


# Comparing the UK EQ-5D-3L and English EQ-5D-5L Value Sets

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## Abstract

**Background** Three EQ-5D value sets (EQ-5D-3L, crosswalk, and EQ-5D-5L) are now available for cost-utility analysis in the UK and/or England. The value sets' characteristics differ, and it is important to assess the implications of these differences.

**Objective** The aim of this paper is to compare the three value sets.

**Methods** We carried out analysis comparing the predicted values from each value set, and investigated how differences in health on the descriptive system is reflected in the utility score by assessing the value of adjacent states. We also assessed differences in values using data from patients who completed both EQ-5D-3L and EQ-5D-5L.

**Results** The distribution of the value sets systematically differed. EQ-5D-5L values were higher than EQ-5D-3L/crosswalk values. The overall range and difference between adjacent states was smaller. In the patient data, the

EQ-5D-5L produced higher values across all conditions and there was some evidence that the value sets rank different health conditions in a similar severity order.

**Conclusions** There are important differences between the value sets. Due to the smaller range of EQ-5D-5L values, the possible change in quality-adjusted life years (QALYs) might be reduced, but they will apply to both control and intervention groups, and will depend on whether the gain is in quality of life, survival, or both. The increased sensitivity of EQ-5D-5L may also favour QALY gains even if the changes in utility are smaller. Further work should assess the impact of the different value sets on cost effectiveness by repeating the analysis on clinical trial data.

## Key Points for Decision Makers

There are differences between the UK EQ-5D-3L and English EQ-5D-5L value sets.

The choice of value set will have implications for the decision-making process carried out by the National Institute of Health and Care Excellence.

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## 1 Introduction

In the economic evaluation of health interventions, the quality-adjusted life year (QALY) is a commonly used metric that combines length and quality of life into a single figure. The quality, or utility, weight used in the estimation of QALYs is anchored on a full health (1) to dead (0) scale, with negative values assigned to health states considered worse than dead. Utility values for health states associated with a particular condition or disease can be derived in

several ways, one of which is via the use of preference-based measures (PBMs) of health. Of currently available PBMs, the EQ-5D [1, 2] is the most widely used.

EQ-5D classifies health on five dimensions: mobility, self-care, usual activities, pain/discomfort, and anxiety/depression. The original version of the EQ-5D (described as EQ-5D-3L) included three severity levels (none, some, extreme/unable to)<sup>1</sup>, thereby describing ( $3^5 =$ ) 243 health states. In the UK, utility values for EQ-5D-3L health states were derived using the time trade-off (TTO) preference elicitation technique [3]. The resulting ‘value set’ has been widely influential, and is preferred by the National Institute for Health and Care Excellence (NICE) for use in the cost-utility analysis of health interventions [4]. EQ-5D-3L values are also accepted by reimbursement agencies worldwide, including the Pharmaceutical Benefits Advisory Committee (PBAC) in Australia [5] and the Canadian Agency for Drugs and Technology in Health (CADTH) in Canada [6]. The instrument itself is also used in a wide range of settings, including population health surveys and routine clinical practice [7].

Notwithstanding the widespread use of the EQ-5D-3L descriptive system and value set, research has suggested that both have a number of limitations. Regarding the descriptive system, it has been shown that the EQ-5D-3L is not sensitive to the important quality of life impacts of all conditions [8, 9]. It may also not be sensitive to smaller changes in health as it only has three response levels in each dimension, and in general public and some patient samples, a substantial proportion of respondents report themselves as being in the best health state, i.e. no problems on any dimension (11111). This is known as a ceiling effect [10]. Regarding the value set, the procedure and modelling used to elicit values for worse than dead health states has been criticised [11]. Furthermore, the EQ-5D-3L valuation data were collected in 1993, and population preferences for different aspects of health and quality of life may have changed in this time given advances in treatment and care. Social and environmental changes may also be important.

In an effort to improve the instrument’s sensitivity and reduce the ceiling effect, a five-level descriptive system, the EQ-5D-5L [12], was developed. The new instrument includes five response levels (none, slight, moderate, severe, extreme/unable to). The wording was also standardised across dimensions so that the worst level of mobility was changed from ‘confined to bed’ to ‘unable to walk about’, which is in line with the severity indicators used for the other functioning dimensions (self-care and usual activities). The intermediate severity level was also standardised

to be ‘moderate’. The EQ-5D-5L increases the number of states described to ( $5^5 =$ ) 3125. Research has shown improved measurement properties of the EQ-5D-5L descriptive system across a number of patient samples when compared to the EQ-5D-3L [13].

