

Valuation of the Weight-Specific Adolescent Instrument for Economic Evaluation using online personal utility functions in an adult population

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Overview

- ▶ This presentation will present the methods and results from the valuation of the Weight-Specific Adolescent Instrument for Economic Evaluation (WAItE) using online personal utility functions (OPUF) in a representative sample of UK adults.
- ▶ Before we dive in...
 - ▶ What is the WAItE?
 - ▶ What is OPUF?

Background

What is the WAItE?

- ▶ The WAItE is the first weight-specific health related quality of life measure designed for use in adolescents which is appropriate for use in economic evaluation.
- ▶ It is composed of 7 attributes and 5 levels (*never, almost never, sometimes, often, always*) within each:
 - **Tired:** I ... get tired.
 - **Walking:** I ... struggle to keep up when walking around with others
 - **Sports:** I ... avoid doing sport
 - **Concentration:** I ... struggle to concentrate on my studies/work
 - **Embarrassment:** I ... feel embarrassed shopping for clothes
 - **Unhappiness:** I ... feel unhappy because I am unable to do the same things as others
 - **Treated differently:** People ... treat me differently when I go out

Background

What is OPUF?

- ▶ OPUF is a new type of online survey for valuing patient reported outcome measures using more efficient, compositional elicitation methods, which even allow estimating value sets on the individual level.
- ▶ Research has shown that the results are comparable with values estimated via discrete choice experiment.
- ▶ OPUF main structure:
 - Attribute weighting
 - Level rating
 - Anchoring task

Aims

- ▶ To undertake a population-based valuation survey with adults using the WAItE OPUF to determine their preferences.
- ▶ To elicit a health state utility value for the WAItE PITS state.
- ▶ To explore preference heterogeneity within our sample and how it varies among different subgroups.

Methods

Recruitment

- ▶ 300 adults were recruited to respond to a quality-of-life survey hosted online.
- ▶ Study participants were recruited based on specific quotas to form a representative sample based on UK census data.
- ▶ The survey was hosted on the Prolific platform which invited paid respondents to complete the WAItE OPUF survey.
- ▶ Participation in this survey was estimated to take approximately fifteen minutes to complete and participants received £2.50 as a payment upon completion.

Survey structure

- ▶ Consent and Prolific ID
- ▶ WAItE descriptive system
- ▶ Attribute selection: *determine most important attribute*
- ▶ Attribute swing weighting: *determine relative importance of other attributes*
- ▶ Level rating: *determine importance of levels within each attribute*
- ▶ Anchoring: *determine the utility value of the worst WAItE health state*
- ▶ Survey feedback and demographic questions

Utility value estimation

- ▶ Attribute ratings are normalised to sum to 1 to denote their relative importance.
- ▶ Attribute weighting is combined with level ratings to yield a coefficient matrix which defines the marginal disutilities associated with each attribute level combination for the WAItE. An example is shown below:

$$L_{ij} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0.14 & 0.26 & 0.21 & 0.15 & 0.16 & 0.12 & 0.19 \\ 0.57 & 0.55 & 0.63 & 0.54 & 0.38 & 0.26 & 0.66 \\ 0.83 & 0.82 & 0.85 & 0.86 & 0.64 & 0.38 & 0.91 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix} \quad (1)$$

$$w_j = [0.08 \quad 0.10 \quad 0.11 \quad 0.14 \quad 0.30 \quad 0.10 \quad 0.17] \quad (2)$$

$$L_{ij} \cdot w_j = \tilde{M}_{ij} \quad (3)$$

Utility value estimation

- Combining attribute weightings with level ratings yields coefficient matrix \tilde{M}_{ij}

$$\tilde{M}_{ij} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0.01 & 0.03 & 0.02 & 0.02 & 0.05 & 0.01 & 0.03 \\ 0.05 & 0.05 & 0.07 & 0.07 & 0.11 & 0.03 & 0.11 \\ 0.07 & 0.08 & 0.09 & 0.12 & 0.19 & 0.04 & 0.15 \\ 0.08 & 0.10 & 0.11 & 0.14 & 0.30 & 0.10 & 0.17 \end{bmatrix} \quad (4)$$

