# 5'th Workshop on Discrete Choice Modelling Warsaw 2016

Location: University of Warsaw, The Old Library building, room 107

Date: 5 – 6 October 2016

## **Program overview**

<b>Wednesday</b> 10:00 – 17:45	Day 1
<b>10:15 – 11:30</b> 11:30 – 11:45	Session 1: Spatially explicit DCE Coffee break
<b>11:45 – 13:00</b> 13:00 – 14:30	Session 2: Technical and methodological issues I Lunch
14:30 – 17:00	Session 3: Random regret Keynote – Caspar G. Chorus
15:45 – 16:00 17:00 – 17:15	Coffee break Coffee break
<b>17:15 – 17:45</b> 18:30 – 20:00 20:30 – 22:00	Special session 1: Basic discrete choice models in R Refreshments Dinner
<b>Thursday</b> 09:00 – 17:30	Day 2
09:00 – 13:00	Session 4: SP elicitation theory and methodology Keynote – Fredrik Carlsson
10:15 – 10:30	Coffee break
12:05 – 12:20	Coffee break
13:10 – 14:30	Lunch
14:30 – 15:00	Special session 2: Good practice recommendations for DCE beginners
15:00 – 17:15	Session 5: Technical and methodological issues II
16:00 – 16:15	Coffee break

#### Additional information

The conference is sponsored by University of Warsaw Foundation and the Department of Economics.

#### Partners





## Agenda for the 5<sup>th</sup> of October 2016

**10:00 – 10:15** Welcome and introduction (15 min)

Mikołaj Czajkowski

#### Session 1: Spatially explicit DCE

<b>10:15 – 10:35</b> (10 + 10 min)	The overview of methods to account for spatially explicit preference heterogeneity	<b>Wiktor Budziński</b> , Danny Campbell, Mikołaj Czajkowski, Urška Demšar, Nick Hanley	
	We review the commonly applied and emerging approaches to model spatia From a traditional two-step approach, through one-step estimation, to geogra models.	ally explicit preference heterogeneity. aphically weighted MNL, LC and MXL	
<b>10:35 – 11:00</b> (15 + 10 min)	Using spatial latent class models to identify willingness to pay hot and cold spots	<b>Danny Campbell,</b> Wiktor Budziński, Mikolaj Czajkowski, Nick Hanley	
	Notwithstanding the ability to include individual characteristics in the latent of possibility that the unobserved factors that explain membership to latent class errors are spatially arranged, meaning that the assumption that the error terr violated. Not addressing this means the model is mis-specified in the syste membership function - in particular, the omission of variables that are spat therefore, lead to bias, poor prediction and missed opportunities for insight. In modelling framework, whereby spatial dependence in class membership is as	lass membership function, there is a ses may be spatially related. If so, the ns are independent of one another is ematic component of the latent class tially clustered. Overlooking this will, this paper, we develop a latent class ddressed.	
<b>11:00 – 11:20</b> (10 + 10 min)	Comparing methods to account for spatial heterogeneity in discrete choice experiments	<b>Julian Sagebiel</b> , Klaus Glenk, Robert Johnston, Jürgen Meyerhoff	
	As willingness to pay values for environmental goods often vary by local incorporated geo-statistical methods in the analysis of discrete choice exper- validity of these approaches and, in many cases, different methods lead to diffi- briefly outline two approaches to model spatial preference heterogeneity and to pay values on a map. The first approach relies on predicting willingness to are interacted with attributes. The second approach makes use of spatial inter distance weighting). Here, geocoded individual willingness to pay values are willingness to pay in unobserved regions. We then propose ideas to compare to of results.	ation, researchers have increasingly periments. It is challenging to test the ferent results. In this presentation, we I use the results to predict willingness pay based on spatial variables which erpolation techniques (kriging, inverse e used as observed points to predict the approaches and to test the validity	
<b>11:20 – 11:30</b> (5 + 5 min)	Spatial sampling strategies: should we be using them	Danny Campbell	
	The main disadvantage of a classical random sampling approach is that it ignores any spatial dependence. If spatial dependence exists, random sampling may lead to data redundancy. For example, many observations may be clustered in one area when perhaps one or two observations might suffice. Thus, in the presence of spatial dependence, random sampling is inefficient. The aim of spatial sampling methods is to get results of a higher quality at a lower cost. In this presentation, I discuss whether or not spatial sampling matters for stated choice experiments.		
<b>11:30 – 11:45</b> (15 min)	Coffee break		
Session 2: Techni	cal and methodological issues I		
<b>11:45 – 12:05</b> (10 + 10 min)	Dealing with endoneity in DCEs	David Hoyos	
	Although dealing with endogeneity in classical regression models is well esta how to deal with this problem in the framework of non-linear models like D received some attention. The aim of this presentation is to discuss this issue of the performance of the Multiple Indicator Solution (MIS) method to deal	ablished in the econometric literature, ICEs is a field that only recently has by presenting an exploratory analysis ling with endogeneity in DCEs. This	

