







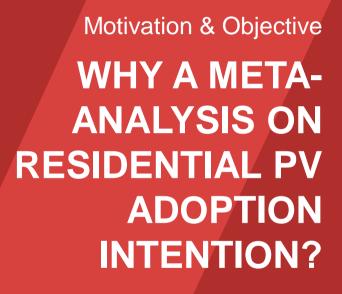
DETERMINANTS OF RESIDENTIAL PHOTOVOLTAIC ADOPTION INTENTION - A META-ANALYSIS

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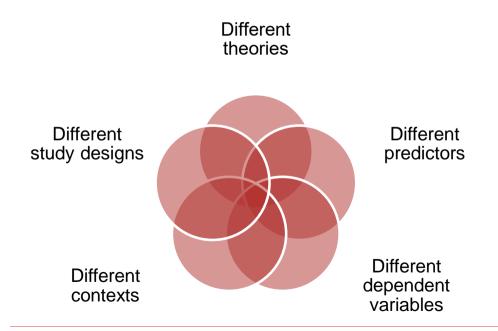


Institute for Infrastructure and Resource Management Energy Management and Sustainability Prof. Dr. Thomas Bruckner



MOTIVATION

META ANALYSES PROVIDE A MECHANISM TO OBJECTIVELY SYNTHESIZE DATA ACROSS STUDIES



- In 173 studies on residential PV adoption, [1] find 333 predictors
- Single studies are subject to issues related to
 - Statistical power
 - Validity and reliability
 - Contextuality
- → Single studies provide no solid base for conclusions about effects of predictors

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OBJECTIVE

REVEAL PATTERNS OF RELATIONSHIPS AMONG PREDICTORS AND RESIDENTIAL PV ADOPTION

- Determine point estimates of relationships between predictors and intention
- Assess the suitability of an (extended) Theory of Planned Behavior framework
- Derive implications to enhance future aggregation of scientific evidence

Methodology

WHAT DID WE DO?



METHODOLOGY

USE MASEM TO SYNTHESIZE AND ANALYZE THE SYSTEMATICALLY SELECTED LITERATURE

Meta-analytic structural equation modeling (MASEM) ^[2, 3, 4]:

- 1. Literature research
- 2. Literature selection
- 3. Code literature
- Pool correlations (Random effect model, inverse variance weighing, REML method)
- 5. Estimate SEMs

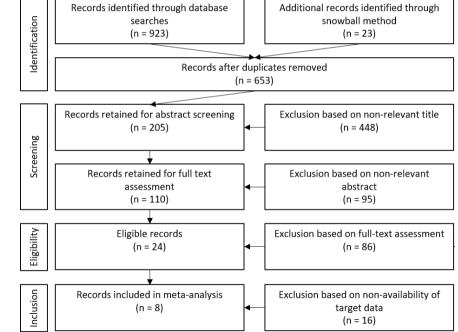


Figure 1: Flow chart describing literature selection. Source: Own illustration

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5. Estimate SEMs

	í A	B	С	D	E	F	G	Н		J	K	L	М
1	no	author	year	country	n	envcon	novseek	bar	ben	sn	gen	edu	inc
2		1 Sun et al.	2020	Taiwan	300	0,631689204	0,75		0,765247931				
3		2 Claudy et al.	2013	Ireland	254			-0,266966639	0,340293929				
4	1	8 Rai and Beck	2015	Texas	522	0,187	0,324		0,349	0,416025147			
5	4	4 Chen	2014	Taiwan	203	0,639736524	0,4						
6	3	5 Arroyo and Carre	2019	Mexico	72	0,046		-0,153			-0,028	0,265	0,442
7	(5 Parkins et al.	2018	Canada	2065	0,048				0,101	-0,009391759	0,011665175	-0,002107285
8		7 Aziz et al.	2017	Malaysia	211	0,354		0,097	0,562	0,397		-0,048	0,142
9	1	8 Wolske et al.	2017	USA	904	0,324442842	0,32439929	-0,116964222	0,530695705	0,384114384			

Figure 2: Input table for correlations between intention and predictors. Source: Own calculations

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Results

POOLED CORRELATION TABLE AND SEM's

POOLED CORRELATION TABLE

	INT	\mathbf{EC}	NS	BA	BE	\mathbf{SN}	GEN	\mathbf{EDU}
EC	.343**							
	(.382), N=7							
\mathbf{NS}	.475**	.445**						
	(.46), N=4	(.187), N=4						
$\mathbf{B}\mathbf{A}$	111	172*	015					
	(.306), N=4	(.397), N=3	(.13), N=1					
\mathbf{BE}	.53**	.693**	.636**	185**				
	(.334), N=5	(.366), N=4	(.458), N=3	(.102), N=3				
\mathbf{SN}	.326**	.283**	.504**	104	.491**			
	(.291), N=4	(.293), N=4	(.752), N=2	(.528), N=2	(.429), N=3			
GEN	01	.049**	0	038	0	059**		
	(.084), N=2	(.082), N=2	(0), N=0	(.463), N=1	(0), N=0	(.086), N=1		
\mathbf{EDU}	.046	.047	0	035	006	.068**	086**	
	(.287), N=3	(.126), N=3	(0), N=0	(.234), N=2	(.27), N=1	(.082), N=2	(.084), N=2	
INC	.183	.15**	0	.002	.085	.037	096**	.194
	(.499), N=3	(.219), N=3	(0), N=0	(.333), N=2	(.687), N=1	(.189), N=2	(.084), N=2	(.508), N=3