- Then anchoring the coefficient matrix using the PITs utility value ($P = 0.2$) yields the anchored coefficient matrix \tilde{V}_{ij}

$$\tilde{M}_{ij} \cdot (1 - P) \rightarrow P = 0.2 \quad (5)$$

$$\tilde{V}_{ij} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0.01 & 0.02 & 0.02 & 0.02 & 0.04 & 0.01 & 0.02 \\ 0.04 & 0.04 & 0.06 & 0.06 & 0.09 & 0.02 & 0.09 \\ 0.06 & 0.06 & 0.07 & 0.10 & 0.15 & 0.03 & 0.12 \\ 0.06 & 0.08 & 0.09 & 0.11 & 0.24 & 0.08 & 0.14 \end{bmatrix} \quad (6)$$

Preference heterogeneity

- ▶ Investigating the heterogeneity of preferences between individuals, required a measure of dis/similarity to quantify how far apart two PUFs are.
- ▶ A utility value set was estimated for each individual in our sample and euclidean distance (EUD) was used to assess dis/similarity between preferences (shown below).
- ▶ We then used permutational analysis of variance (PERMANOVA) to explore which factors were influencing preference heterogeneity in our sample.

$$d_{EUD}(i, j) = \sqrt{\sum(u_i(s_1) - u_j(s_1))^2 + \dots + (u_i(s_{78125}) - u_j(s_{78125}))^2} \quad (7)$$

where $s = \{1111111, 2111111, \dots, 5555555\}$

- Results

Participant Characteristics

Characteristic	N (%)	Characteristic	N (%)
Age		Ethnicity	
18-24	32 (10.9%)	White	251 (84%)
25-34	50 (17%)	Asian	23 (8%)
35-44	48 (16.3%)	Black	11 (4%)
45-54	49 (16.7%)	Mixed	10 (3%)
55-64	81 (27.6%)	Other	5 (2%)
65-90	34 (11.6%)		
Not Stated	6 (2.0%)	Weight Status	
		Normal	154 (51%)
Gender		Overweight	104 (35%)
Female	154 (51%)	Obese	30 (10%)
Male	144 (48%)	Underweight	8 (3%)
Non-binary	1 (0%)	Prefer not to say	4 (1%)
		Underweight	8 (3%)
Education		Prefer not to say	4 (1%)
Degree	147 (49%)		
A Level	64 (21%)	Occupation	
Higher Education	46 (15%)	Full-time	130 (43%)
Other	20 (7%)	Part-time	62 (21%)
GCSE A-C	18 (6%)	Not Paid	30 (10%)
GCSE D-G	5 (2%)	Other	31 (10%)
		Starting a New Job	3 (1%)
Waite		Mean (SD)	
Tiredness	3.4 (0.8)		
Walking	2.1 (1.1)		
Sport	3.3 (1.3)		
Concentration	2.7 (1.0)		
Embarrassment	2.2 (1.2)		
Unhappiness	2.3 (1.0)		
Treated differently	1.9 (0.9)		
Total	17.8 (4.8)		

Level ratings

Level rating ^α	Mean (SD)	Median (Q1; Q3)	Min	Max
Tired				
Almost never	20.3 (23.2)	10 (5; 25)	0	100
Sometimes	36.3 (19.2)	33.5 (20; 50)	0	100
Often	62.2 (23.9)	70 (50; 80)	0	100
Walking				
Almost never	19.4 (21.8)	10 (6; 21)	0	100
Sometimes	37.7 (19.4)	40 (24; 50)	0	100
Often	63 (26.2)	71 (50; 80)	0	100
Sports				
Almost never	16.6 (21)	10 (5; 20)	0	100
Sometimes	29.5 (22)	25 (10; 45)	0	100
Often	49.8 (29.6)	50.5 (24.5; 75)	0	100
Concentration				
Almost never	21.4 (22.1)	14 (7; 25)	0	100
Sometimes	41.6 (20.1)	40 (25.8; 53.2)	0	100
Often	64.5 (26.2)	73 (50; 80.2)	0	100
Embarrassment				
Almost never	16.6 (22.3)	10 (4; 20)	0	100
Sometimes	29.4 (21.6)	25 (10; 50)	0	100
Often	47.9 (30.4)	50 (20; 75)	0	100
Unhappiness				
Almost never	21.1 (22.2)	13 (6; 25)	0	100
Sometimes	41.4 (22.1)	41.5 (25; 56)	0	100
Often	63.6 (28.2)	75 (50; 85)	0	100
Treated differently				
Almost never	20.9 (24.4)	11 (5; 25)	0	100
Sometimes	35.5 (22.8)	34.5 (19.8; 50)	0	100
Often	55.9 (30.6)	60.5 (31; 80)	0	100

