



Australia Awards
Cambodia



A Thousand Stories and Counting!

The Role of Digital Financial Literacy in Facilitating Climate-Smart Agriculture (CSA) Adoption and Enhancing Food Security:

Evidence from a Repeat Cross-Sectional Household Survey in Cambodia

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Themes will be covered!

Stylize Facts

Smallholder climate vulnerability, digital financial deficits, and food insecurity parameters.

Methodology

Pooled cross-sectional data (N=816) across Cambodia provinces with nearest-neighbor imputation.

Research Gap/ Literature

Evidence disconnects and the limitation of typical simple binary digital ownership proxies.

Key Findings

Regressions of DFL inequalities, climate-smart drivers, and final structural equation models (Tables 4, 5, 6).

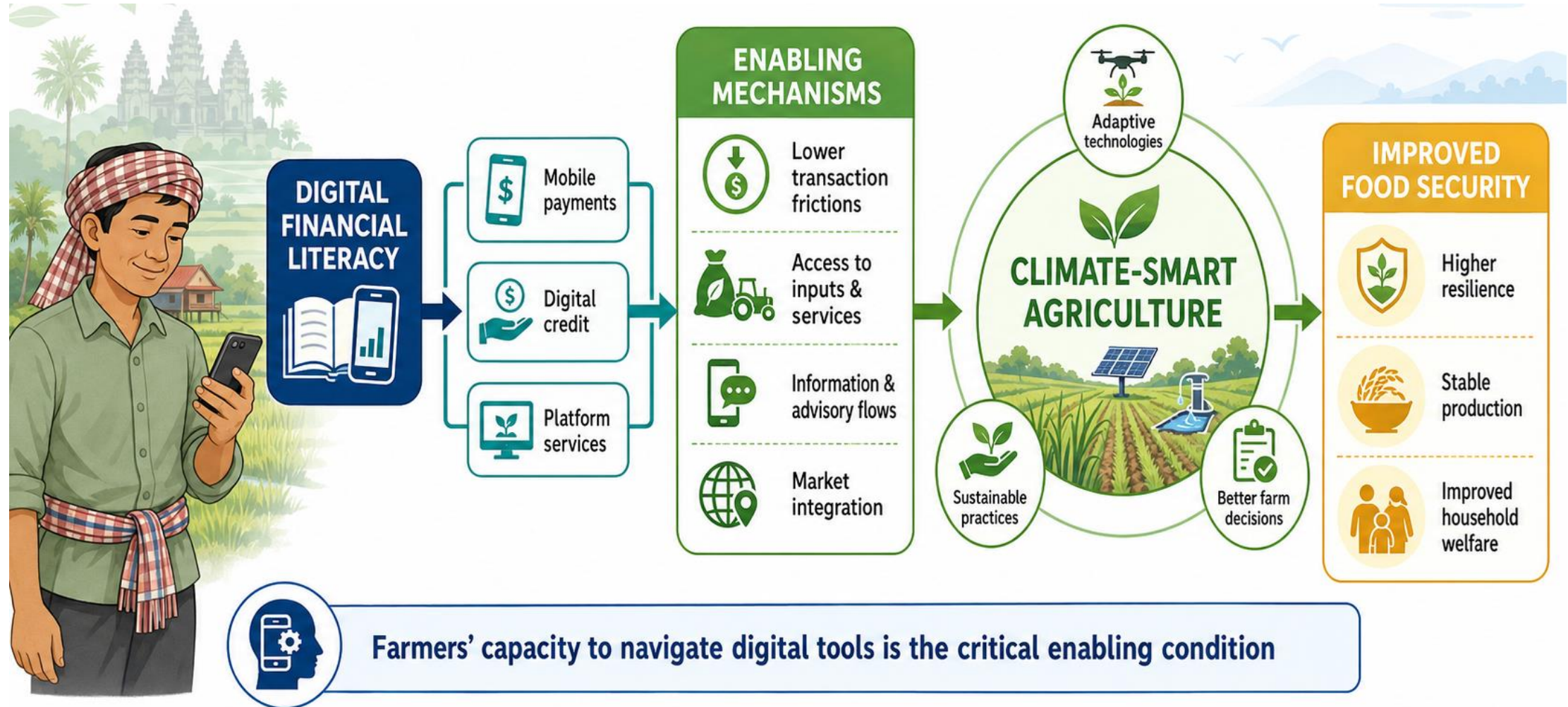
Research Questions

Mapping drivers of digital financial capability, agricultural adaptation, and direct mediation paths.