One consequence of this initiative was the need to develop value sets for the new descriptive system that reflect more up-to-date preferences of the population for health and quality of life, and this resulted in two separate developments. Firstly, an interim ‘crosswalk’ value set was developed so that EQ-5D-3L values could be used to predict EQ-5D-5L values [14]. Secondly, in order to elicit values for health states generated by the EQ-5D-5L descriptive system, a new valuation protocol combining TTO and discrete choice experiment (DCE) methods was developed [15]. This protocol used a ‘composite’ TTO approach combining standard and ‘lead time’ TTO [15–17]. In England, health states generated by the EQ-5D-5L were valued during 2012 and 2013 using this protocol and subsequently modelled using newly developed techniques that combined TTO and DCE data in a hybrid model to produce an EQ-5D-5L value set [18, 19].

Three EQ-5D value sets are therefore now available for use in cost-utility analysis in the UK and/or England, those being the EQ-5D-3L value set, the crosswalk value set mapping the EQ-5D-5L descriptive system onto the EQ-5D-3L value set, and the EQ-5D-5L value set. The first two of these were developed based on valuations from respondents in the UK, while the latter was based on valuations from respondents in England only. However, this is only one way in which they differ. As noted, they are also based on different descriptive systems, valuation protocols, and modelling methods. Given widespread and increasing use of the EQ-5D-5L in decision making, it is important to systematically assess the differences between the value sets, and the implications of the new values. For example, in recent work, it has been found that quality of life changes are valued less using the EQ-5D-5L value set [20]. At the end of 2017, NICE released a position statement regarding the use of the EQ-5D-5L stating that “the mapping function developed by van Hout et al. [14] [i.e. the crosswalk value set] should be used for reference-case analyses” until its position is reviewed in 2018 [21]. This means that the UK crosswalk is currently important in health technology assessment (HTA) carried out by NICE, and the results of studies comparing the new EQ-5D-5L value set with the crosswalk and EQ-5D-3L will inform future decisions about which measure to use. Therefore the aim of this paper is to add to the literature in this area by comparing the UK EQ-5D-3L and English EQ-5D-5L value sets, and the EQ-5D-5L and crosswalk value sets.

<sup>1</sup> In the EQ-5D-3L, level 3 mobility was described as ‘confined to bed’ not ‘unable to’.

## 2 Methods

### 2.1 The Value Sets

In the sections below, EQ-5D health states are described using five digits corresponding to each dimension and each level. The dimensions are listed in the order presented on the questionnaire (mobility, self-care, usual activities, pain/discomfort, anxiety/depression). For the EQ-5D-3L, 1 represents no problems, 2 some problems, and 3 extreme problems/confined to bed. Therefore EQ-5D-3L state 22222 describes some problems on each of the five dimensions. For the EQ-5D-5L, 1 represents no problems, 2 slight problems, 3 moderate problems, 4 severe problems, and 5 extreme problems/unable to. Therefore EQ-5D-5L state 22222 describes slight problems on each dimension.

#### 2.1.1 EQ-5D-3L

The UK EQ-5D-3L value set [3] was developed using data collected in 1993 from 2997 general population respondents who were sampled from the Postcode Address File. Respondents were recruited to be representative of the non-institutionalised adult population of England, Scotland, and Wales, and had similar characteristics to the UK General Health Survey sample [3]. Each respondent completed a face-to-face interview and valued 13 states (12 EQ-5D-3L profiles plus ‘unconscious’) using TTO which included one procedure for states valued better than dead and a different process for states valued worse than dead. In total, 42 of the 243 EQ-5D-3L states were valued, with an overrepresentation of the mildest health states. The data were modelled using additive generalised least squares (GLS) regression to produce a value set ranging from 1 (for the best state, 11111) to  $-0.594$  (for the worst state, 33333), with 34.6% of states valued as worse than dead. The model includes a constant subtracted for any move away from full health, a further decrement for each move away from ‘no problems’ for each dimension, and an additional term that is subtracted if any dimension is at the worst level (known as the ‘N3 term’). The value set also has a large change in utility between 11111 and the next best state (11211, which is scored at 0.883).

#### 2.1.2 Crosswalk

Crosswalk value sets were developed by van Hout et al. [14] from a multicountry study of respondents who completed both the EQ-5D-3L and EQ-5D-5L in 2010. The crosswalk used a non-parametric response mapping method to predict values that are linked to the EQ-5D-3L value set. The decrements for the ‘equivalent’ levels of the two

descriptive systems are the same. This means that the decrements for level 3 of the EQ-5D-5L (moderate problems) are the same as level 2 of the EQ-5D-3L (some problems), and those for level 5 of the EQ-5D-5L are the same as level 3 of the EQ-5D-3L. This means that the range of values is the same (55555 on the EQ-5D-5L has the same value as 33333 on the EQ-5D-3L, and an example intermediate state 35353 on the EQ-5D-5L has the same value as 23232 on the EQ-5D-3L). The crosswalk can link EQ-5D-5L data to a range of existing international EQ-5D-3L value sets. For the purposes of this paper, we focus on the crosswalk to the UK value set.

