





NUDGE has received funding from the European Union's Horizon 2020 Research and innovation programme under grant agreement No 957012.

Intervening me Softly - Modeling Nudging Interventions to Change Electric Vehicle User Preferences (6-161-22)

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¹Fraunhofer Institute for Systems & Innovation Research ISI, ² Delft University of Technology ECEEE Conference, 6-10/06/2022



Simulations of rational decision-making on flexibility provision by electric vehicles vs. intuitive decision-making

In everyday life:

Simulating everyday life in agent-based models:

Decision-making on providing flexibility from electric

vehicles based on... (Kahnemann 2003)

Rational choices – e.g., cost-minimazation	Cost-minimisation optimisation
• Intuitive choices – e.g., based on social interaction and provided information	Our paper: social network simulation
Incentives for providing more flexibility from electric	
vehicles	
For rational choices: monetary incentives	\mathbf{H}
For intuitive choices: nudging interventions	μ





Research question

• How do nudging interventions based on social interaction influence the flexibility provision by electric vehicles?

Research assumptions

- Flexibility provision by electric vehicles by minimizing the required final stage of charge (SoC)
- Nudging interventions based on social interaction in the network influence the SoC of the agents

Research objectives

- 1. Setting up the agent-based model based on a social network
- 2. Testing and demonstrating the model functionalities in two sub-steps:
 - 1. Influence of nudging interventions (and external factors) on the SoC (experiment)
 - 2. Influence of the agent settings on the results (sensitivity analysis)

Next steps: Parameterization based on results of NUDGE field experiments and model coupling

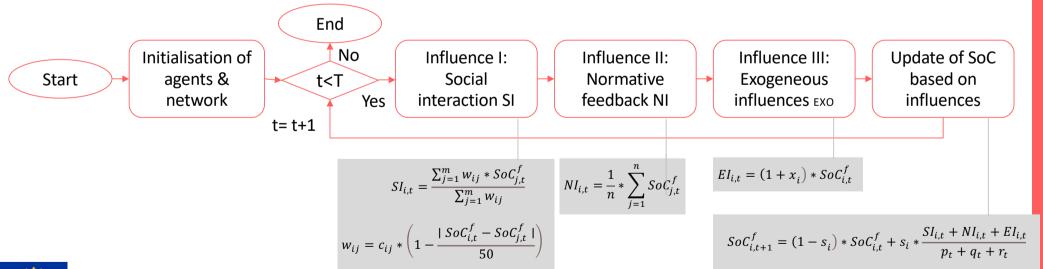




Setting up the agent-based model based on a social network

Nudging interventions based on social interaction extracted from the literature:

- Social interaction (influence I): Distributed information on <u>favorable and unfavorable behaviour</u> of the <u>own</u> <u>network (Zarei and Maghrebi 2020, Duggins 2014)</u> uncontrolled interventions, such as by information events
- **Normative feedback (influence II)**: Distributed information on <u>favorable behaviour</u> of the <u>entire network</u> (Friedkin 2001) selective interventions, such as by smart home systems or energy management apps



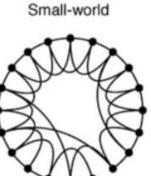


Experimental setup for testing the model

Nudging consumers towards energy efficiency through behavioural science

Scenario	Intervention / Influences
All (EXO + SI + NI)	All
EXO + SI	Social interaction + external influences
EXO	External influences
EXO + NI	Normative feedback + external influences

Parameter	Value
Number of agents	100
Number of simulation runs	100
Size of social contacts	12
Limit behaviour adjustment	18%
Simulation steps	52
Social network type	small-world



Agent initialization

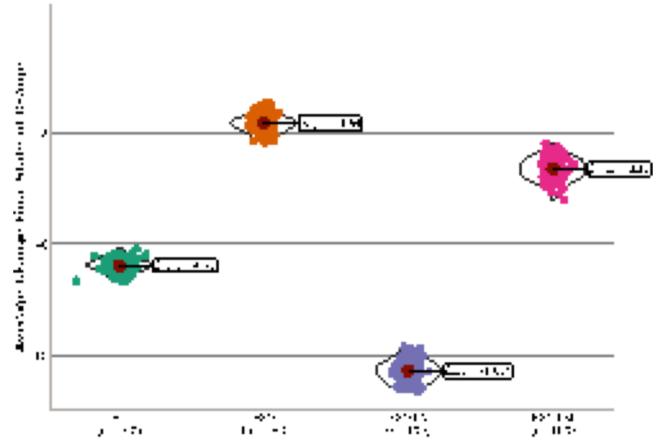
- EV battery capacity (75.7 kWh, 47.4 kWh, 26.3 kWh) (Gnann and Speth 2021)
- EV consumption (16-25 kWh/100 km)
- Distribution of initial SoC based on Quirós-Tortós et al. 2015
- Daily mobility demand of agents (= minimum technical SoC)





Experiment results: SoC reduction for all scenarios strong influence of exposure to unfavorable behavior