Policy & Conclusion

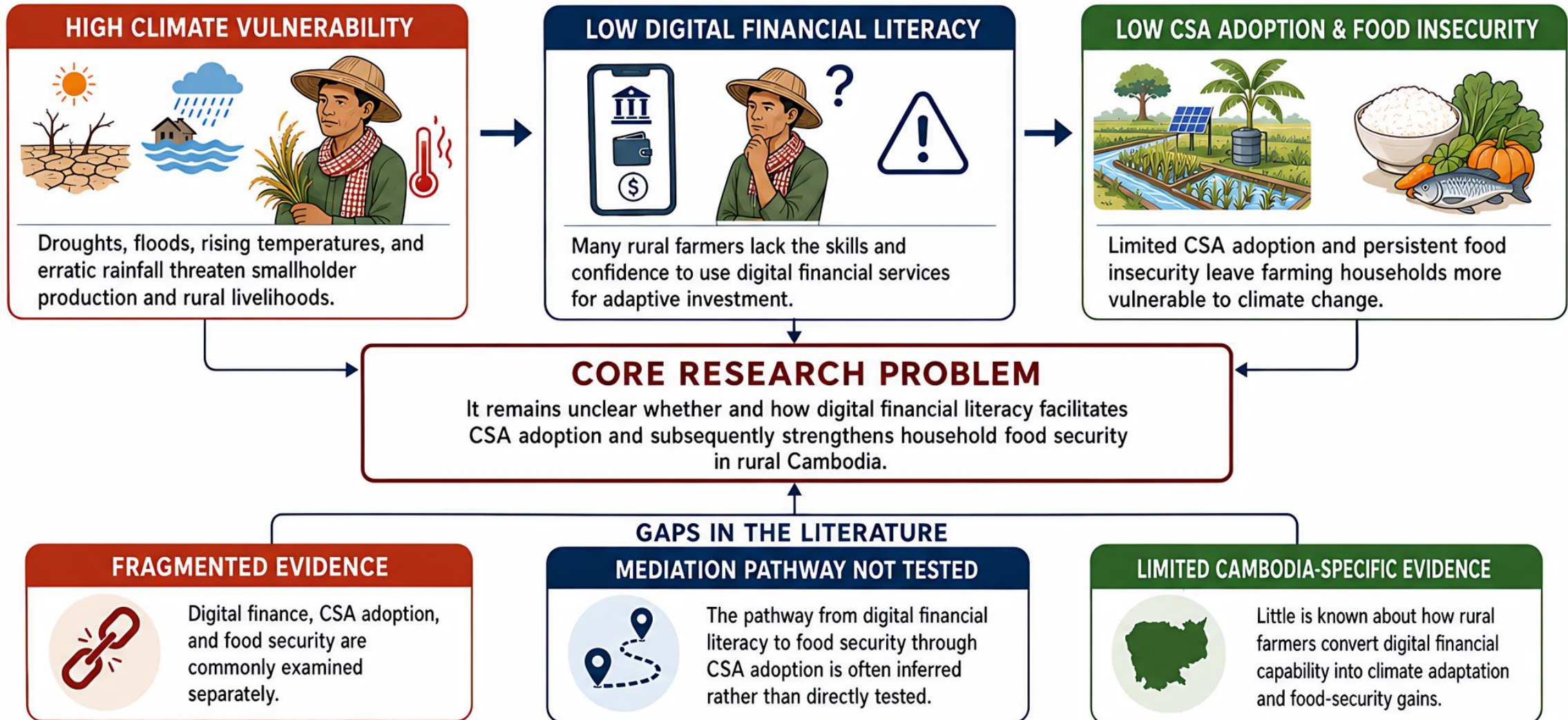
Targeted ministerial actions, addressing gender gaps, and structural social protection systems.

Why This Matters!!!

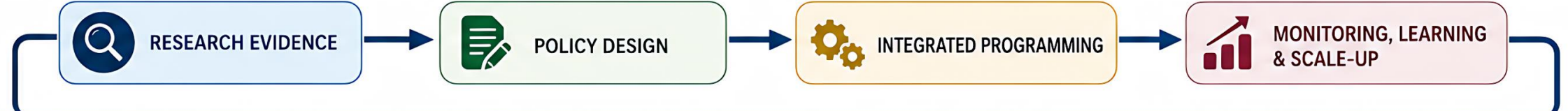


Sources: Yang et al. (2024); Mushi et al. (2022); Liu & Ren (2023); Asante et al. (2025); Hussain et al. (2025); Liu & Li (2025); Mao et al. (2024); Balyan et al. (2024); USAID (2021).

Stylize Facts !!!



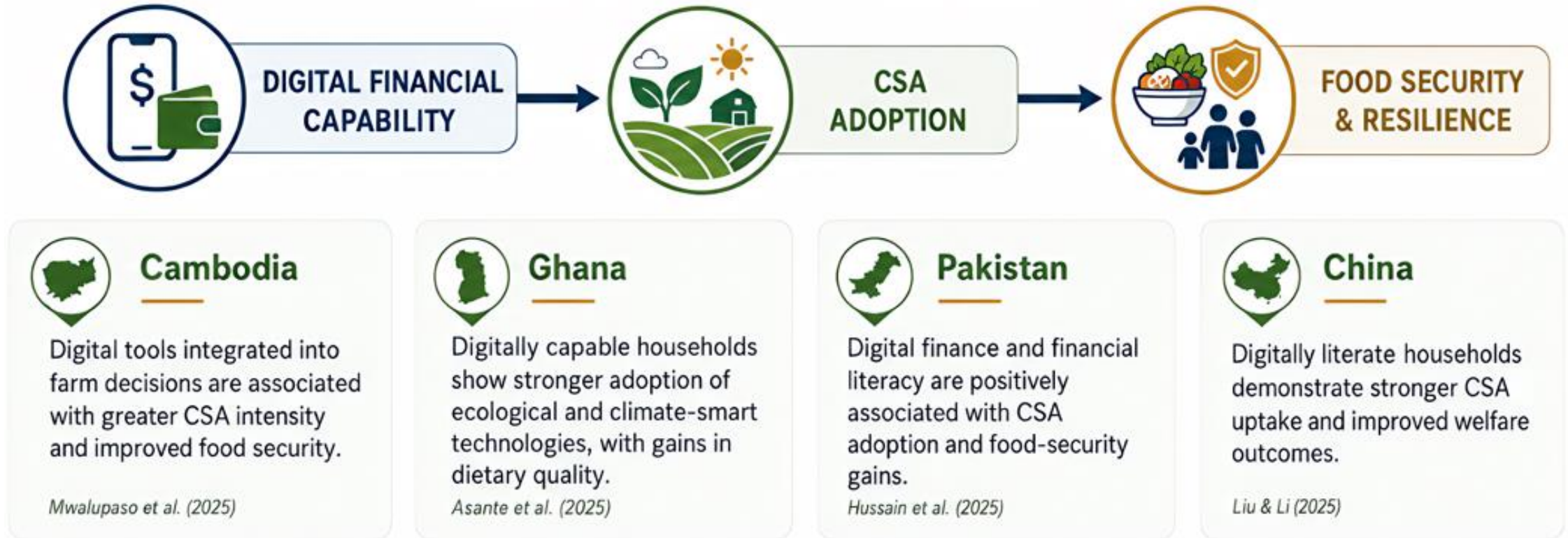
Contribution of the Study



POLICY VALUE: The study helps translate broad climate, digital, and inclusion commitments into practical interventions for Cambodia's smallholder farmers.

