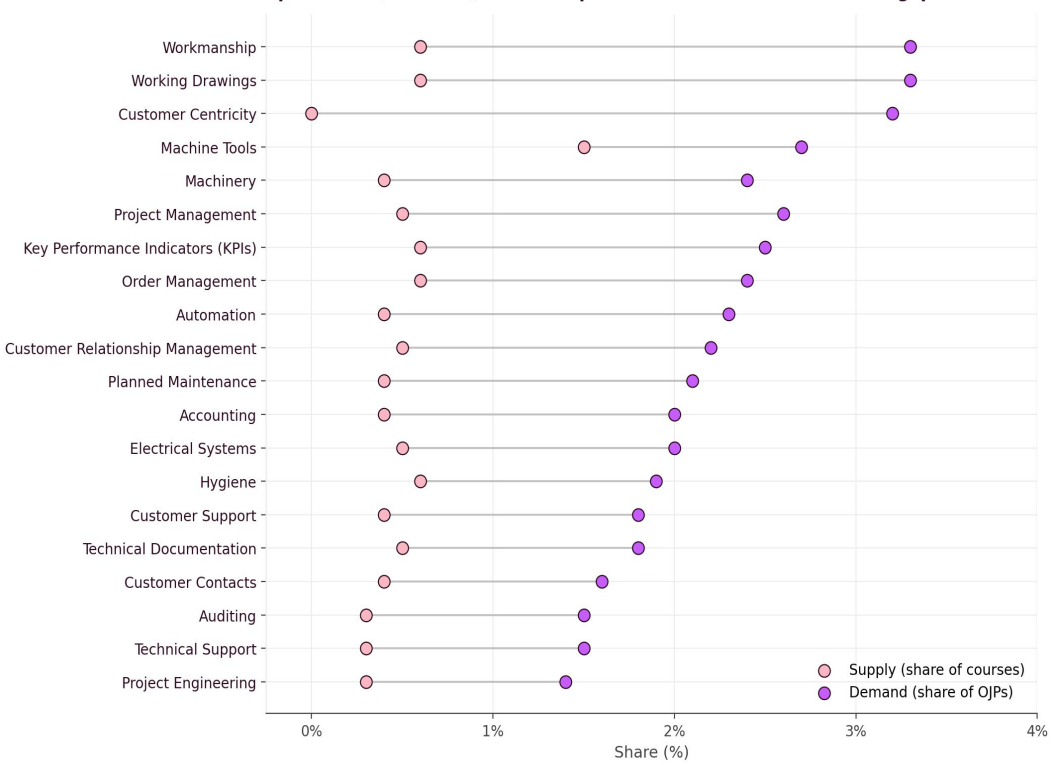
SKILL	DEMAND	SUPPLY	GAP (PP)
<b>English Language</b>	20.5%	0.1%	<b>20.4</b>
<b>Management</b>	20.3%	0.2%	<b>20.1</b>
<b>Teamwork</b>	19.7%	0.1%	<b>19.6</b>
<b>Communication</b>	17.3%	0.4%	<b>16.9</b>
<b>Detail Oriented</b>	15.8%	0.2%	<b>15.6</b>
<b>Problem Solving</b>	11.8%	0%	<b>11.8</b>
<b>Italian Language</b>	10.6%	0%	<b>10.6</b>
<b>Social Skills</b>	9.1%	0.3%	<b>8.8</b>
<b>Microsoft Office</b>	8.3%	0.1%	<b>8.2</b>

# Specialised skills: a sector-specific mismatch

Specialised gaps are spread thinly: no single skill exceeds 5% of postings, but a long tail of mid-frequency competences accumulates.