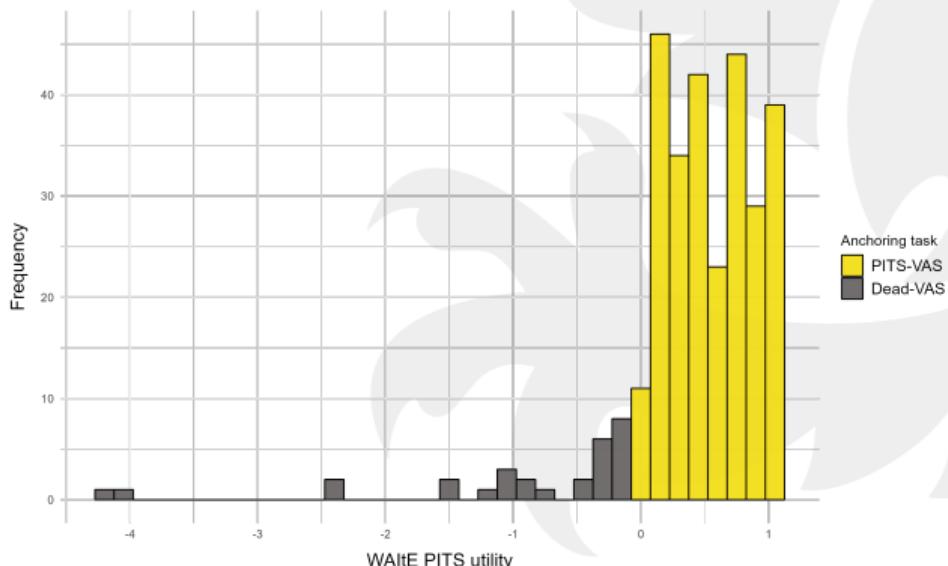
^α Levels Never and Always were fixed at 0 and 100 respectively

Attribute weighting and anchoring

OPUF Section	Mean (SD)	Median (Q1; Q3)	Min	Max
Domain weighting				
Tired	76.5 (28.4)	90 (60; 100)	1	100
Walking	65.5 (32.5)	75 (40; 100)	0	100
Sports	42.3 (32.8)	35 (11; 70)	0	100
Concentration	67.9 (30.9)	80 (44; 99.2)	0	100
Embarrassment	40.1 (34.3)	30 (9; 70)	0	100
Unhappiness	70 (31.9)	80 (50; 100)	0	100
Treated differently	52.1 (35.6)	50 (15.8; 86)	0	100
Anchoring				
PITS preferred to death	87.9% (32.7%)	1 (1; 1)	0	1
PITS-VAS	56.1 (31.3)	54 (30; 85)	0	100
Dead-VAS	42.5 (31.6)	38.5 (13.2; 63.5)	1	100
PITS VAS uncensored	-0.025 (5.95)	0.5 (0.2; 0.8)	-99	1
PITS VAS censored	0.431 (0.485)	0.5 (0.2; 0.8)	-1	1
PITS Utility Value (winsorized & imputed)	0.282 (1.456)	0.5 (0.2; 0.8)	-14.3	1

Anchoring distribution

- The distribution of PITS utility values were significantly left skewed (illustrated in the figure below). To mitigate this winsorization of values which lay in the outer 0.1% of the distribution was conducted. Missing values were imputed using multiple imputation by chained equations.