Literature Review

A CONSISTENT POSITIVE LINK ACROSS SETTINGS

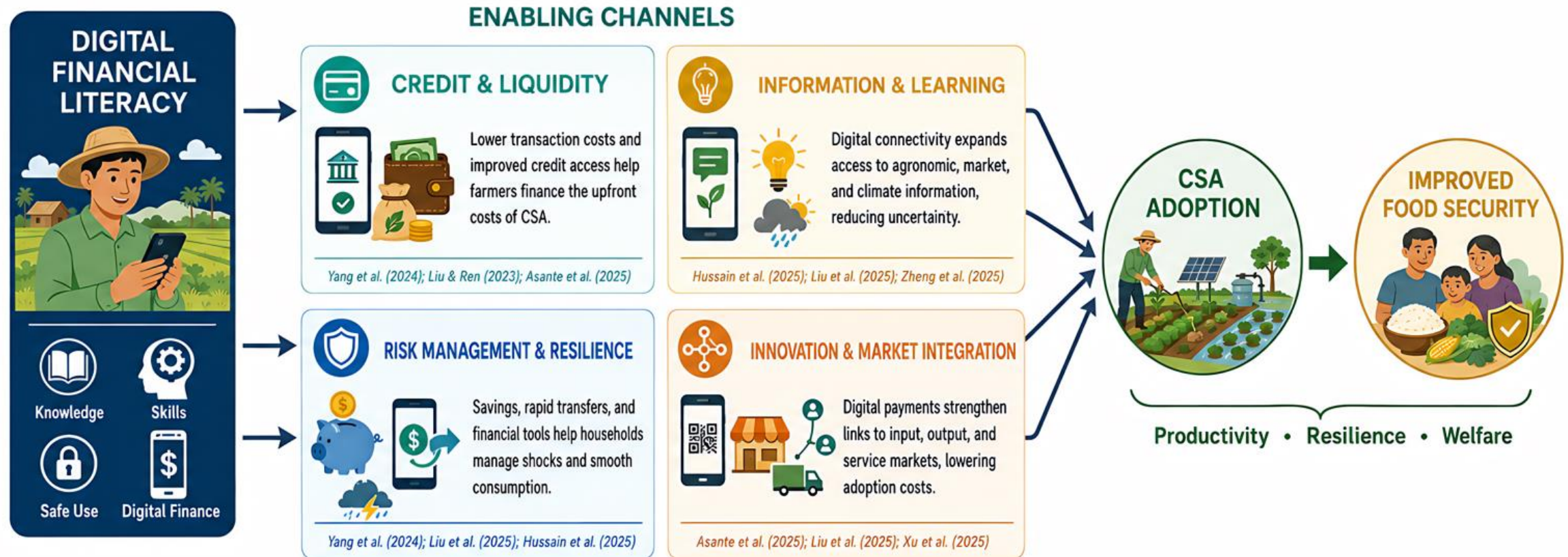


COMMON FINDING: Digital financial capability can support adaptive farm investment, CSA uptake, and household resilience.