method will be tested using three environmental valuation datasets conducted in recent years.

<b>12:05 – 12:25</b> (10 + 10 min)	Hybrid Choice Models and accounting for the endogeneity of indicator variables: a Monte Carlo investigation	<b>Wiktor Budziński</b> , Mikołaj Czajkowski	
	We dispel the common misconception that the hybrid choice models address Through a Monte Carlo analysis, we demonstrate that similarly to directly inte choice attributes, including them in the measurement or structural component o automatically account for possible correlations. We propose a few solutions, all yet to be confirmed.	endogeneity of indicator variables. eracting indicator variables with the f the hybrid choice models does not Ithough their practical usefulness is	
<b>12:25 – 12:35</b> (5 + 5 min)	The importance of variable order when constraining correlation patterns between random parameters	<b>Tobias Borger</b> , Joseph Cook	
	When correlation is allowed between parameters in the random parameters logit, and particularly when some of these pairwise correlations are constrained, the order in which variables enter the model appears to have an influence on the estimates. Is this issue of concern and if so, how can we best deal with it?		
<b>12:35 – 13:00</b> (15 + 10 min)	The impact on welfare analysis of not modelling scale heterogeneity: a Monte Carlo experiment.	<b>Marco Boeri</b> , Alberto Longo	
	This note investigates, by mean of Monte Carlo simulation, the bias caused by the presence of a difference scale parameter across groups and the presence of individual scale parameter not incorporated in the model. The Monte Carlo study is conducted generating 1000 samples assuming the presence of 3 groups with different scale parameters – specifically equal to 1, 0.5 and 2. Main Findings: Estimating MNL models on datasets with a DGP that includes scale heterogeneity (both continuous and discrete) does not have a strong impact on parameter estimations different from cost, however the presence of scale impacts strongly on the cost coefficient. This has an important effect on Welfare analysis. In fact the WTPs are biased much more than the corresponding parameters. This finding is in line with what has been assumed so far in the literature and highlights the importance of including scale heterogeneity in modeling people preferences.		
<b>13:00 – 14:30</b> (90 min)	Lunch		

#### Session 3: Random regret

<b>14:30 – 15:45</b> (90 + 15 min)	Keynote address	Caspar G. Chorus,
	New insights on random regret minimization models	Sander van Cranenburgh, Cristian Angelo Guevara
	This paper develops new methodological insights on Random Regret I showing that the classical RRM model is not scale-invariant, and that minimization behavior imposed by the classical RRM model depends crup parameters in combination with the distribution of attribute-values in the commakes three methodological contributions: (1) it clarifies how the estimate are related to one another; (2) it introduces the notion of "profundity of re- this concept; and (3) it proposes two new family members of random regree and the Pure-RRM model. These new methodological insights are illust have been used to compare linear-additive RUM and classical RRM model analyses reveal that the degree of regret minimizing behavior imposed be very limited. This insight explains the small differences in model fit that literature between the classical RRM model and the linear-additive RUM re-	Minimization (RRM) models. It starts by t – as a result – the degree of regret cially on the sizes of the estimated taste lata. Motivated by this insight, this paper d taste parameters and the decision rule gret", and presents a formal measure of t minimization models: the µRRM model, rated by re-analyzing 10 datasets which les in recently published papers. Our re- by the classical RRM model is generally thave previously been reported in the model. Furthermore, we find that on 4 out

