

Comparing Pre-service and In-service Teachers' AI Competence Self-Efficacy

-- Measurement Invariance and Latent Mean Differences

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Theoretical and Empirical Background

01 Research Background

- AI has reshaped education
- AI potential in education remains underutilized
- AI Literacy Framework (OECD and EU, 2025), primary, secondary education
- Teachers' roles in K-12 stages, cultivate AI-literate citizens and future labor
- Significance of pre- and in-service teachers' AI competence

02 Theoretical Frameworks

- TPACK, DigCompEdu, Teacher Digital Competence (TDC) framework (Mishra and Koehler, 2006; Redecker, 2017; Falloon, 2020)
- UNESCO: AI Competency Framework for Teachers (Miao & Cukurova, 2024)
- Five dimensions: Human-centered Mindset, Ethics of AI, AI Pedagogy, AI Foundations and Applications, AI for Professional Learning

03 Standardized Instruments

- Based on TPACK model: Intelligent/AI-TPACK scales (e.g. Celik, 2023; Ning et al., 2024, 2025) or based on TDC framework (e.g. Falloon, 2020)
- UNESCO: AI competency framework for Teachers (Miao & Cukurova, 2024)
- An extra dimension, AI assessment from DigCompEdu (Redecker, 2017)
- Teacher AI Competence Self-Efficacy (TAICS) scale (Chiu et al., 2024)

04 Research Gaps

- Research on teachers' AI competence is still in its infancy
- Theoretical frameworks just emerged
- Standardized measurements are scarce
- Chinese policy momentum is accelerating (Ministry of Education, 2025)
- Chinese mainland lacks culturally adapted tools for empirical research

Research Questions

RQ 1: To what extent does the TAICS scale demonstrate a **valid factorial structure** in assessing AI competence self-efficacy among pre- and in-service teachers in mainland China?

RQ 2: Is **measurement invariance** between the groups of pre- and in-service teachers supported?

RQ 3: Are there significant **latent mean differences** across TAICS dimensions between the groups of pre- and in-service teachers?

Results

RQ1 Table 1 & Table 2. Standardized Factor Loadings of EFA and ESEM

	AIK	AIP	AIA	AIE	HCE	PEN		AIK	AIP	AIA	AIE	HCE	PEN
AIK_1	0.651						AIK_1	0.660	0.037	-0.001	0.064	0.111	0.003
AIK_2	0.437	0.342					AIK_2	0.451	0.338	0.048	0.050	-0.002	0.084
AIK_3	0.762						AIK_3	0.817	-0.087	0.070	0.015	-0.038	0.004
AIK_4	0.487	0.363					AIK_4	0.579	0.304	-0.014	0.046	0.006	0.078
AIP_1	0.840						AIP_1	0.048	0.779	0.052	0.060	0.066	0.036
AIP_2	0.767						AIP_2	0.056	0.718	0.102	0.055	0.067	0.043
AIP_3	0.714						AIP_3	0.090	0.612	0.150	0.035	0.061	0.103
AIP_4	0.530						AIP_4	0.150	0.430	0.236	0.037	-0.035	0.223
AIA_1	0.519						AIA_1	0.027	0.405	0.399	0.058	0.111	0.042
AIA_2		0.699					AIA_2	0.058	0.040	0.766	0.022	0.044	0.059
AIA_3		0.670					AIA_3	0.043	0.047	0.758	0.049	0.060	0.074
AIA_4		0.664					AIA_4	0.044	0.120	0.721	0.080	0.067	0.007
AIE_1			0.725				AIE_1	0.008	0.061	-0.121	0.740	0.080	-0.013
AIE_2			0.889				AIE_2	0.017	-0.062	0.038	0.914	-0.038	-0.011
AIE_3			0.801				AIE_3	0.051	-0.001	-0.010	0.852	0.003	0.030
AIE_4			0.809				AIE_4	-0.049	0.034	0.086	0.864	0.015	0.007
HCE_1				0.730			HCE_1	0.040	0.115	-0.068	0.099	0.731	0.048
HCE_2				0.897			HCE_2	0.011	-0.005	0.059	-0.015	0.931	-0.053
HCE_3				0.679			HCE_3	-0.012	-0.014	0.060	0.053	0.588	0.157
HCE_4				0.694			HCE_4	0.121	-0.062	0.097	0.013	0.603	0.155
PEN_1					0.417		PEN_1	0.101	0.143	-0.087	0.143	0.251	0.415
PEN_2					0.605		PEN_2	0.037	0.050	-0.101	0.191	0.158	0.571
PEN_3					0.756		PEN_3	0.012	0.077	0.000	0.018	0.043	0.815
PEN_4					0.714		PEN_4	0.052	-0.052	0.227	-0.006	-0.012	0.717