Literature Review

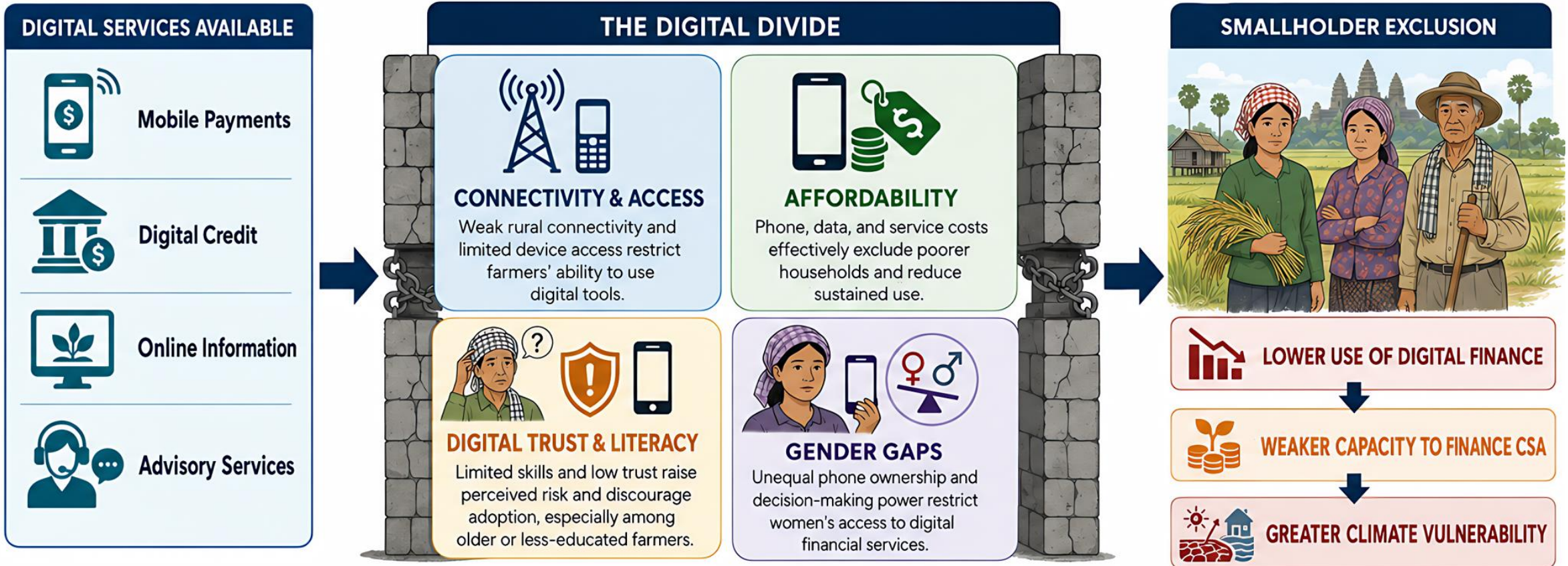
Mechanisms Linking DFL to CSA Adoption

Four interrelated channels connect digital financial literacy to climate-smart agriculture and food security.



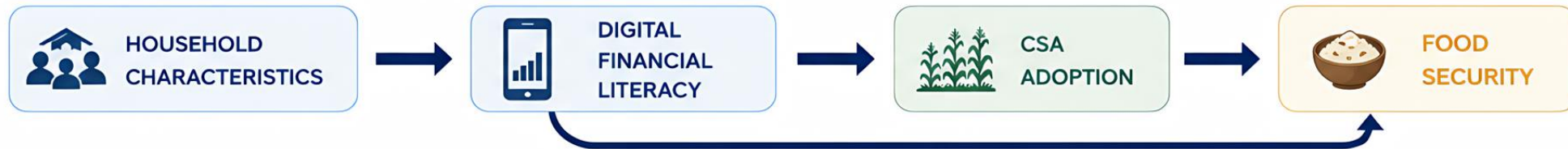
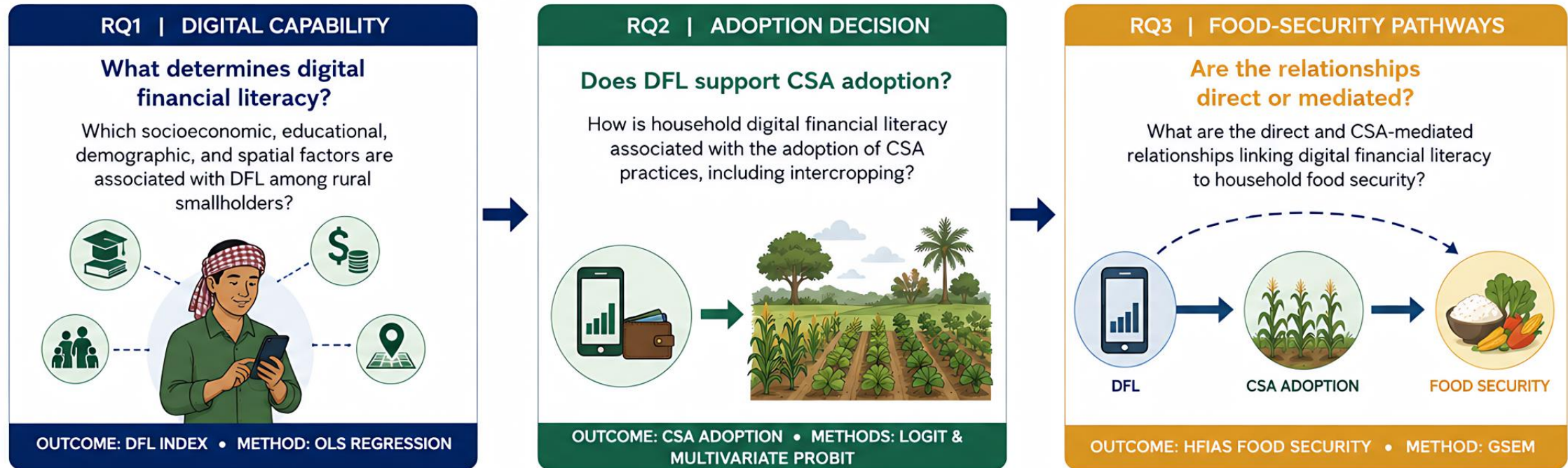
Literature Review

Structural barriers can exclude smallholders even when digital services nominally exist.