15:45 - 16:00 (15 min)

16:00 - 16:20

(10 + 10 min)

#### **Coffee break**

RRM model.

Information and choice paradigms in the preferences for renewable energy

Marco Boeri, Alberto Longo

This study aims to explore the impact of using two choice paradigms in deriving preferences for a stated discrete choice experiment on renewable energy programmes: the Random Utility Maximization (RUM) and the Random Regret minimization (RRM). In general RRM described better respondents' choices. When considering both choice paradigms in a hybrid model Hybrid we included 3 classes RRM completely free, we get one class with scale = 0 (pure RR) and one with scale very high (RU) plus a third class with scale not significantly different from zero (but very low probability). Furthermore these models are not identifiable (numerical problem when scale = 0). So given Van Cranenburgh et al, (2015), we suggest to change the class with scale = 0 with a pure RR model and the class with very high scale with a RUM model. We do so and we get a significant model using LC without heterogeneity (RRM has higher probability and members of environmental org have higher prob of being RUM). When accounting for preference heterogeneity, hybrid models are not identifiable. Indeed, when considering preference heterogeneity the best model is to estimate P-RR, with truncated distribution to avoid to cross zero. To test the robustness of the survey instrument, we assessed whether additional information can affect variance of the utility function, profundity of regret or impact on results. When considering RUM we find that varying the level of information has no impact on either preferences or variance, while on regret having less information results in higher profundity of regret (PRR).

of 10 datasets the µRRM model improves model fit very substantially as compared to the RUM and the classical

<b>16:20 – 16:45</b> (15 + 10 min)	What do we gain from introducing different decision r in non-market valuation?	ules Romain Crastes
	This paper introduces the µRandom Regret Minimization (µRRM) approach has been recently developed by Cranenburgh et al. ( minimization behaviour imposed by a RRM model. In this paper w of regret is allowed to vary across respondents and across attrib model to linear-additive RUM and classical RRM.	to the field of non-market valuation. The $\mu$ RRM (2015). It allows to model the degree of regret we introduce the mixed $\mu$ RRM where the degree outes and we compare the performances of this
<b>16:45 – 17:00</b> (5 + 10 min)	The regret of not modelling regret in choice experime Monte Carlo investigation.	ents: a Marco Boeri, Alberto Longo
	The Random Regret Minimization (RRM) approach to discrete choice analysis has been developed in the context of modelling the demand for travel, and, since then, it has been used in other fields including the demand for outdoor recreation and health. It presents a tractable, regret-based alternative to the dominant choice-modelling paradigm based on Random Utility Maximization (RUM). The idea that regret is an important determinant of choice behaviour is acknowledged, theoretically and empirically, in many fields including marketing, microeconomics, psychology, the management sciences, transportation and health. However, no previous study has yet measured the bias that the presence of a RRM behaviour in a dataset can create to estimations and welfare analysis based only on the RUM assumption. This paper explores and measures, by means of Monte Carlo simulations, the bias caused by estimating a multinomial logit model assuming that the data conforms to the RUM choice behaviour only, whilst the data presents a mixture of the two choice paradigms, both the RUM and the RRM. This bias is investigated with a gradually higher presence of the RRM choice behaviour in the Data Generation Process (DGP). We simulated 13 different datasets generating 1,000 samples each. The DGP is based on estimates from data of a real study in health economics aimed at testing the trade-off that people are willing to make between life style choices, in terms of diet and physical activity, and the risk of dying from cardiovascular disease in the next 10 years. We find that MNL models based on the RUM paradigm on datasets with a DGP that includes choices generated by a RRM approach results in biased parameters estimation. As expected, this bias is intensified by the increased proportion of choices driven by the RRM is about 50%. A further finding is interesting and counterintuitive: the bias is not as strong on willingness to pay estimates as on parameter estimates. We finally conclude supporting the idea of developing methods that allow for	
<b>17:00 – 17:15</b> (15 min)	Coffee break	
<b>17:15 – 17:45</b> (30 min)	Basic discrete choice models in R	Petr Mariel
	<i>R</i> is a language and environment for statistical computing and gra terms of the Free Software Foundation's GNU General Public Lic on a wide variety of UNIX platforms and similar systems, Wind introduction to the discrete choice modelling using <i>R</i> . There are more we focus on the core ones: multinomial logit, latent class and mixed on simulation exercises will be used to show in a didactic way estimation analysis.	phics. It is available as Free Software under the ense in source code form. It compiles and runs dows and MacOS. This presentation is a short any different models applied in the literature, but d logit models. Detailed worked examples based the estimation procedures as well as the post-
18:30 – 20:00	Refreshments	PiwPaw, Foksal 16
20:30 – 22:00	Dinner	Enoteka, Byrack Newcore Mieste 12/15