#### 2.1.3 EQ-5D-5L

The English EQ-5D-5L value set [18] was developed from 996 members of the general population who were purposively sampled from the Postcode Address File. In contrast to the EQ-5D-3L, respondents representative of the population of England (as opposed to the wider UK; a UK value set reflecting the preferences of respondents in England, Scotland, Wales, and Northern Ireland is due to follow) were recruited. The sample used differed slightly to the actual population as there were more older and retired people. Preferences were elicited using computer-assisted face-to-face valuation interviews that were conducted in 2012 and 2013. Respondents valued ten EQ-5D-5L states using composite TTO [15, 16] and completed seven DCE paired comparison tasks. In total, 86 states were valued in the TTO exercise and 196 pairs in the DCE tasks. The data was modelled using heterogeneous hybrid approaches combining the TTO and DCE data [19]. The resulting tariff ranges from 1 to  $-0.285$ , with 5.1% of the states valued as worse than dead. The model includes a decrement for each dimension for each move away from full health, and an extra ‘scalar’ coefficient. The range of values is therefore smaller than for the EQ-5D-3L, despite the considerable increase in the number of possible health states. The value of the mildest health states other than 11111 (12111 and 11211) is 0.950.

## 2.2 Analysis

We carried out an analysis comparing the predicted values from each of the three value sets, and also using patient reported EQ-5D-3L and EQ-5D-5L data. The patient data were taken from the crosswalk development study dataset where all respondents self-reported their health using both the EQ-5D-3L and EQ-5D-5L descriptive systems, thereby enabling direct comparisons. The key comparisons carried out were between the EQ-5D-3L and EQ-5D-5L value sets, and the EQ-5D-5L value set and the crosswalk tariff.

## 2.3 Comparison of Predicted Values

### 2.3.1 Comparing Value Set Models

Firstly we compared the coefficient models used to calculate the values. This was done to assess the overall magnitude of the coefficients for each dimension, and the impact of the various interaction coefficients included in each model on the values produced. We also compared the process for calculating values using an example health state.

### 2.3.2 Comparing Value Set Characteristics

We assessed a range of descriptive statistics of the possible theoretical values (i.e. 243 for the EQ-5D-3L and 3125 for the EQ-5D-5L). This included the value set range, the percentage of states valued as worse than dead, and the state with the smallest utility decrement from 11111. We looked at the modality of the overall distributions using kernel density histograms, and compared the values of selected states to demonstrate differences between the value sets.

### 2.3.3 Comparing Value Set Characteristics for Matched States

We carried out a comparative analysis on the states that are comparable across the EQ-5D-3L and EQ-5D-5L (i.e. the matched 243 states). The crosswalk value set is not relevant here, as for these states, the values are the same as the EQ-5D-3L tariff due to the response mapping procedure used. We considered comparable states to be those from the intermediate levels of the EQ-5D-5L descriptive system (i.e. none, moderate, and extreme/unable to) which, to a certain extent, ‘match’ the three-level states (as an example, the EQ-5D-3L state 12321 is defined as comparable to 13531 on the EQ-5D-5L). We assessed similarities and differences, both for individual states and at the overall level, to highlight where the largest differences occur across the value sets.

### 2.3.4 Comparing Differences in Utility Between Adjacent States

Analysis was also carried out to understand how changes in severity on the descriptive system are reflected by changes in utility. This was done by assessing the values of adjacent states within the descriptive system, and comparing the differences across the three value sets. An adjacent state pair was defined as having one dimension with a one-level difference (for example, calculating the change in utility between 21111 and 11111). This was done for states where

only one dimension changed at a time, so we focused on the change in utility between level 3/5 and level 1 on one dimension, with the other four dimensions held at the same level. For example, for mobility, we compared the increase in utility between 51111, 41111, 31111, 21111, and 11111, and we repeated this for all five dimensions. The magnitude of the change between all level changes, and the matched states, was assessed. This analysis reflects the coefficient decrements in a different way and provides an insight about how change in self-reported health would lead to change in utility in the absence of longitudinal data.

## 2.4 Analysis on Patient Data

### 2.4.1 Data Used

The data used to develop the crosswalk value sets were used for the analysis. The data were collected online across a range of patient groups with different health conditions who completed both the EQ-5D-3L and EQ-5D-5L descriptive systems. More information about the data collection procedure is provided in van Hout et al. [14]. Respondents from seven countries took part, but the analysis reported here used only the English and Scottish data. The characteristics of the 1501 respondents included are reported in Table 1.