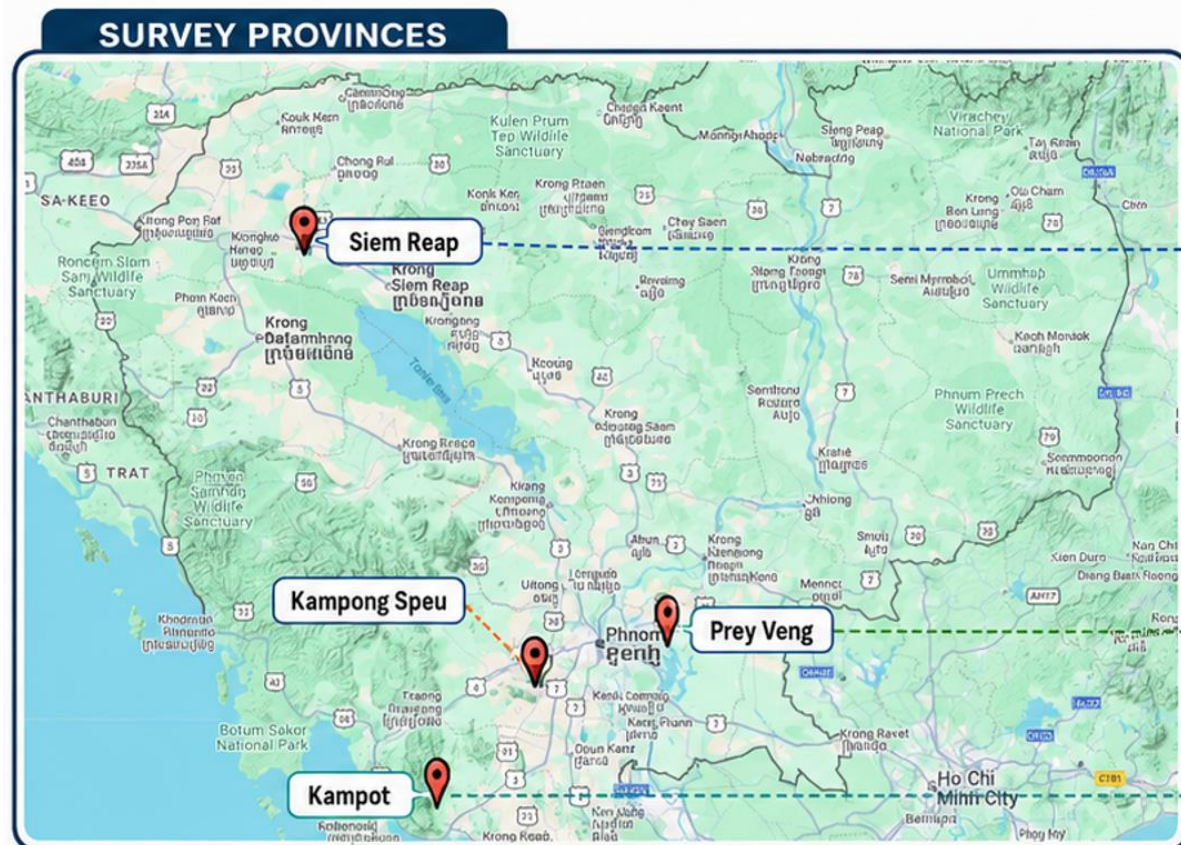
Research Questions and Empirical Strategies

Questions mapped to outcomes and analytical methods



The empirical design progresses from determinants, to adoption, to direct and mediated food-security relationships.

Study Locations & Scope



SIEM REAP TONLE SAP BASIN

Agricultural systems exposed to seasonal flood pulses.



PREY VENG MEKONG FLOODPLAIN

Intensive lowland rice production across fertile floodplain ecosystems.



KAMPONG SPEU DROUGHT-PRONE UPLANDS

Upland farming systems with low-moisture-retention soils.



KAMPOT SOUTHERN COASTAL ZONE

Coastal agriculture exposed to salinity and seawater intrusion.



Study sites span distinct hydrological, climatic, and agricultural conditions.



FOUR PROVINCES



816 FARMING HOUSEHOLDS



REPEAT CROSS-SECTIONS: 2023 & 2025

Sample distribution 2023 & 2025

Repeat Cross-Sectional Household Surveys



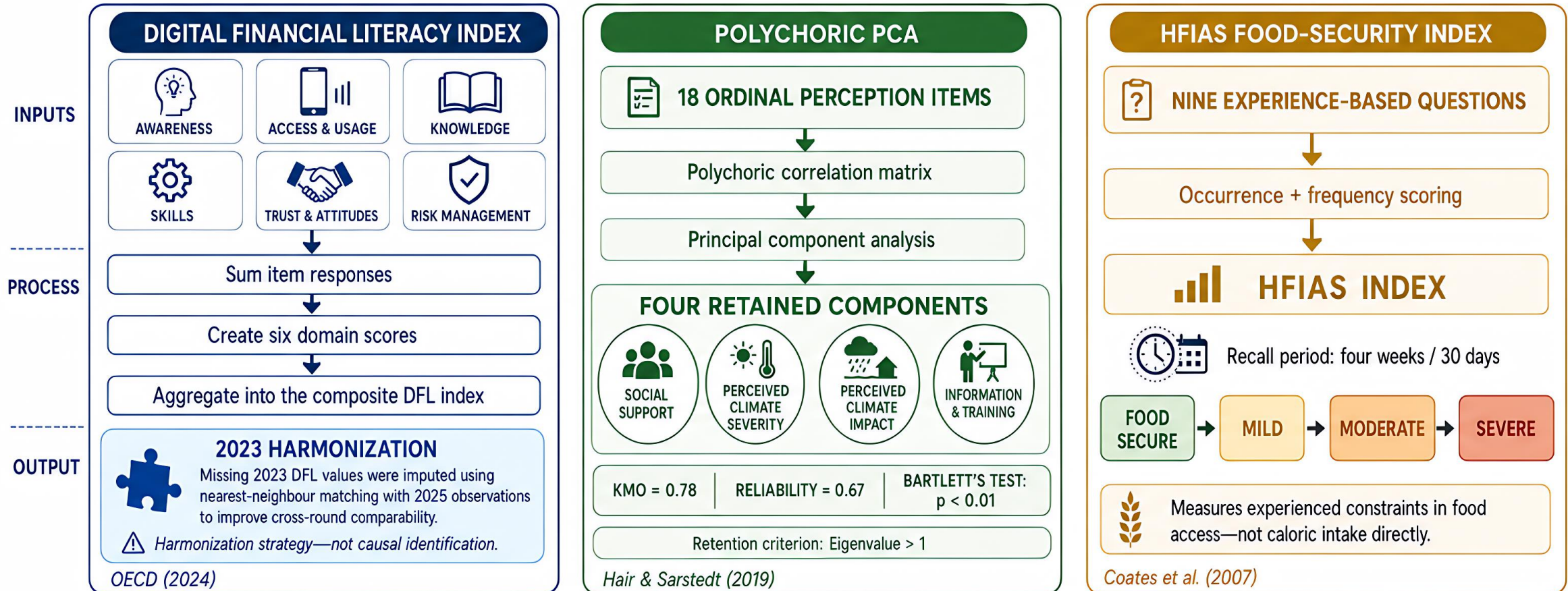
The same four provinces were surveyed in both rounds, but the same households were not necessarily re-interviewed.

