

Technological Change and Learning Curves In the Context of the TranSust.Scan Network

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• Workpackage 2, Task 2:

Learning curves have been suggested as meaningful presentations of technological change in economyenvironment models, and their strengths and weaknesses in the context of a sustainability perspective are to be investigated

⇒ literature review on learning curves and technological change



Paper addresses theory of technological change in a broader context

- o Learning Curve Concept
- o Endogenous Growth Theory
- o Innovation Theory
- o Theory of Technology Diffusion

A special focus remains on learning curves:

- due to description in Workpackage 2, Task 2 and
- the prevalence of the concept in empirical modeling



Learning curves capture various learning effects besides learning by doing :

- o Learning by using
- o Learning by interacting
- o Learning from inter-industry spill-overs
- o Learning by researching
- o Learning by exploring

Extension of the concept of learning curves:

The recently developed two-factor learning curves that can capture two types of learning at the same time (mostly learning by doing and learning by researching)



- Bottom up concepts show a pronounced overlapping with respect to the drivers of technological change
 - o Focus on particular drivers mainly determines the specific categorisation:
 - Learning curves primarily address experience
 - Innovation theory mainly addresses R&D
 - Diffusion theory primarily addresses adoption of technologies
- The drivers of technological change also establish a link to endogenous growth theory (top down perspective).
 - o Spill-overs are a main driver of technological change
 - R&D is considered as contributing to the accumulation of the knowledge capital stock and to economic growth

WFOE Links between the theoretical approaches





- Review of empirical modelling of ETC
 - Literature survey on learning curves and energy technologies
 - o Review of the TranSust.Scan models that explicitly tackle endogenous technological change
- Basis of TranSust.Scan review
 - o Questionnaires on technological change
 - o Model description provided

WIFO Characterisation of TranSust.Scan models (I)

 6 out of the 12 models are explicitly dealing with endogenous technological change

Model	Institution	ETC	No ETC	Description
DART	IfW	x		Global multi-regional, multi-sectoral recursive dynamic model
DEMETER	IVM-VU	x		Long-term integrated assessment model
GAIN	WIFO	x		Extended energy model
IMACLIM-S	SMASH-CIRED		x	Multisectoral macroeconomic general equilibrium framework
IMACLIM-R	SMASH-CIRED		x	Energy systems model
IMPEC	LIFEA		x	Multisectoral macro model
KLUM	University of Hamburg		x	Global agricultural land allocation model
MARKAL	ECN	x		Bottom-up energy system optimisation model
PACE	ZEW		x	Comparative-static multi-region, multisectoral CGE model
POLES	SMASH-CIRED			
WITCH	FEEM	x		Regional integrated assessment hard-link hybrid model
W8D	LIFEA	x		Macroeconomic model of the Polish economy

WIFO Characterisation of TranSust.Scan models (II)

- learning curves are the prevailing concept in empirical modelling in TranSust.Scan models
 - o 4 models use learning curves
 - GAIN and WITCH use a hybrid modelling approach
 GAIN: learning curves and innovation theory
 WITCH: learning curves and endogenous growth theory

Model ²	LC ¹	EGT ²	IT/DT ³	Model Description		
DART			x	top-down model		
DEMETER	x			top-down model that is calibrated to fit bottom-up descriptions of certain energy technologies		
GAIN	x		x	bottom-up		
MARKAL	x			bottom-up		
WITCH	x	x		top-down structure with a detailed description of the energy sector.		
 Learning Curves Endogenous Growth Theory Innovation / Diffusion Theory 						



- Review of theoretical and empirical modelling of ETC
 - o Focus on the links between various theoretical approaches
 - o Application in TranSust.Scan models
- LC predominant concept in empirical modelling
 Rather straight forward empirical implementation
 - Evidence on ex post observations and ex ante estimation match considerably well