### 2.4.2 Comparing the Descriptive System and Value Sets

Firstly, we compared the number of respondents reporting each level of the two descriptive systems. This was done to understand how the addition of the two extra levels changes response patterns. We compared the values using density plots, and also by assessing the scores overall and across patient groups (with the exception of those with a sample size of less than 50) using one-way ANOVA and mean difference statistics. We also compared the agreement between the value sets using Bland–Altman plots [22]. These present the mean of two scores on the  $x$ -axis and the difference on the  $y$ -axis, with lines indicating the upper and lower limits of agreement [calculated as the mean difference  $\pm 1.96 \times$  standard deviation (SD)] added. Agreement across the full severity range can then be assessed, with points outside the limits indicative of outliers.

**Table 1** Demographic characteristics of the crosswalk data used for the comparative analysis

Demographic	<i>N</i> (%)
<i>N</i>	1501
Country	
Scotland	500
England	1001
Age	
Mean (SD)	57 (16)
Range	19–94
Gender male	734 (49)
Education	
Left school with no qualifications	485 (32)
Left school with some qualifications	339 (23)
College degree/further education	377 (25)
Degree/postgraduate/professional	300 (20)
EQ-5D visual analogue scale	
Mean (SD)	60.3 (21.4)
Range	0–100
Condition	
COPD	320 (21)
Heart problems	251 (17)
Arthritis	250 (17)
Depression	250 (17)
Rheumatoid arthritis	87 (6)
Stroke	85 (6)
Back pain	70 (5)
ADHD	69 (5)
Diabetes	45 (3)
Parkinson's	37 (3)
Breathing problems	22 (2)
Multiple sclerosis	15 (1)

*ADHD* attention deficit hyperactivity disorder, *COPD* chronic obstructive pulmonary disease, *SD* standard deviation

## 3 Results

### 3.1 Comparison of Predicted Values

#### 3.1.1 Comparing Value Set Models

The models used to derive EQ-5D-3L and EQ-5D-5L values are displayed in Table 2. In each case, the coefficient decrements are larger for the more severe levels of each dimension and are therefore ordered as expected. Both models include a constant term, and in the EQ-5D-3L, this involves a decrement of 0.081 for the move away from the best health state (11111). The EQ-5D-5L constant is 1, and the coefficients are the mean coefficients from the modelling process after the application of the latent class

adjustments. The magnitude of the dimension level coefficients between the EQ-5D-3L and EQ-5D-5L varies (for example, pain/discomfort has a larger overall decrement on the EQ-5D-3L and anxiety/depression has a larger decrement on the EQ-5D-5L). The EQ-5D-3L N3 term is an extra decrement when at least one of the levels is at the most severe (i.e. level 3), and therefore this reduces the value of the more severe states. Table 2 also displays how to calculate a value for a state. The calculation is for the value for EQ-5D-5L state 21223 and the equivalent EQ-5D-5L state 31335, and this shows that the EQ-5D-3L value is substantially lower (0.186 vs 0.488).

#### 3.1.2 Comparing Value Set Characteristics

Table 3 (adapted from Devlin et al. [18]) compares the descriptive characteristics of the three value sets. The EQ-5D-5L value set has a higher value for the worst health state and substantially fewer worse than dead values. Also, the decrement from the best (11111) to next best health state (11211) is smaller for the EQ-5D-5L value set. This is expected given differences in labelling (e.g. 11211 describes 'slight' problems performing usual activities in the five-level instrument and 'some' problems in the three-level version). In all three value sets, pain/discomfort has the largest overall decrement (but not at the less severe levels), while self-care and usual activities have the smallest.

Figure 1 compares all unique theoretical values for the three value sets. The results demonstrate that the range for the EQ-5D-3L and crosswalk is different from the EQ-5D-5L. The large coefficients for level 3 on the EQ-5D-3L (and the impact of the N3 term) means that there is a higher density of lower values. The EQ-5D-5L is unimodal, whereas the EQ-5D-3L has multiple clusters, as has previously been observed [23].

#### 3.1.3 Comparing Value Set Characteristics for Matched States

Figure 2 displays the values of the comparable states from the EQ-5D-3L and the EQ-5D-5L value sets ordered by descending EQ-5D-5L value. EQ-5D-3L values are consistently lower across the full severity range. Figure 3 shows a histogram of the differences for each comparable state across the value sets, and a box plot of the mean difference by utility score category as a proxy for severity (1 to 0.500; 0.499 to 0.200; 0.199 to 0; <0). The mean difference is large overall at 0.312 (SD 0.102; range 0–0.484), and significantly increases as severity increases ( $F_{3,239} = 196.0$ ,  $p < 0.001$ ). Only 16 (6.6%) of 243 states have a mean difference smaller than 0.1, and 40 (16.5%) of the states have a difference of at least 0.4. The state with