Specialised (technical) skills — top 20 most-demanded with widest gaps

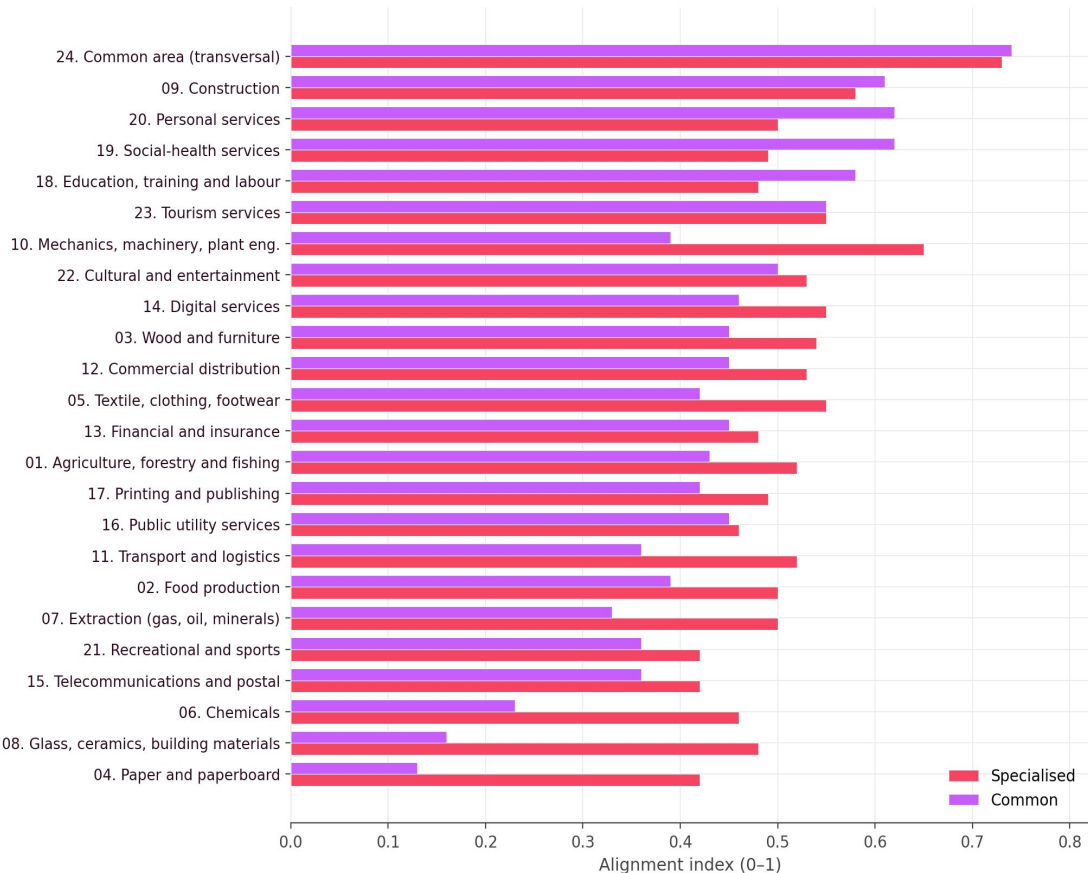


TOP UNDER-SUPPLIED SPECIALISED SKILLS

SKILL	DEMAND	SUPPLY	GAP (PP)
<b>Loading and Unloading</b>	4.9%	0.0%	<b>4.9</b>
<b>Packaging and Labeling</b>	3.6%	0.0%	<b>3.6</b>
<b>Workmanship</b>	3.3%	0.6%	<b>2.7</b>
<b>Working Drawings</b>	3.3%	0.6%	<b>2.7</b>
<b>Customer Centricity</b>	3.2%	0.0%	<b>3.2</b>
<b>Results Orientation</b>	3.2%	0.5%	<b>2.7</b>
<b>Project Management</b>	2.6%	0.5%	<b>2.1</b>
<b>Invoicing</b>	2.4%	0.0%	<b>2.4</b>
<b>Order Management</b>	2.4%	0.6%	<b>1.8</b>
<b>Customer Relationship Mgmt</b>	2.2%	0.5%	<b>1.7</b>