- EXO + N: Strongest average SoC reduction (-10.54 %) compared to initialisation
- All: Medium reduction (-6.76%)
- **EXO + SI**: Lower reduction (-3.28%)
- **EXO:** Also reduction due to high SoC at initialisation (-1.64%)
- All scenarios on average below the limit of behavioral adjustments (18%)



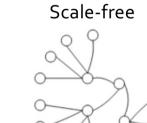


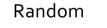
through behavioural science

Sensitivity analysis: Limit of behaviour adjustment impacts the results strongly

Parameter	Value in the experiment	Value in the sensitivity analysis	Impact on scenario "all"
Size of social contacts	12	4 – 22	Decreasing SoC change with increasing number of contacts
Limit behaviour adjustment	18%	5 – 50 %	Strong impact: 15% results in highest available flexibility
Social network type	small-world	random, regular, scale-free	Minor impact: max. o.3% smaller change for random network

Regular















- Creation of agent-based model based on social network, in which agents adapt their final SoC depending on their social interaction
- Normative feedback (=selective information of only favorable behavior of entire network) results in strongest reduction of SoC (=highest provision of flexibility)

Discussion and next steps

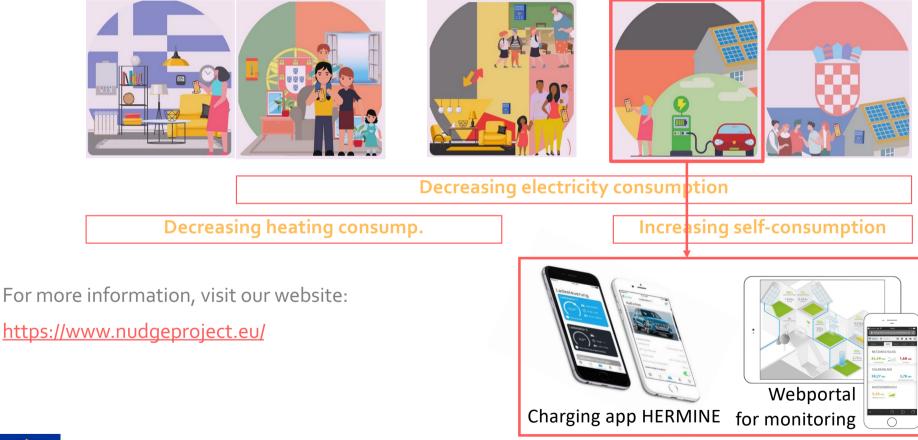
- Exposure to unfavorable behavior (boomerang effect) and limit of behavior adjustments influences the results strongly
- Parameterization (esp. w.r.t. adjustment limit and heterogenity of agents) and validation based on results of NUDGE field experiments
- Medium-term step: Model coupling with cost-minimisation model

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H2020 project NUDGE: 5 pilots testing the impact of nudging interventions on energy-saving behavior of households

Nudging consumers towards energy efficiency through behavioural science











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Thank you for your attention!

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Zelena Energetska Zadruga beegy

the energy manager







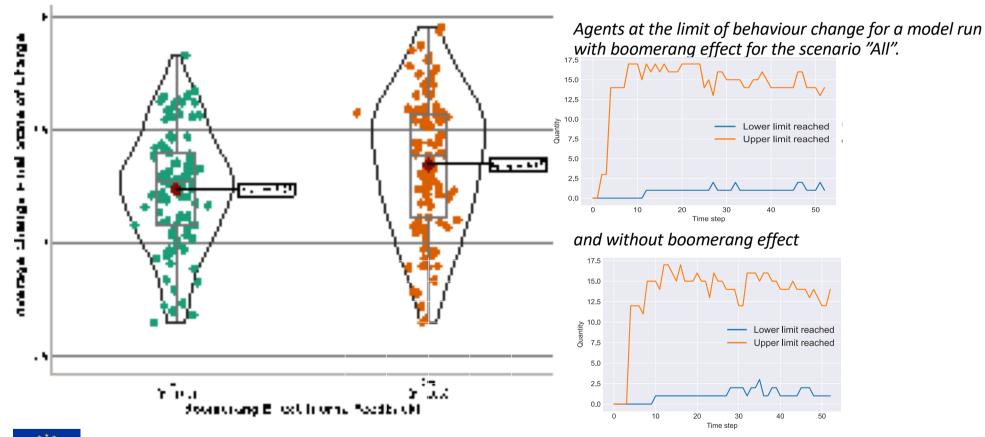
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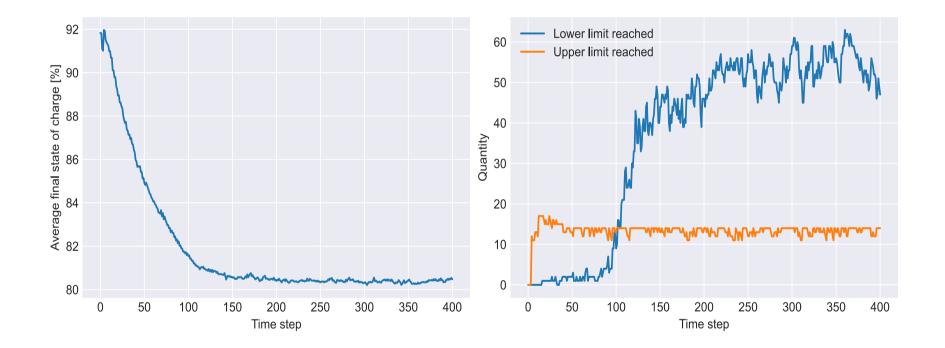