**Table 2** Comparing the EQ-5D-3L and EQ-5D-5L models

Parameters	EQ-5D-3L	EQ-5D-5L <sup>a</sup>	Value calculation (21223/31335)	
			EQ-5D-3L	EQ-5D-5L
Constant	0.081			
<i>EQ-5D dimensions</i>				
Mobility				
None	0	0		
Slight		0.058		
Some/moderate	0.069	0.076	0.069	0.076
Severe		0.207		
CTB/unable to	0.314	0.274		
Self-care				
None	0	0	0	0
Slight		0.050		
Some/moderate	0.104	0.080		
Severe		0.164		
Unable to	0.214	0.203		
Usual activities				
None	0	0		
Slight		0.050		
Some/moderate	0.036	0.063	0.036	0.063
Severe		0.162		
Unable to	0.094	0.184		
Pain/discomfort				
None	0	0		
Slight		0.063		
Some/moderate	0.123	0.084	0.123	0.084
Severe		0.276		
Extreme	0.386	0.335		
Anxiety/depression				
None	0	0		
Slight		0.078		
Some/moderate	0.071	0.104		
Severe		0.285		
Extreme	0.236	0.289	0.236	0.289
Interactions				
N3 term	0.269		0.269	
Value of state			$1 - 0.081 - 0.069 - 0 - 0.036 - 0.123 - 0.236 - 0.269 = 0.186$	$1 - (0.076 + 0 + 0.063 + 0.084 + 0.289) = 0.488$

CTB confined to bed

<sup>a</sup>Mean coefficient from the Bayesian regression with the latent class adjustment applied

the largest difference is 32131 (53151 on EQ-5D-5L) (0.484), and the state with the smallest difference (excluding the best state) is 11212 (11313 on EQ-5D-5L) (0.023).

### 3.1.4 Comparing Differences in Utility Between Adjacent States

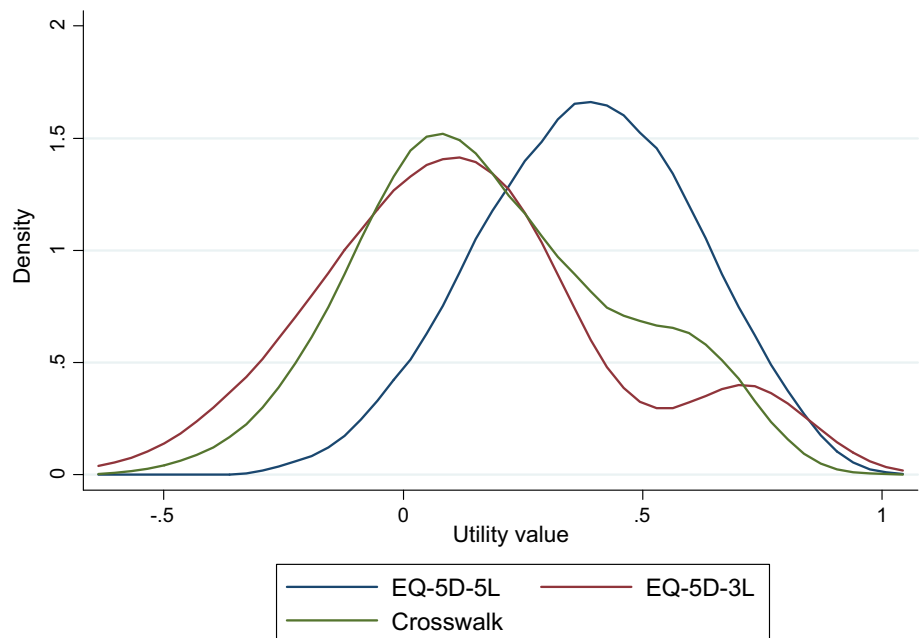
Table 4 displays the change in utility between adjacent and matched states. Comparisons of the matched states demonstrate that the change in adjacent states is substantially larger for the three-level tariff across all five

**Table 3** Overall descriptive characteristics of the three value sets (modelled values)

	EQ-5D-3L value set	EQ-5D-5L crosswalk	EQ-5D-5L value set
Range	1 to -0.594	1 to -0.594	1 to -0.285
% health states worse than dead	34.6%	26.7%	5.1%
Dimension importance order <sup>a</sup>	Pain/discomfort Mobility Anxiety/depression Self-care Usual activities	Pain/discomfort Mobility Anxiety/depression Self-care Usual activities	Pain/discomfort Anxiety/depression Mobility Self-care Usual activities
Health state values			
‘Mildest’ state (11211)	0.883	0.906	0.950 (11211/12111)
‘Moderate’ state [22222 (3L) or 33333 (5L)]	0.516	0.516	0.593
‘Worst’ state [33333 (3L) or 55555 (5L)]	-0.594	-0.594	-0.285