③ ESEM (full sample)

- Six-factor: good fit ($\chi^2/df = 3.684$, robust CFI = 0.989, robust TLI = 0.979, robust RMSEA = 0.032, SRMR = 0.008)
- AIK_2 and AIK_4: primary loadings on AIK; secondary loadings on AIP
- AIA_1: high cross loadings on AIP; removed
- Revised ESEM model: slightly improved fit ($\chi^2/df = 3.652$, robust CFI = 0.990, robust TLI = 0.980, robust RMSEA = 0.032, SRMR = 0.008)

RQ1 Table 3. Concurrent Validity with Adapted DigCompEdu Scale

	AIK	AIP	AIA	AIE	HCE	PEN
ADMIN	.564***	.539***	.488***	.539***	.541***	.574***
LESS	.576***	.559***	.492***	.589***	.581***	.600***
TEACH	.563***	.576***	.539***	.579***	.571***	.604***
ASSES	.563***	.620***	.660***	.482***	.542***	.598***
EMPOW	.618***	.638***	.609***	.597***	.602***	.640***
COMP	.612***	.675***	.694***	.556***	.593***	.656***
COPY	.593***	.584***	.548***	.664***	.590***	.576***

RQ2: Table 4. Measurement Invariance

Model	χ^2/df	CFI	TLI	RMSEA	SRMR	Model	ΔCFI	$\Delta RMSEA$	Decision
Overall	6.872	0.965	0.959	0.064	0.029	-	-	-	-
Pre-service Teachers	1.778	0.956	0.949	0.053	0.042	-	-	-	-
In-service Teachers	2.971	0.972	0.967	0.065	0.027	-	-	-	-
M1: Configural Invariance	4.007	0.965	0.959	0.064	0.033	-	-	-	-
M2: Metric Invariance	3.970	0.965	0.960	0.063	0.035	M2-M1	0.000	-0.001	Accept
M3: Scalar Invariance	4.342	0.960	0.956	0.066	0.037	M3-M2	-0.005	0.003	Accept
M4: Strict Invariance	5.564	0.940	0.938	0.079	0.041	M4-M3	-0.020	0.013	Reject

RQ2: Table 5. Latent Mean Differences

	ΔM	P	Cohen's d	Size effect
AIK	-0.344	< 0.001	-0.517	medium
AIP	-0.462	< 0.001	-0.585	medium-to-large
AIA	-0.406	< 0.001	-0.478	near-medium
AIE	-0.163	< 0.001	-0.275	small
HCE	-0.213	< 0.001	-0.290	small-to-medium
PEN	-0.191	< 0.001	-0.265	small

① EFA (half-sample)

- Supported six-factor
- AIK_2 & AIK_4: cross-loadings on AIP
- AIA_1: misloaded on AIP
- Excluded in later CFA

② CFA (half-sample)

- Revised six-factor: good fit ($\chi^2/df = 4.229$, robust CFI = 0.965, robust TLI = 0.958, robust RMSEA = 0.068, SRMR = 0.030)
- Factors: inter-correlated
- Factor loadings: >0.75

④ Concurrent Validity (full sample)

- Adapted DigCompEdu Scale (Quast et al., 2025)
- Seven factors: ADMIN, LESS, TEACH, ASSES, EMPOW, COMP, COPY
- 0.5 < r < 0.7, all factors inter-correlated

➤ Final model (RQ1)

- Supported a six-factor, 23-item TAICS model
- excluded AIA_1 for the mainland Chinese context

⑤ Measurement (RQ2) Invariance

- Configural, metric & scalar invariance were supported

⑥ Latent Mean (RQ3) Differences

- pre-service teachers scored significantly higher than in-service teachers across all dimensions
- most substantial gaps: AI Pedagogy (d = -0.585), AI Knowledge (d = -0.517), AI Assessment (d = -0.478)