Impact of the Boomerang Effect (considering unfavorable behavior for normative feedback)





towards energy efficiency through behavioural science Long-term development of average SoC of all agents (left) & number of agents at the limit of behaviour change (right)





towards energy efficiency through behavioural science

Attribut	Abkür- zung	Beschreibung	Тур	Ausprägung
Agent ID	ID	Identifikator	int	-
Batterie- größe	BS	Kapazität des Elektroautos	float	3 Kategorien (Groß/Mittel/Klein)
Endlade- stand	SoC ^f	Ladestand der Batterie nach Lade- vorgang	float	prozentual
Mobilitäts- bedarf	MN	Mobilitätsbedarf des Agenten	float	in km/d
Sensibilität	S	Empfindlichkeit des Agenten ggü. externen Einflüssen	float	zwischen 0 und 1
Verbrauch	С	Verbrauch des Elektroautos	float	gleichverteilt (16- 25kWh/100km)



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Verhaltensänderung:	Parameter	
-	BS _i	[kWh]
$SoC_{i,t+1}^{f} = (1 - s_{i}) * SoC_{i,t}^{f} + s_{i} * \frac{SI_{i,t} + NI_{i,t} + EI_{i,t}}{p_{t} + q_{t} + r_{t}}$	C _i	[kWh/100km]
$s_{i,t+1} = (1 - s_i) * s_{i,t} + s_i * p_t + q_t + r_t$	C _{ij}	[-]
Einflüsse:	$EI_{i,t}$	[%]
Σ^m a c^{f}	m	[-]
$SI = -\frac{\sum_{j=1}^{m} w_{ij} * SoC_{j,t}}{\sum_{j=1}^{m} w_{ij} * SoC_{j,t}}$	MN _i	[km/d]
$SI_{i,t} = \frac{\sum_{j=1}^{m} w_{ij} * SoC_{j,t}^{f}}{\sum_{i=1}^{m} w_{ii}}$	n	[-]
y-1 ·y	$NI_{i,t}$	[%]
$w_{ij} = c_{ij} * \left(1 - \frac{ SoC_{i,t}^f - SoC_{j,t}^f }{50}\right)$	p_t	[-]
n	q_t	[-]
$NI_{i,t} = \frac{1}{n} * \sum_{j=1}^{n} SoC_{j,t}^{f}$	r _t	[-]
$EI_{i,t} = (1+x_i) * SoC_{i,t}^f$	s _i	[-]
$E_{i,t} = (1 + x_i) * SOC_{i,t}$	SI _{i,t}	[%]
Nebenbedingung:	$SoC_{i,t}^{f}$	[%]
$MN_{i} * C_{i/i}$	W _{ij}	[-]
$SoC_{i}^{f} \ge \frac{MN_{i} * C_{i}}{BS_{i}} * 100 + 10$	x _i	[-]

eter		
	[kWh]	Kapazität des Elektroautos von Agent <i>i</i>
	[kWh/100km]	Verbrauch des Elektroautos von Agent i
	[-]	Anzahl identischer sozialer Kontakte von
		Agent <i>i</i> und <i>j</i>
	[%]	Exogener Einfluss auf Agent i zum
		Zeitpunkt <i>t</i>
	[-]	Größe soziales Netzwerk
	[km/d]	Täglicher Mobilitätsbedarf von Agent i
	[-]	Gesamtzahl Agenten
	[%]	Einfluss von normativem Feedback auf
		Agent <i>i</i> zum Zeitpunkt <i>t</i>
	[-]	Binärvariable, Vorhandsein sozialer
		Interaktion zum Zeitpunkt t
	[-]	Binärvariable, Vorhandsein normatives
		Feedback zum Zeitpunkt t
	[-]	Binärvariable, Vorhandsein exogener
		Einfluss zum Zeitpunkt t
	[-]	Sensibilität von Agent i
	[%]	Einfluss sozialer Interaktion auf Agent i
		zum Zeitpunkt <i>t</i>
	[%]	Endladestand von Agent i zum Zeitpunkt t
	[-]	Gewichtungsfaktor der Stärke der
		Beziehung zwischen den Agenten <i>i</i> und <i>j</i>
	[-]	Zufallszahl zwischen -0,1 und +0,1