Rynek Nowego Miasta 13/15

## Agenda for the 6<sup>th</sup> of October 2016

#### Session 4: SP elicitation theory and methodology

<b>9:00 – 10:15</b> (60 + 15 min)	Keynote address	Fredrik Carlsson,	
	Old and new aspects of respondent behavior in stated preference surveys	Mitesh Kataria, Elina Lampi	
	We discuss the various aspects of why we sometimes observe a difference in and the corresponding survey situation. In particular, we investigate the role and experimenter demand effects. We find that an explicit budget exercise of improved water, primarily as a shift to an opt-out. An explicit mentioning of sub environmental projects) has little effect on respondent behavior). Finally we of demand script, which we find reduce willingness to pay for some of the attribut	n behavior between a real situation of the budget, available substitutes decreases the willingness to pay to ostitute uses of the money (on other developed a so-called experimenter es of the experiment.	
<b>10:15 – 10:30</b> (15 min)	Coffee break		
<b>10:30 – 10:55</b> (15 + 10 min)	Are the Effects of Real Incentives in Stated Choice Experiments Context Dependent? A Comparison of Choice Behavior in Online and Field Environments	<b>Ulf Liebe</b> , Klaus Glenk	
	We compare the results of hypothetical and real choice experiments on ethical consumption – the organic and fair trade tea – carried out in an online survey and field setting (i.e. "research si supermarket). We use propensity score matching to make the online and field data comparable (cirr group, online vs. field as well as hypothetical vs. real). Our findings indicate that the social context i willingness to pay is higher in the field setting than in the online setting. Second, we find a hypothetic tends to be larger in the field setting. Third, men and women seem to react differently to real incentive contexts. This study contributes to the research on the hypothetical bias in stated choice data by she social context is relevant for the effects of real incentives on individual's choices.		
<b>10:55 – 11:20</b> (15 + 10 min)	Rewarding truthful-telling in stated preference studies	<b>Romain Crastes</b> , Pierre-Alexandre Mahieu, Jordan Louviere, Ewa Zawojska	
	Stated preference surveys rarely provide respondents with such conditions in which a responsist to answer truthfully. As a result, reliability of stated preference data is often questioner method, grounded in economic theory, to incentivize respondents to answer truthfully. Our lie detector coupled with a reward. We discuss theoretical predictions of the method, and text split sample choice experiment dealing with a tree planting program. We find that the lie det time spent to complete the valuation tasks and (ii) decreases the variance of the error term b model that accounts for possible endogeneity. Our results are encouraging but more resear the validity of this new approach.		
<b>11:20 – 12:05</b> (15 + 10 min)	Do social norms matter for environmental preferences?	<b>Katarzyna Zagórska</b> , Mikołaj Czajkowski, Nick Hanley, Jacob LaRiviere, Natalia Letki	
	Do social norms matter? We investigate by experimentally varying the information about the social norm communicated to respondents in two empirical studies dealing with household recycling and GMO foods.		
<b>12:05 – 12:20</b> (15 min)	Coffee break		
<b>12:20 – 12:40</b> (10 + 10 min)	Show me the money	<b>Søren Bøye Olsen</b> , Kennet Uggeldahl	
	We test whether illustrating the cost attribute with pictures of real money can help reduce the welfare estimates derived from hypothetical discrete choice experiments, arguing that this method might help mitigate hypothetical bias. In a between sample design, we vary the presentation of the cost attribute, finding that pictures of real money significantly reduce willingness to pay estimates. The effect cannot be attributed to the visual presentation alone, as estimates do not differ between the control treatment and a treatment with a generic illustration of money, but only appear when real money is used in the illustration of the cost. These results are in line with previous findings		