PROVINCE-LEVEL SAMPLE COMPOSITION

Values shown as Total | Mono-cropping | Intercropping

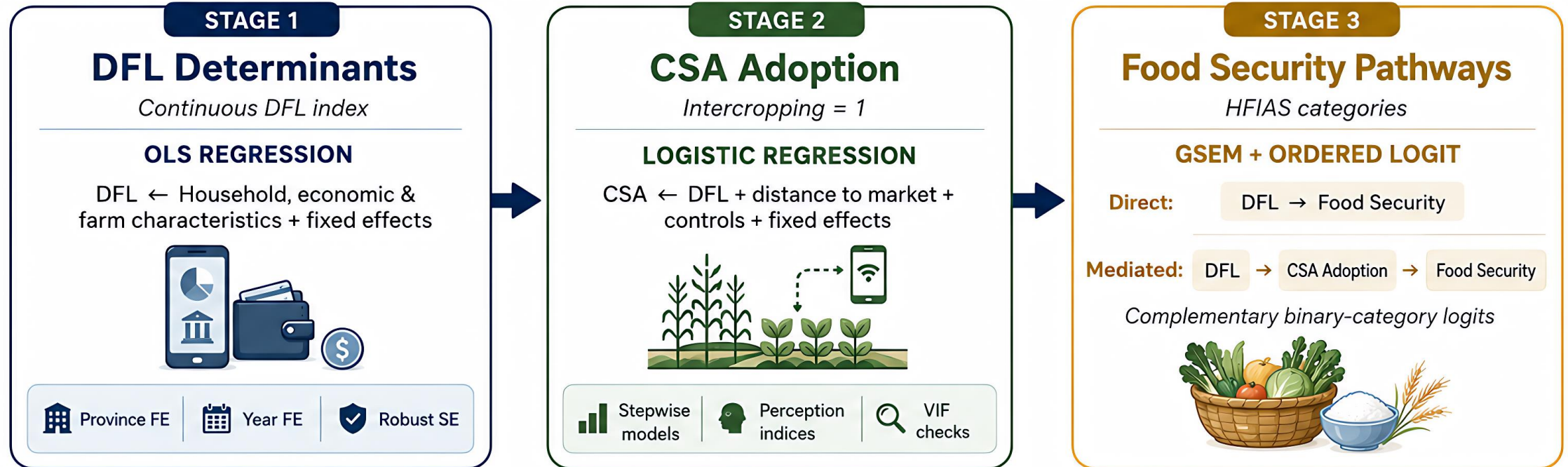
 KAMPONG SPEU 2023: 88 80 8 2025: 94 82 12 POOLED TOTAL: 182	 KAMPOT 2023: 76 63 13 2025: 86 63 23 POOLED TOTAL: 162	 PREY VENG 2023: 141 73 68 2025: 130 127 3 POOLED TOTAL: 271	 SIEM REAP 2023: 86 66 20 2025: 115 108 7 POOLED TOTAL: 201
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Measuring DFL, Climate change, and Food Security



MEASUREMENT LOGIC: Capability → Perceptions and enabling conditions → Experienced food-access outcomes

Empirical Approaches



COMMON COVARIATES



Age & gender



Household size & education



Log expenditure



Farm size & land plots



Association membership



Province and survey-year fixed effects

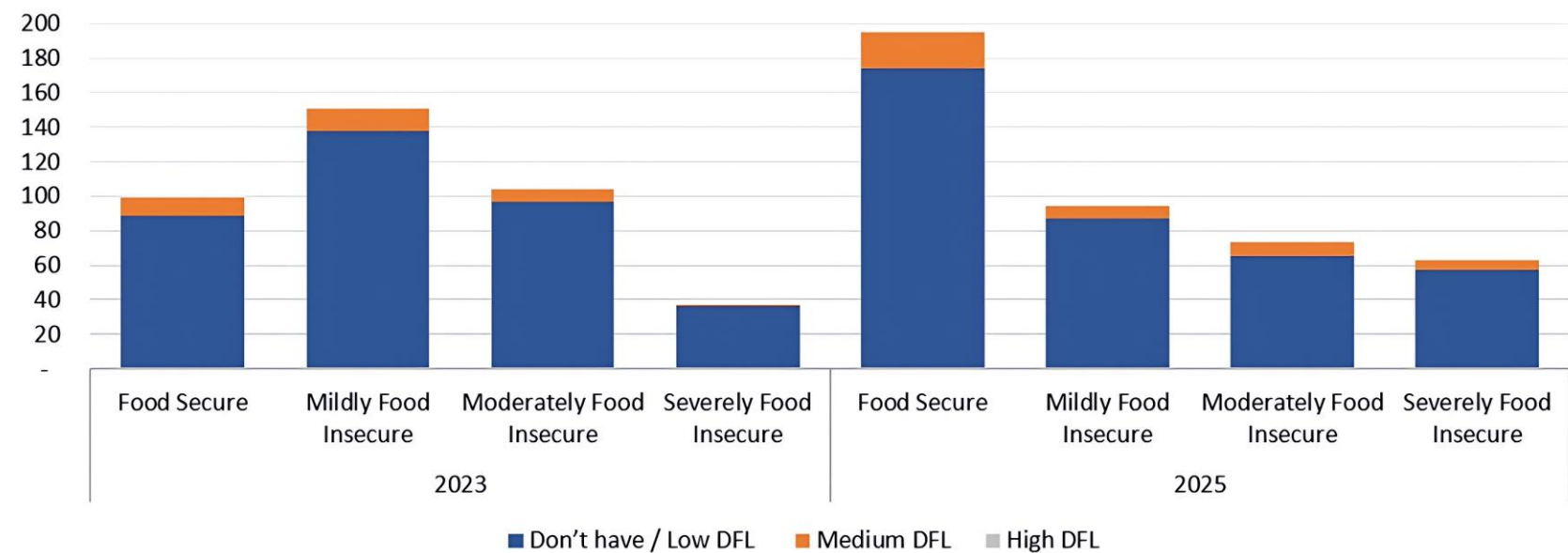


INTERPRETATION: Associations—not strictly causal effects—because unobserved selection may remain.