Method

Participants and Procedure

N = 2590 K-12 Teachers

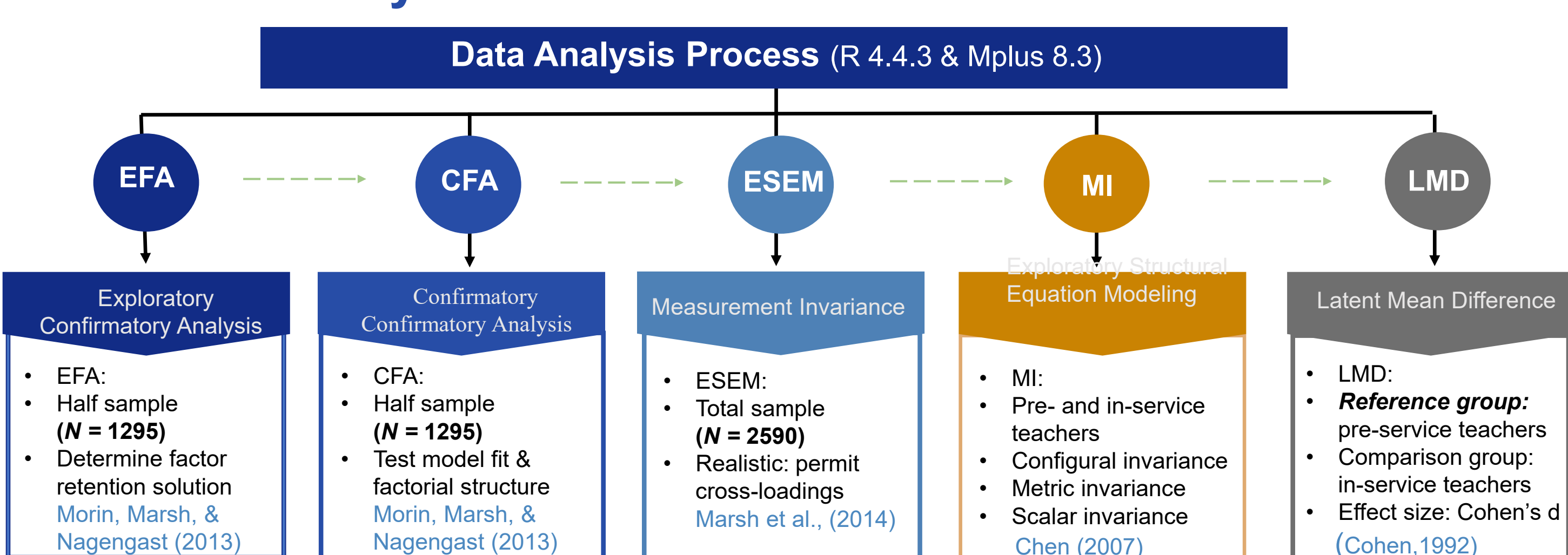
Pre-service teachers (n = 770, 29.73%)		In-service teachers (n = 1820, 70.27%)	
M = 19.92, SD = 1.33	81.04%, 18.96%	M = 37.59, SD = 9.13	75%, 25%

- Data collection: Wis 2024/25, online survey, convenient sampling
- Completion time: 15 mins

Instrument

- Teacher AI Competence Self-efficacy (TAICS) scale (Chiu et al., 2024)
- Six dimensions, 24 items, five-point Likert scale
- AI Knowledge (AIK) ($\Omega = 0.930$, 95%CI: [0.928, 0.932])
- AI Pedagogy (AIP) ($\Omega = 0.980$, 95%CI: [0.980, 0.981])
- AI Assessment (AIA) ($\Omega = 0.970$, 95CI: [0.970, 0.971])
- AI Ethics (AIE) ($\Omega = 0.947$, 95%CI: [0.946, 0.949])
- Human-Centered Education (HCE) ($\Omega = 0.946$, 95%CI: [0.944, 0.947])
- Professional Engagement (PEN) ($\Omega = 0.940$, 95%CI: [0.939, 0.941])
- McDonald Omega: excellent internal consistency (George & Mallery, 2003)
- Standardized forward-backward translation (Brislin, 1970)

Statistical Analysis



Discussion

- Culturally adapted scale: replicated 6-factor-structure, 23 items, removed item AIA_1
- Potential reasons: a) contextual & sampling differences b) model-related assumptions
- Latent mean differences: pre-service scored significantly higher than in-service teachers
- Potential reasons: a) exposure to generative AI tools b) motivational and career-stage differences c) AI-integrated teacher preparation programs

Scan for Reference!



Reference

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