in the behavior economics literature, and could improve the design of stated preference surveys.

<b>12:40 – 12:55</b> (5 + 10 min)	Time preferences and DCE	Morten Raun Mørkbak	
	Time preferences are important determinants of health related behavior – since such behavior often implies ar intertemporal choice. An often used approach in eliciting time preferences is a choice experiment in the sense of a multiple price list, where respondents have to trade-off a smaller-sooner reward over a larger-later reward – the switching point is then used as the interval of the discount rate. Alternatively – one a more regular choice experiment can be used, but were we run into usual scaling issue when comparing discount rates across groups/segments. Finally, one can elicit time preferences using a more simple time-trade-off method, and then use this as e.g. class specific variable in a latent class model – segmenting individuals according to time preferences and specific preferences in a giving DCE.		
<b>12:55 – 13:10</b> (5 + 10 min)	A different approach to stated choice experiments: new developments in political science	Ulf Liebe	
	Stated choice experiments are increasingly used in political science research. Here, researchers focus on causal inference, use fully randomized designs and non-parametric models. I would like to discuss how this perspective might complement choice experiment research in environmental economics and vice versa.		
<b>13:10 – 14:30</b> (80 min)	Lunch		
<b>14:30 – 15:00</b> (30 min)	Good practice recommendations for DCE beginners – but not only	<b>Jürgen Meyerhoff</b> , Klaus Glenk, David Hoyos, Jette Jacobsen, Petr Mariel, Søren Olsen,	
	The good practice recommendations are meant to be a list of issues to consider when designing a choice experiment, subsequently analysing data and reporting results. They were (and will be) chosen because a) we found them to be relevant when doing choice modelling, b) students raised them, or c) we stumbeld over them as reviewers. As our background is within environmental economics, the issues raised are with inspiration from applications in this field. The main objective is to provide info on good practise to help PhD students and practitioners improving at the same time hopefully the quality of studies and published papers especially in interdisciplinary contexts.		
Session 5: Technica	I and methodological issues II		
<b>15:00 – 15:20</b> (10 + 10 min)	Preference matching effects - it's always good to have more choice options, isn't it?	<b>Jürgen Meyerhoff</b> , Katrin Rehdanz,	

Katrin Rehdanz, **Christine Bertram** 

Present studies investigating the effects of the number of alternatives presented on a choice set have mainly found that more alternatives seem to be beneficial. However, mostly only choice sets with two and three alternatives, including a status quo alternative, have been compared. Thus, we use five split samples (300 respondents each) to vary the number of alternatives from two to six alternatives on a choice set, keeping all other design dimensions equal. Overall, respondents were presented 12 sets, first eight from the experimental design, and afterwards four sets randomly drawn out of the first eight sets. In addition, individual decision making styles a captured through various items batteries concerning, for example, maximization, regret minimization, or impulsivity. One of the study objectives is to examine whether more is always good or only applies to the move from two to three alternatives.