Rich controls, fixed effects, and explicit pathway modelling strengthen internal validity. Repeat cross-sections capture population-level change between 2023 and 2025.

DFL and Food Security

FIGURE 1. Digital Financial Literacy by Household Food Security Status, 2023–2025



KEY TAKEAWAY (DESCRIPTIVE)



Digital financial literacy remains generally low among smallholder farmers across all food-security categories in both 2023 and 2025.

Households that are food secure show relatively higher levels of medium DFL, especially in 2025.

Findings- Determinants of Digital Financial Literacy

VARIABLES	Model (1)	Model (2)	Model (3)	Model (4)
HH age	-0.255*** (0.036)			-0.256*** (0.037)
HH gender (1=Female)	-3.246*** (0.974)			-3.104*** (0.992)
HH size	-0.536*** (0.202)			-0.485** (0.227)
HH primary educ	1.162 (1.035)			1.126 (1.024)
HH lower secondary educ	4.110*** (1.347)			3.879*** (1.339)
HH upper secondary / Higher educ	6.480*** (1.851)			5.907*** (1.840)
Economic dependency ratio		1.491* (0.793)		0.471 (0.745)
Expenditure (log)		-0.334 (0.835)		-0.409 (0.891)
Farm size (log)		3.859** (1.848)		3.101* (1.867)
Member of association		3.745** (1.651)		3.435** (1.555)
Number of land plots		-0.488 (0.309)		-0.462 (0.333)
Perceive climate impact			-1.163** (0.523)	-0.687 (0.493)
Perceive inform			1.678* (0.902)	1.471* (0.881)
Perceive support			-0.029 (0.441)	0.111 (0.426)
Perceive climate severity			0.772 (0.632)	1.016* (0.593)
Constant	35.467*** (2.702)	21.201*** (4.876)	17.229*** (3.124)	32.399*** (5.930)
Observations	816	816	816	816
R-squared	0.199	0.115	0.108	0.217
Province FE	YES	YES	YES	YES
Year FE	YES	YES	YES	YES
Mean VIF	1.68	1.71	1.60	1.84

Note: OLS regression is used for the estimation. Dependent variable: Digital Financial Literacy Index (DFL). / Robust standard errors in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

HIGHER DFL ↑

Positive significant correlates



Upper secondary / higher education

+5.907***



Lower secondary education

+3.879***



Association membership

+3.435**



Farm size (log)

+3.101*



Access to climate information

+1.471*



Perceived climate severity

+1.016*

LOWER DFL ↓

Negative significant correlates



Female household head

-3.104***



Household size

-0.485**



Household-head age

-0.256***



Education gradient

The DFL advantage rises sharply beyond primary education.



Gender gap

Female-headed households record a substantially lower DFL index, holding other factors constant.



Institutions matter

Association membership and access to climate information are linked to stronger DFL.



Education is the strongest positive correlate of DFL, while older age, female household headship, and larger households are associated with lower DFL.

Findings- Determinants of CSA (Intercropping)

VARIABLES	Model (1)	Model (2)	Model (3)	Full Model (4)
DFL Index	0.013 (0.008)	0.012 (0.008)	0.013* (0.008)	0.017** (0.009)
HH distance to market	-0.035 (0.026)	-0.040 (0.026)	-0.034 (0.026)	-0.032 (0.026)
HH age	0.026*** (0.009)			0.017* (0.010)
HH gender (1=Female)	-0.268 (0.233)			-0.307 (0.232)
HH marital status	-0.195 (0.346)			-0.115 (0.353)
HH size	-0.047 (0.053)			0.008 (0.060)
HH primary educ	-0.328 (0.315)			-0.530* (0.321)
HH lower secondary educ	0.062 (0.370)			-0.129 (0.374)
HH upper secondary / Higher educ	0.336 (0.411)			0.137 (0.429)
Economic dependency ratio		-0.343** (0.171)		-0.226 (0.182)
Expenditure (log)		-0.326* (0.191)		-0.370* (0.216)
Farm size (log)		-0.394 (0.392)		-0.470 (0.413)
Number of land plots		0.181*** (0.067)		0.193*** (0.073)
Access to pipe water		0.031 (0.230)		0.043 (0.245)
Member of association		0.016 (0.305)		-0.222 (0.333)
Perceive climate impact			0.419*** (0.110)	0.455*** (0.116)
Perceive inform			0.461** (0.182)	0.468** (0.190)
Perceive support			0.110 (0.100)	0.096 (0.100)
Perceive climate severity			-0.234 (0.147)	-0.222 (0.155)
Constant	-2.351*** (0.782)	-0.188 (1.104)	-1.760** (0.771)	-0.520 (1.603)
Observations	816	816	816	816
Province FE	YES	YES	YES	YES
Year FE	YES	YES	YES	YES
Mean VIF	1.75	1.83	1.72	1.88
Log Pseudolikelihood	-354.397	-354.968	-351.447	-337.428

Note: Logistic regression is used for the estimation. Dependent variable: Inter-Cropping. Robust standard errors in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$



Higher DFL is positively associated with intercropping adoption

DFL Index: +0.017**

Logit coefficient; holding observed characteristics and fixed effects constant



POSITIVE SIGNIFICANT CORRELATES



Climate information

+0.468**

Weak significance:
 $p < 0.10$



Perceived climate impact

+0.455***

Weak significance:
 $p < 0.10$



Number of land plots

+0.193***

Weak significance:
 $p < 0.10$



Household-head age

+0.017*

Weak significance:
 $p < 0.10$



Digital capability matters

The positive DFL coefficient supports the proposed DFL → CSA pathway.