<sup>a</sup>Importance is judged by the size of the coefficient for level 5 in each dimension

**Fig. 1** All unique theoretical values



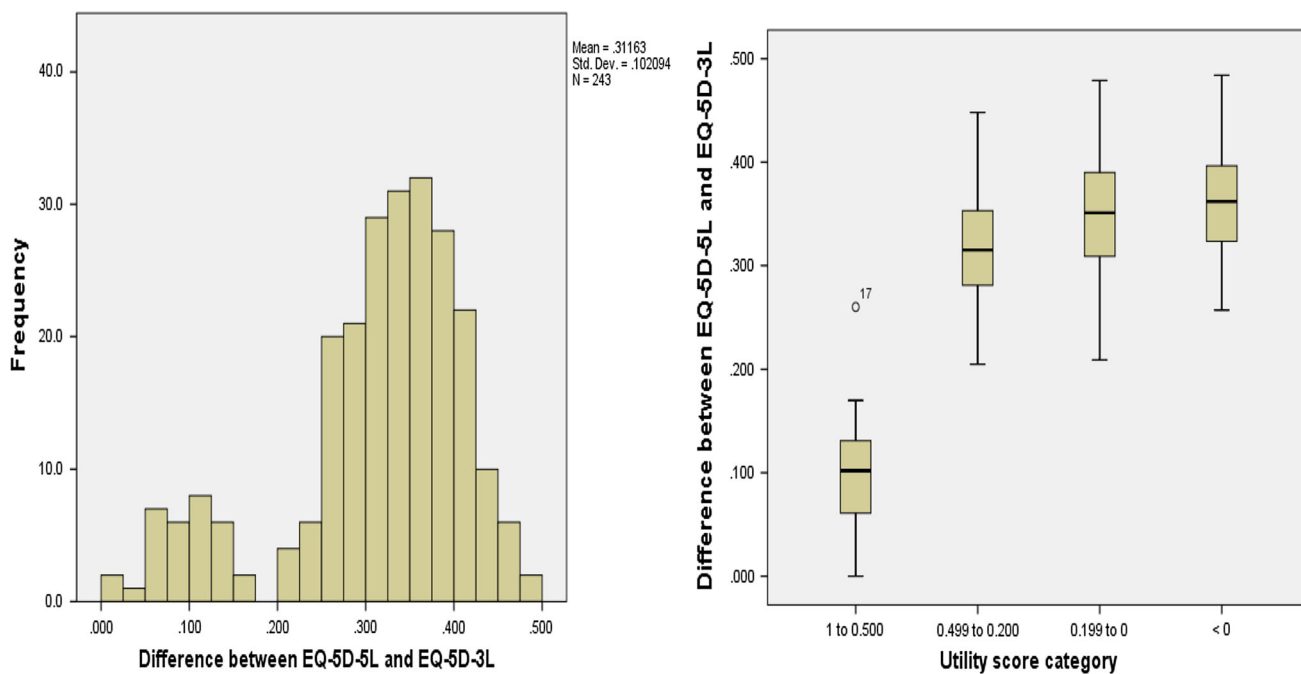
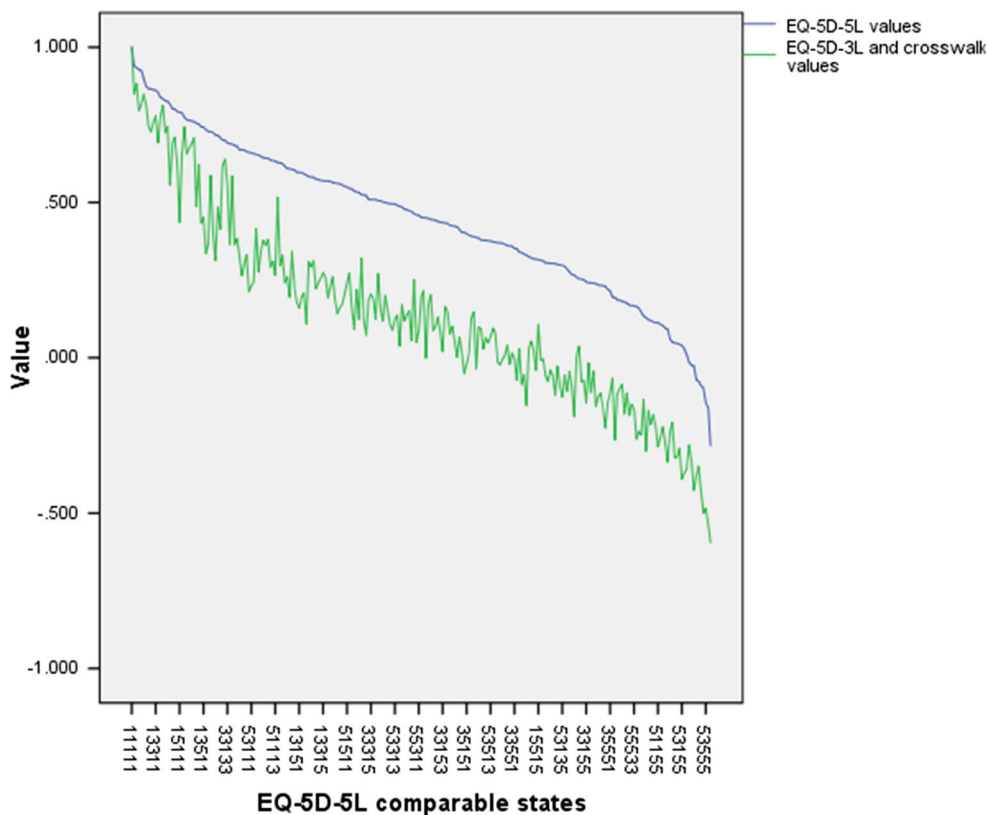
dimensions. This may suggest that the use of the EQ-5D-3L value set would tend to result in larger QALY gains for purely quality of life-improving interventions. Regarding EQ-5D-5L, the largest change in value occurs in the move from severe (level 4) to moderate (level 3) reported health problems. In contrast, the largest change in the crosswalk value set is between extreme/unable to (5) and severe (4) which is comparatively small in the EQ-5D-5L value set. The change in the crosswalk values from slight (2) to no problems (1) is larger than for EQ-5D-5L. This means that interventions resulting in an improvement in both mild and more severe health may result in larger QALY gains if the crosswalk values were used.