Upper number: Pearson's r with significance level (*: p<.1; **: p<.05); Number in brackets: Width of 95% CI; N: Number of studies INT Intention; EC Environmental concern; NS Novelty Seeking; BA Barriers; BE Benefits; SN subjective norm; GEN Gender; EDU Education; INC Income

Figure 3: Pooled correlation table. Source: Own calculations

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POOLED CORRELATION TABLE

DETERMINE POINT ESTIMATES OF RELATIONSHIPS BETWEEN PREDICTORS AND INTENTION

	INT	EC	NS	BA	BE	SN	GEN	EDU		
\mathbf{EC}	.343**									
	(.382), N=7								_	Strongest correlation
\mathbf{NS}	.475**	.445**								between intention and
		(.187), N=4								benefits
BA		172*	015							
BE	(.306), N=4 .53**	(.397), N=3 .693**	(.13), N=1 .636**	185**					-	No correlation between intention and income and
	(.334), N=5	(.366), N=4	(.458), N=3	(.102), N=3						
\mathbf{SN}	.326**	.283**	.504**	104	.491**					barriers
	(.291), N=4	(.293), N=4	(.752), N=2	(.528), N=2	(.429), N=3				_	Benefits strongly correlated
GEN	01	.049**	0	038	0	059**				with environmental
	(.084), N=2	$(.082), N{=}2$	(0), N=0	(.463), N=1	(0), N=0	(.086), N=1				
\mathbf{EDU}	.046	.047	0	035	006	.068**	086**			concern, novelty seeking
	(.287), N=3	(.126), N=3	(0), N=0	(.234), N=2	(.27), N=1	(.082), N=2	(.084), N=2			and subjective norm
INC	.183	.15**	0	.002	.085	.037	096**	.194		
	(.499), N=3	(.219), N=3	(0), N=0	(.333), N=2	(.687), N=1	(.189), N=2	(.084), N=2	(.508), N=3		

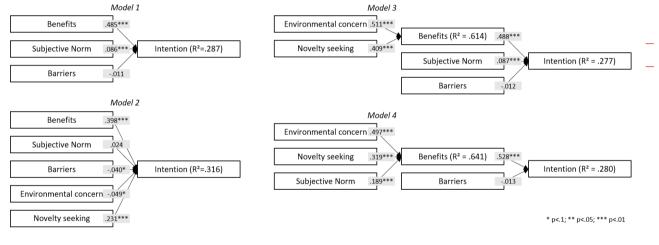
Upper number: Pearson's r with significance level (*: p<.1; **: p<.05); Number in brackets: Width of 95% CI; N: Number of studies

Figure 3: Pooled correlation table. Source: Own calculations

INT Intention; EC Environmental concern; NS Novelty Seeking; BA Barriers; BE Benefits; SN subjective norm; GEN Gender; EDU Education; INC Income

RESULTS OF STRUCTURAL EQUATION MODELLING

ASSESS THE SUITABILITY OF AN (EXTENDED) THEORY OF PLANNED BEHAVIOR FRAMEWORK



- Useful to limited extent
- Suggestion to use a modified version of TPB:
 - attitudes operationalized as benefits
 - attitudes explained by environmental concern, novelty seeking, and subjective norms

	N	Chi ² (LR test model vs. saturated)	df	CFI	TLI	RMSEA, 90% CI	SRMR	AIC	BIC	CD
Model 1	1640	0	0	1.000	1.000	.000 [0]	.000	17,556.435	17,578.044	.287
Model 2	1714	0	0	1.000	1.000	.000 [0]	.000	25,789.611	25,822.290	.316
Model 3	1714	222.595 (p = .000)	4	.910	.797	.179 [.159 – .199]	.048	26,014.206	26,052.332	.618
Model 4	1714	113.798 (p = .000)	4	.964	.920	.127 [.107 – .147]	.039	25,905.409	25,943.535	.641

Figure 4: Results of structural equation modeling. Source: Own calculations

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Discussion & Conclusion

PROBLEMS, PLAUSIBILITY AND PROPOSITIONS

PROBLEMS

DUE TO HETEROGENEOUS LITERATURE BODY, SYNTHETIZATION RESULTS ARE LIMITED

- Small study sample
- No consistent use of predictors
- Different contexts _

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 \rightarrow Limited informative value

PLAUSIBILITY

RELATIONS BETWEEN INCOME, BARRIERS AND INTENTION MUST BE ANALYZED MORE THOROUGHLY

1. Correlation Income – Intention

[7] and [8] both find positive effects of income on intention

- Considerations preceding intentions less concrete?^[9]
- Role of policy schemes? [10, 11]
- Biased samples?

2. Correlation Barriers – Intention

- Smaller than correlation in metaanalyses on pro-environmental behavior ^[3, 4]
- Reasons for outlier can't be analyzed due to missing information

PROPOSITIONS – FOR FUTURE RESEARCH

DERIVE IMPLICATIONS TO ENHANCE FUTURE AGGREGATION OF SCIENTIFIC EVIDENCE

Systematization of research is crucial for more meaningful derivations:

- Use consistent predictors
- Use consistent measures for adoption
- Collect contextual variables
- Comply with reporting standards

PROPOSITIONS – FOR POLICY-MAKERS

- Enhance benefits instead of reducing barriers
- Focus on environmental benefits and innovativeness
- Innovativeness plays particularly large role in regions with low diffusion rates
- Consider tailoring strategies to consumer segments of like-minded people



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THANK YOU FOR YOUR ATTENTION!

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