Climate awareness matters

Information access and perceived climate impacts are linked to intercropping uptake.



Farm structure matters

Households operating more land plots are more likely to diversify through intercropping.

N = 816

Province FE: Yes

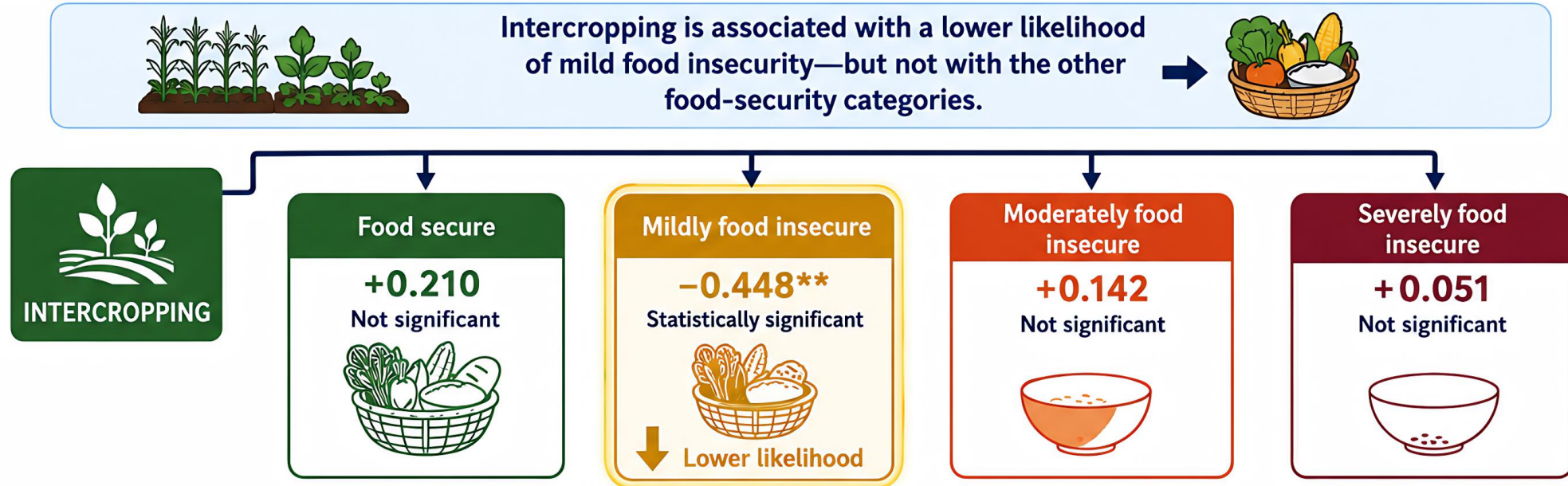
Year FE: Yes

Robust SE

Mean VIF = 1.88

Log pseudolikelihood = -337.428

Findings- Effects of DFL and CSA Adoption on Food Security using GSEM



MEDIATION EVIDENCE: PARTIAL — CSA adoption is linked to reduced mild food insecurity, but the pathway is not statistically supported for food security or more severe deprivation.



Education

Primary education is negatively associated with moderate (-0.519^{**}) and severe insecurity (-0.771^{**}); higher education is also negative for moderate and severe categories.



Farm structure

More land plots are associated with food security ($+0.197^{***}$) and lower moderate insecurity (-0.312^{***}), while larger farm size is positively associated with moderate insecurity ($+0.880^{***}$).



Climate perceptions

Perceived climate impact and severity are positively associated with food security but negatively associated with mild insecurity.



N = 816 per equation



Province FE: Yes



Year FE: Yes



Robust SE



Separate binary logits



Coefficients are log-odds associations, not marginal effects or causal estimates.
Each HFIAS category is estimated separately; results do not constitute a single ordered-outcome comparison.

*** $p < 0.01$ ** $p < 0.05$ * $p < 0.10$

Policy Implications



Move from stand-alone training to an integrated, farmer-centred delivery model.

1



Make DFL Practical

Design training around receiving payments, purchasing inputs, saving for lean seasons, managing risk, and preventing fraud.



Use cases + safe use

2



Integrate DFL with CSA Services

Deliver digital financial literacy alongside CSA training, extension advice, demonstration farms, and peer learning.



One coordinated service package

3



Close the Gender Gap

Provide women farmers and female-headed households with tailored training, phone-based support, and access to farmer groups.



Gender-responsive delivery

4



Strengthen Learning Systems

Expand agricultural extension, practical demonstrations, farmer training, and peer-to-peer learning to support CSA adoption.



Learn by seeing and doing

5



Turn Climate Awareness into Action

Use community-based support to convert climate information and risk awareness into feasible CSA practices.



Information → local action



DESIGN WITH FARMERS



DELIVER THROUGH TRUSTED LOCAL CHANNELS



TRACK ADOPTION AND FOOD-SECURITY OUTCOMES



EVIDENCE LINK: Recommendations respond to the observed DFL gender gap, the positive DFL–CSA relationship, and the importance of climate information, farm structure, and community support.

Questions & Answers

Dr. OUM Chan Mono

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