### 3.2 Comparisons Using Patient Data

#### 3.2.1 Comparing the Descriptive System and Value Sets

Table 5 displays the dimension level responses to the EQ-5D-3L and EQ-5D-5L and shows that the largest impact of the addition of the two intermediate levels (slight and severe) is to spread the ‘some’ responses on the EQ-5D-3L between levels 2–4 on the EQ-5D-5L. The introduction of ‘slight’ modestly reduces the ceiling effect as respondents move away from reporting no problems given the increased sensitivity. There is clear dispersion of scores from ‘some’ on the EQ-5D-3L across ‘slight’, ‘moderate’, and ‘severe’ on the EQ-5D-5L.

**Fig. 2** Values of comparable states ordered by EQ-5D-5L value



**Fig. 3** Histogram and box plot of differences between the EQ-5D-3L and EQ-5D-5L value sets

Figure 4 compares the EQ-5D-3L, EQ-5D-5L, and EQ-5D-5L crosswalk values. For the EQ-5D-3L, there is not only a large decrease in values in the very mild area (due to the upper gap reflected by the large constant), but also in

the moderate area around the values 0.25–0.45. In contrast, the EQ-5D-5L has a smoother distribution. This reflects a benefit of EQ-5D-5L: the increased sensitivity results in a much smoother transition between adjacent values that are



**Table 4** Comparing the change in utility between adjacent health states

EQ-5D-5L state	EQ-5D-5L value set			Crosswalk value set			EQ-5D-3L state	EQ-5D-3L value set	
	Value	Difference	Difference matched <sup>a</sup>	Value	Difference	Difference matched		Value	Difference
11111	1.000			1.000			11111	1.000	
21111	0.942	0.058		0.877	0.123				
31111	0.924	0.018	0.076	0.850	0.027	0.150	21111	0.850	0.150
41111	0.793	0.131		0.813	0.037				
51111	0.726	0.067	0.198	0.336	0.477	0.514	31111	0.336	0.514
11111	1.000			1.000			11111	1.000	
12111	0.950	0.050		0.846	0.154				
13111	0.920	0.030	0.080	0.815	0.031	0.185	12111	0.815	0.185
14111	0.836	0.084		0.723	0.092				
15111	0.797	0.039	0.123	0.436	0.287	0.379	13111	0.436	0.379
11111	1.000			1.000			11111	1.000	
11211	0.950	0.050		0.906	0.094				
11311	0.937	0.013	0.063	0.883	0.023	0.117	11211	0.883	0.117
11411	0.838	0.099		0.776	0.107				
11511	0.816	0.022	0.121	0.556	0.220	0.327	11311	0.556	0.327
11111	1.000			1.000			11111	1.000	
11121	0.937	0.063		0.837	0.163				
11131	0.916	0.021	0.084	0.796	0.041	0.204	11121	0.796	0.204
11141	0.724	0.192		0.584	0.212				
11151	0.665	0.059	0.251	0.264	0.320	0.532	11131	0.264	0.532
11111	1.000			1.000			11111	1.000	
11112	0.922	0.078		0.879	0.121				
11113	0.896	0.026	0.104	0.848	0.031	0.152	11112	0.848	0.152
11114	0.715	0.181		0.635	0.213				
11115	0.711	0.004	0.185	0.414	0.221	0.434	11113	0.414	0.434