Utility value set

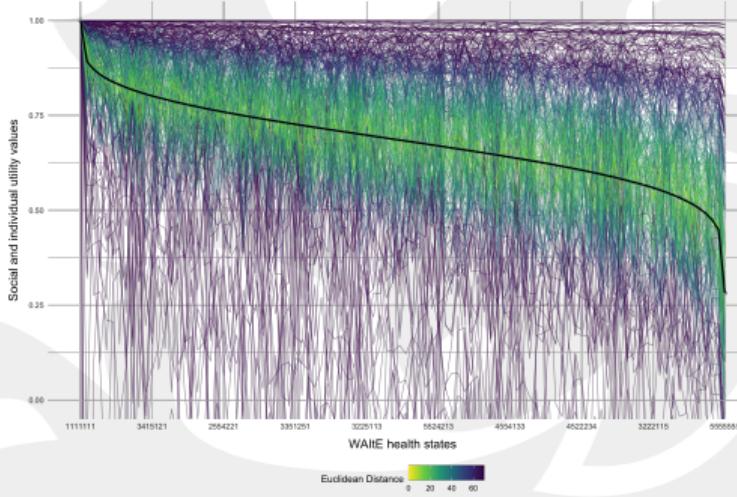
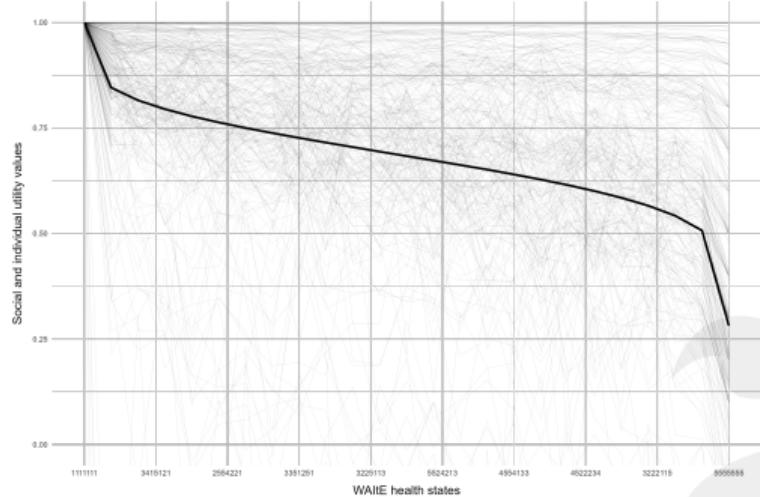
Dimension level	Mean ^a (95% CI) ^b	Median (Q1; Q3)	Min	Max
Tired				
Almost never	0.029 (0.025; 0.033)	0.029 (0.027; 0.030)	0.021	0.039
Sometimes	0.052 (0.047; 0.058)	0.052 (0.050; 0.054)	0.041	0.063
Often	0.088 (0.082; 0.094)	0.088 (0.085; 0.090)	0.077	0.103
Always	0.140 (0.133; 0.148)	0.140 (0.137; 0.143)	0.125	0.157
Walking				
Almost never	0.021 (0.018; 0.024)	0.021 (0.020; 0.022)	0.016	0.027
Sometimes	0.045 (0.041; 0.049)	0.045 (0.043; 0.046)	0.037	0.052
Often	0.075 (0.069; 0.082)	0.075 (0.073; 0.077)	0.064	0.087
Always	0.116 (0.108; 0.124)	0.115 (0.113; 0.118)	0.101	0.131
Sports				
Almost never	0.012 (0.010; 0.015)	0.012 (0.012; 0.013)	0.009	0.017
Sometimes	0.023 (0.020; 0.025)	0.023 (0.022; 0.024)	0.018	0.029
Often	0.038 (0.034; 0.044)	0.038 (0.037; 0.040)	0.029	0.052
Always	0.069 (0.063; 0.076)	0.069 (0.067; 0.071)	0.058	0.084
Concentration				
Almost never	0.026 (0.023; 0.030)	0.026 (0.025; 0.028)	0.019	0.034
Sometimes	0.051 (0.047; 0.055)	0.051 (0.049; 0.052)	0.044	0.060
Often	0.080 (0.074; 0.086)	0.080 (0.078; 0.082)	0.069	0.093
Always	0.121 (0.114; 0.128)	0.121 (0.118; 0.123)	0.107	0.138
Embarrassment				
Almost never	0.012 (0.010; 0.014)	0.012 (0.011; 0.013)	0.008	0.017
Sometimes	0.022 (0.019; 0.025)	0.022 (0.021; 0.023)	0.016	0.027
Often	0.034 (0.030; 0.038)	0.034 (0.032; 0.035)	0.025	0.043
Always	0.061 (0.056; 0.067)	0.061 (0.059; 0.063)	0.051	0.072
Unhappiness				
Almost never	0.025 (0.022; 0.029)	0.025 (0.024; 0.026)	0.019	0.033
Sometimes	0.054 (0.049; 0.059)	0.054 (0.052; 0.056)	0.045	0.064
Often	0.083 (0.076; 0.090)	0.083 (0.081; 0.086)	0.070	0.101
Always	0.124 (0.117; 0.133)	0.124 (0.122; 0.127)	0.110	0.142
Treated differently				
Almost never	0.019 (0.016; 0.022)	0.019 (0.017; 0.020)	0.013	0.025
Sometimes	0.035 (0.030; 0.039)	0.035 (0.033; 0.036)	0.026	0.043
Often	0.052 (0.048; 0.057)	0.052 (0.050; 0.054)	0.042	0.062
Always	0.087 (0.079; 0.095)	0.087 (0.084; 0.089)	0.071	0.101