# Sectoral alignment and the type-asymmetry divide

Sector alignment index — Specialised vs Common skills



**WORST ALIGNED (LOW VAI)**

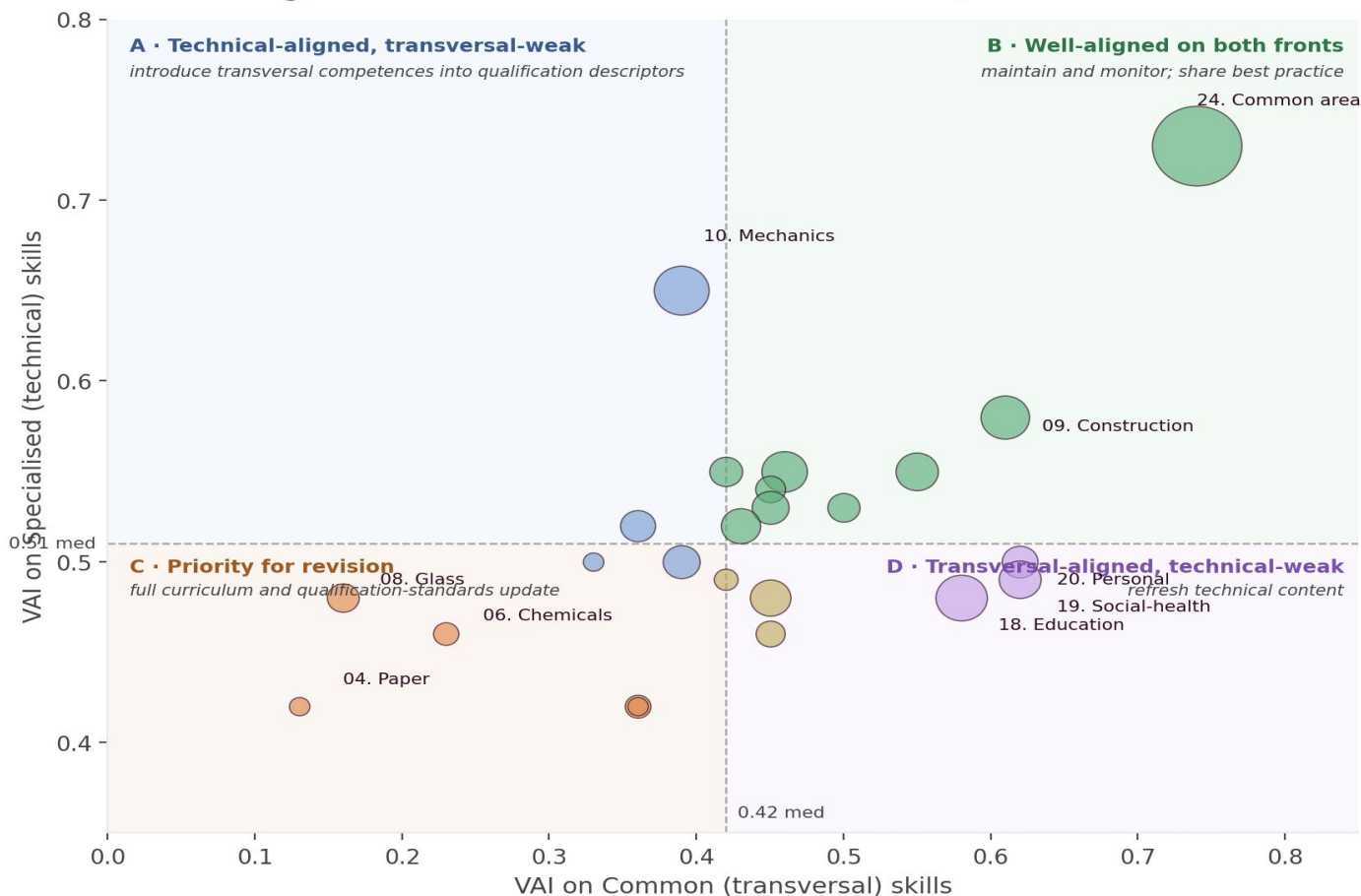
<b>04. Paper and paperboard</b>	<b>0.276</b>	n=15
<b>08. Glass, ceramics, building materials</b>	<b>0.322</b>	n=108
<b>06. Chemicals</b>	<b>0.350</b>	n=68
<b>BEST ALIGNED (excl. Common area)</b>		
<b>09. Construction</b>	<b>0.584</b>	n=250
<b>20. Personal services</b>	<b>0.560</b>	n=136
<b>19. Social-health services</b>	<b>0.558</b>	n=187

**STRUCTURAL DIVIDE**

*Eighteen of twenty-four substantive sectors show VAI(Spec) > VAI(Common), with a mean asymmetry of +0.13. The asymmetry is institutional, not pedagogical — manufacturing codifies technical work; care and service sectors codify transversal work.*

# From evidence to action: a four-quadrant priority matrix

**Sectoral alignment matrix — boundaries at median VAI; bubble area  $\propto$  course count**



## QUADRANT A

### Technical-aligned, transversal-weak

Mechanics, Glass, Chemicals · integrate transversal competences into qualification descriptors.

## QUADRANT B

### Well-aligned on both

Construction, Tourism, Common area · maintain and disseminate as benchmark cases.

## QUADRANT C

### Priority for revision

Paper, Telecoms, Recreational · full curriculum and qualification-standards update.