<b>15:20 – 15:35</b> (5 + 10 min)	DCE, eyetracking and "gaze-contingency"	<b>Søren Bøye Olsen</b> , Kennet Uggeldahl
	A new eyetracking experiment that includes a CE study. We hope to contingency" stuff, where the an alternative/attribute in a choice sets will on looks at it. The aim is to be able to learn more about true non-attendance.	be able to incorporate some "gaze- aly be visible if the respondent actually
<b>15:35 – 15:50</b> (5 + 10 min)	Speedy Gonzales! Some thoughts on speeders and what to do with them	Danny Campbell
	There are a growing number of papers looking response time. But there are presentation, I will give my tuppence worth on these issues.	still some unresolved issues. In this
<b>15:50 – 16:00</b> (5 + 5 min)	The effects of different specifications of standard deviations in the MXL model	<b>Mikołaj Czajkowski</b> , Wiktor Budziński
	We present a Monte Carlo simulation results demonstrating the effects of di deviations in the MXL model. It turns out that operationalizing them as linea. exp etc. substantially impacts model results. The conclusions for the field ar	ifferent specifications for the standard r (and ignoring sign), absolute value, re discussed.
<b>16:00 – 16:15</b> (15 min)	Coffee break	
<b>16:15 – 16:25</b> (5 + 5 min)	Choice task blocking and design efficiency	<b>Mikołaj Czajkowski</b> , Wiktor Budziński
	We share some insights from the investigation of how blocking function work ways to improve design efficiency.	ks in NGENE. We suggest practical
<b>16:25 – 16:50</b> (15 + 10 min)	Functional forms considerations in Maximum Acceptable Risk calculations	<b>Marco Boeri</b> , Juan Marcos Gonzalez
	Maximum acceptable risk (MAR) is commonly used to incorporate patient benefit-risk profiles for medications. Often, relative preferences are elicited particular treatment adverse event. The researcher must define a specifical choice model, which implies making assumptions about respondents' rela- excluded from the experimental design, but fall within the range of the risks these assumptions on the calculation of MARs and evaluate how differen- choices could affect benefit-risk evaluations of treatments.	ts' preferences into the evaluation of d for more than two levels of risk for a ation for the inclusion of the risk in the ative preferences for levels that were considered. We explore the impact of ces in MARs induced by specification
<b>16:50 – 17:15</b> (15 + 10 min)	Handling resolvable uncertainty from incomplete choice set scenarios – choice probabilities versus discrete choices	<b>Morten Raun Mørkbak</b> , Line Bjørnskov Pedersen, Riccardo Scarpa
	Forecasting choice behavior for new health care, environmental or transportation programs and servic challenging because actual data is often unavailable. In order to derive estimates of the demand for such pro- and services researchers often must resort to data derived from hypothetical market scenarios. An incl popular way of doing this is by means of hypothetical (Discrete) Choice Experiments (DCE). Respo- participating in a hypothetical discrete choice experiment are likely to be provided with only a subset information deemed relevant or even necessary for conducting a real life choice. Manski (1990) argues the under best case hypothesis, intentions stated during DCE survey will not be good predictors of future be since scenario descriptors will always be at least in part "incomplete". Such unavoidable incompleteness wil least in part resolved in a real choice context, which gives rise to a component of uncertainty referred "resolvable" because once faced with a real choice scenario subjects will have some uncertainty ref Cognizant of this fact analysts are faced by an extrapolation problem in which assumptions are likely to be and hence matter. However, eliciting choice probabilities (ECP) instead of stated choices could pot overcome this issue, by allowing respondents to explicitly state uncertainty about their stated choice. It tu that this approach might afford the additional advantage of being less econometrically demanding. In the paper we compare the elicited subjective choice probabilities approach with the standard DCE approach of split sample design in a health care context. The very preliminary results show large differences with resp willingness-to-pay estimates, but remarkable similarities with respect to forecasting abilities, suggesting the of the far less econometrically demanding ECP approach, which would seem to be at least as good as the more demanding DCE approach. Furthermore, we extend the model of the ECP approach by disting between those with at least some resolvable uncertainty and	
<b>17:15 – 17:30</b> (15 min)	Closing session	Mikołaj Czajkowski