^aCoefficients were anchored using a PITS utility value of 0.282.

^bConfidence intervals were estimated from bootstrap resampling with 10,000 iterations.

- Results

Preference heterogeneity



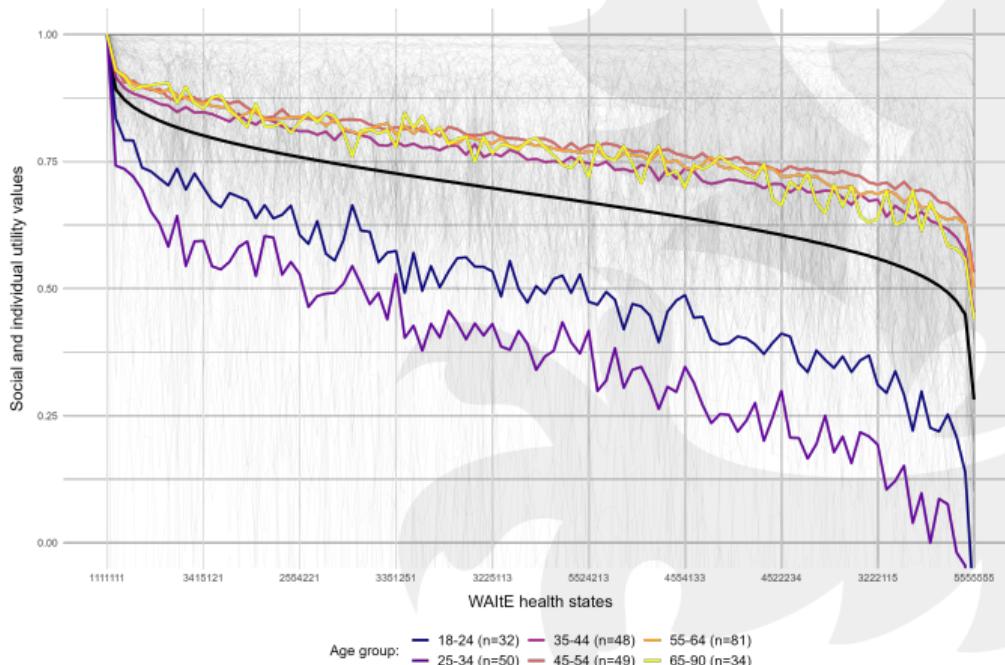
PERMANOVA

Group variable	Df	SS_W	R ²	F	p
Age	6	663590.648	0.057	3.018	0.030*
Gender	3	57313.577	0.005	0.521	0.362
Weight status	1	6892.412	0.001	0.188	0.727
Education	5	33542.464	0.003	0.183	0.967
Employment status	7	290563.598	0.025	1.133	0.269
Ethnicity	4	521334.829	0.045	3.557	0.056
Residual	273	10003165.515	0.864		
Total (SS_T)	299	11576403.042	1.000		

Abbreviations: df, degrees of freedom; F, pseudo F statistics; SS_T , total sum-of-squares;
 SS_W , within-group sum-of-squares.

p values based on 10,000 permutations; * = p < 0.05.