## QUADRANT D

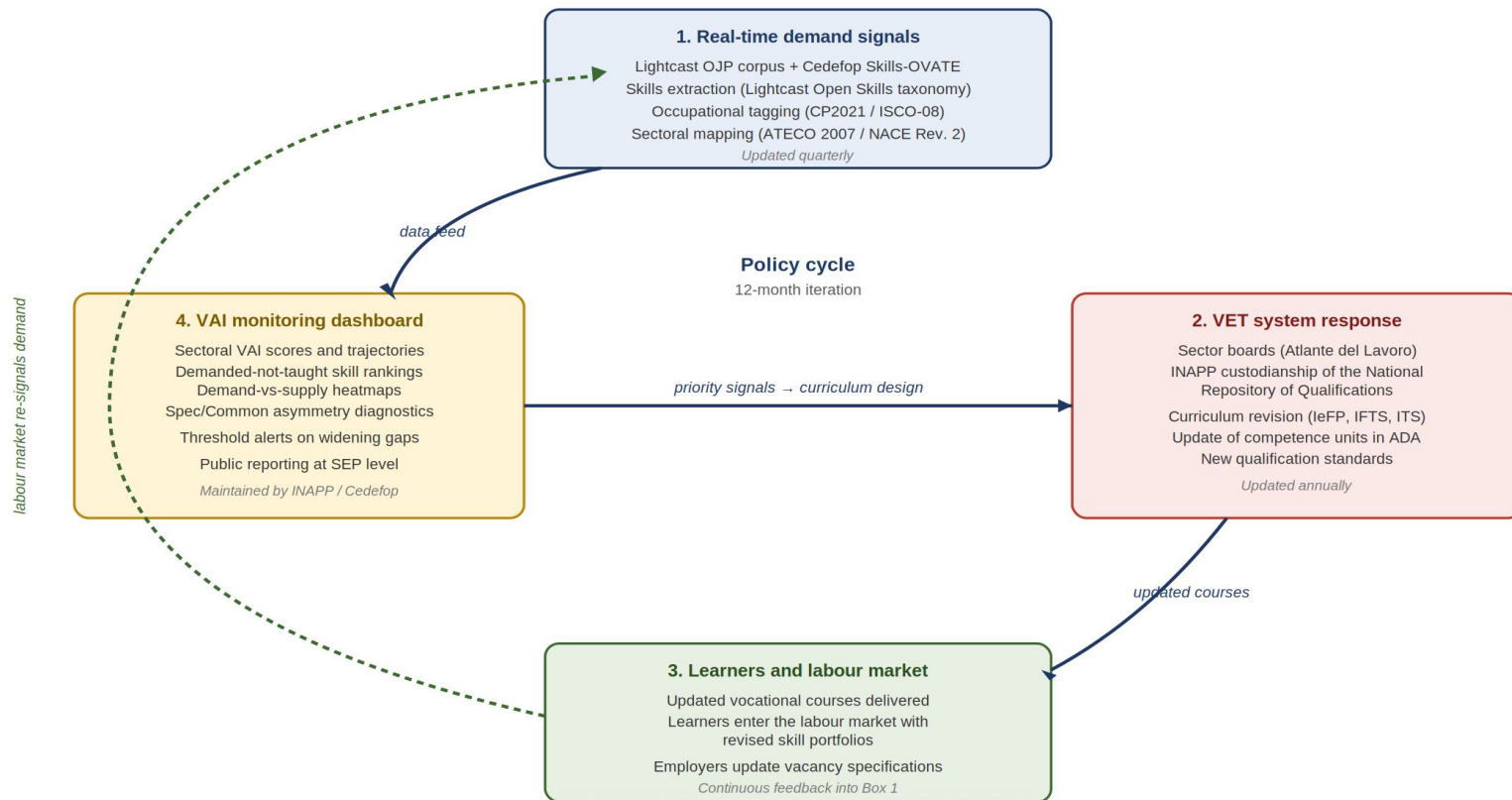
### Transversal-aligned, technical-weak

Personal services, Social-health, Education · refresh technical content; preserve transversal.

# A continuous-monitoring policy cycle

Figure 12. A continuous-monitoring policy cycle anchored on the VAI

From real-time labour market signals to curriculum revision and back



Actors involved at every stage: Cedefop, INAPP, Ministry of Labour, Ministry of Education, Regions, social partners and ITS Foundations.

# Thank you

LIGHTCAST . IO



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