## Locations:

#### Venue

University of Warsaw, The Old Library building, room 107

https://www.google.pl/maps/place/52%C2%B014'25.7%22N+21%C2%B001'09.8%22E/@52.2405507,21.017 8916,653m/data=!3m1!1e3!4m5!3m4!1s0x0:0x0!8m2!3d52.24046!4d21.01939?hl=pl

#### Refreshments, Tuesday 2016-10-04, 18:00+

The Alchemist, Plac Pilsudskiego 3

https://www.google.pl/maps/place/The+Alchemist+Gastropub/@52.2428383,21.0097043,17z/data=!3m1!4b1! 4m5!3m4!1s0x471ecc6133806a09:0x1d2066f748fb4fb8!8m2!3d52.242835!4d21.011893?hl=en

http://www.thealchemist.pl/

#### Refreshments, Wednesday 2016-10-04, 18:30 - 20:00

PiwPaw, Foksal 16

https://www.google.pl/maps/dir/52.2403242,21.0190482/PiwPaw+Beer+Heaven,+Foksal,+Warsaw/@52.2368 818,21.0160223,16z/am=t/data=!3m1!4b1!4m9!4m8!1m0!1m5!1m1!1s0x471eccf66fc8fad5:0x2ebff706c3cef6e 9!2m2!1d21.0209666!2d52.2338022!3e2?hl=en

http://www.piwpaw.pl/

## Diner, Wednesday 2016-10-04, 20:30 – 22:00

Enoteka, rynek Nowego Miasta 13/15

https://www.google.pl/maps/dir/PiwPaw+Beer+Heaven,+Foksal,+Warszawa/Enoteka+Polska,+rynek+Nowego +Miasta+15,+Warsaw/@52.2431531,21.005503,15z/data=!3m1!4b1!4m14!4m13!1m5!1m1!1s0x471eccf66fc8 fad5:0x2ebff706c3cef6e9!2m2!1d21.0209666!2d52.2338022!1m5!1m1!1s0x471ecc652f9b057f:0xb06f510255 bcbc2c!2m2!1d21.0075489!2d52.2528889!3e2?hl=en

https://www.enotekapolska.pl/

## List of participants:

- 1. Fredrik Carlsson (keynote)
- 2. Caspar Chorus (keynote)
- 3. Anna Bartczak
- 4. Marco Boeri
- 5. Tobias Borger
- 6. Wiktor Budziński
- 7. Danny Campbell
- 8. Romain Crastes
- 9. Mikołaj Czajkowski
- 10. Marek Giergiczny
- 11. David Hoyos
- 12. Ulf Liebe
- 13. Petr Mariel
- 14. Jürgen Meyerhoff
- 15. Morten Mørkbak
- 16. Søren Olsen
- 17. Julian Sagebiel
- 18. Erlend Sandorf
- 19. Katarzyna Zagórska

### University of Warsaw:

- 1. Karolina Safarzyńska
- 2. Jerzy Śleszyński

### Notorious absent:

- 1. Sergio Colombo
- 2. Thijs Dekker
- 3. Klaus Glenk
- 4. Stephane Hess
- 5. Jette Jacobsen
- 6. Thomas Lundhede
- 7. Riccardo Scarpa
- 8. Mara Thiene