Preference heterogeneity

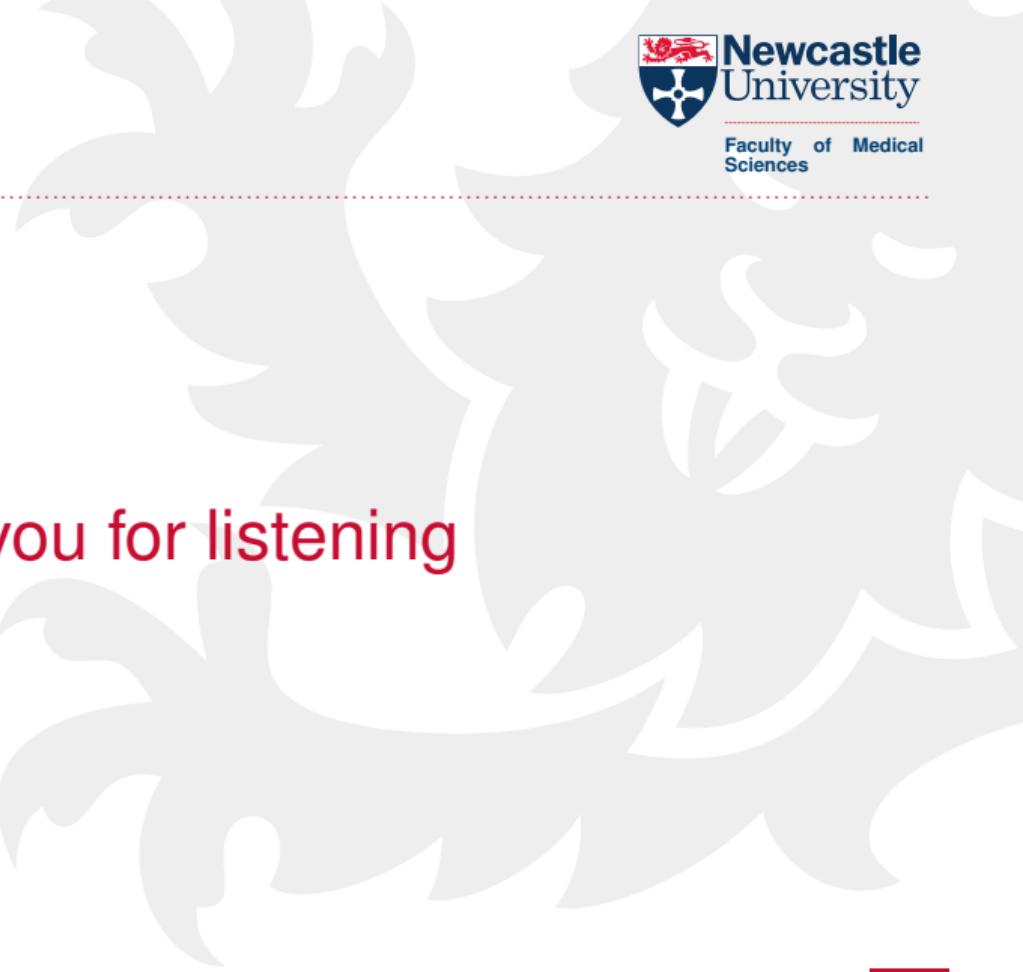


Discussion

- ▶ Winsorization of PITS utility values
- ▶ Tiredness and Unhappiness were considered the most important domains while Embarrassment and Sports the least.
 - ▶ This was consistent with prior valuation studies that have been completed with the WAItE.
- ▶ EUD and PERMANOVA were used to explore preference heterogeneity and age was shown to have a significant impact on preferences.
 - ▶ Younger participants generally had lower health state utility values for equivalent health states than older participants.

Conclusion

- ▶ Preferences varied significantly by age. This could indicate that adults and adolescents are likely to have systematically different preference values for the same health states.
- ▶ Another study in an adolescent population is ongoing to elicit an adolescent valueset for the WAItE using the OPUF.



Thank you for listening