<sup>a</sup>The 'difference matched' calculation refers to the difference between states that are matched across the EQ-5D-3L and EQ-5D-5L (e.g. 31111 on the EQ-5D-5L is equivalent to 21111 on the EQ-5D-3L)

closer together than on the EQ-5D-3L. The crosswalk value set distribution is more similar to the EQ-5D-5L, and the lack of EQ-5D-3L values in the range between approximately 0.25 and 0.45 is not apparent.

Figure 5 compares the EQ-5D-3L and crosswalk with the EQ-5D-5L and shows that there are differences in values across the entire severity scale, but greater variation for more severe health states (where the mean utility value is lower for the EQ-5D-3L and crosswalk). Figure 6 displays Bland–Altman plots comparing EQ-5D-3L and EQ-5D-5L, and EQ-5D-5L and crosswalk scores. There is evidence of disagreement between values across the severity scale, where the difference is outside the  $\pm 2$  SD range. Disagreement means more diverse utility scores for states of a similar severity.

The mean difference between the EQ-5D-3L and EQ-5D-5L values as reported by the patient sample is 0.073 (range  $-0.944$  to  $0.880$  calculated as EQ-5D-5L minus

EQ-5D-3L). Some respondents gave apparent inconsistent responses, and this results in the wide range overall. For example, the difference of  $-0.944$  results from a patient reporting 21111 on EQ-5D-3L and 44444 on EQ-5D-5L. Comparing the EQ-5D-5L and crosswalk value sets, the mean absolute difference is 0.085 and ranges from 0.002 for the states with the smallest non-zero difference (44431, 42433, 43441, and 41231) to 0.429 (for state 51131).

Table 6 compares the value set scores overall and across the different health conditions, with significance statistics reported for the conditions with more than 50 patients. As would be expected, the EQ-5D-5L values are higher, and the difference is significant for the four conditions with the largest sample size [chronic obstructive pulmonary disease (COPD), heart problems, arthritis, and depression]. Of the four conditions with a sample size of between 50 and 100, the difference tends towards significance for stroke and back pain, but not for attention deficit hyperactivity

**Table 5** Dimension level responses across the EQ-5D-3L/EQ-5D-5L (English and Scottish data)

Dimension responses	EQ-5D-3L, <i>n</i> (%)	EQ-5D-5L, <i>n</i> (%)
<b>Mobility</b>		
None	506 (33.7)	435 (29.0)
Slight		392 (26.1)
Some/moderate	983 (65.5)	377 (25.1)
Severe		277 (18.5)
CTB/unable to	12 (0.8)	20 (1.3)
<b>Self-care</b>		
None	951 (63.4)	907 (60.4)
Slight		301 (20.1)
Some/moderate	517 (34.4)	201 (13.4)
Severe		74 (4.9)
Unable to	33 (2.2)	18 (1.2)
<b>Usual activities</b>		
None	464 (30.9)	390 (26.0)
Slight		447 (29.8)
Some/moderate	881 (58.7)	358 (23.9)
Severe		228 (15.2)
Unable to	156 (10.4)	78 (5.2)
<b>Pain/discomfort</b>		
None	380 (25.3)	303 (20.2)
Slight		447 (29.8)
Some/moderate	947 (63.1)	449 (29.9)
Severe		243 (16.2)
Extreme	174 (11.6)	59 (3.9)
<b>Anxiety/depression</b>		
None	672 (44.8)	571 (38.0)
Slight		444 (29.6)
Some/moderate	721 (48.0)	324 (21.6)
Severe		111 (7.4)
Extreme	108 (7.2)	51 (3.4)

CTB confined to bed (level 3 of the EQ-5D-3L mobility dimension)

disorder (ADHD) or rheumatoid arthritis. The percentage of states worse than dead overall and also across each condition is lower for the EQ-5D-5L. Table 6 also displays the rank order of the severity of the conditions according to the mean utility values. There is evidence of consistency for seven of the 12 conditions, including the most (Parkinson's disease) and third most (back pain) severe conditions, and the five least severe (ADHD, breathing problems, arthritis, depression, and diabetes). The most variable condition is multiple sclerosis, which is second most severe according to the EQ-5D-3L, but fifth and sixth equal overall according to the crosswalk and EQ-5D-5L value sets, respectively.

## 4 Discussion

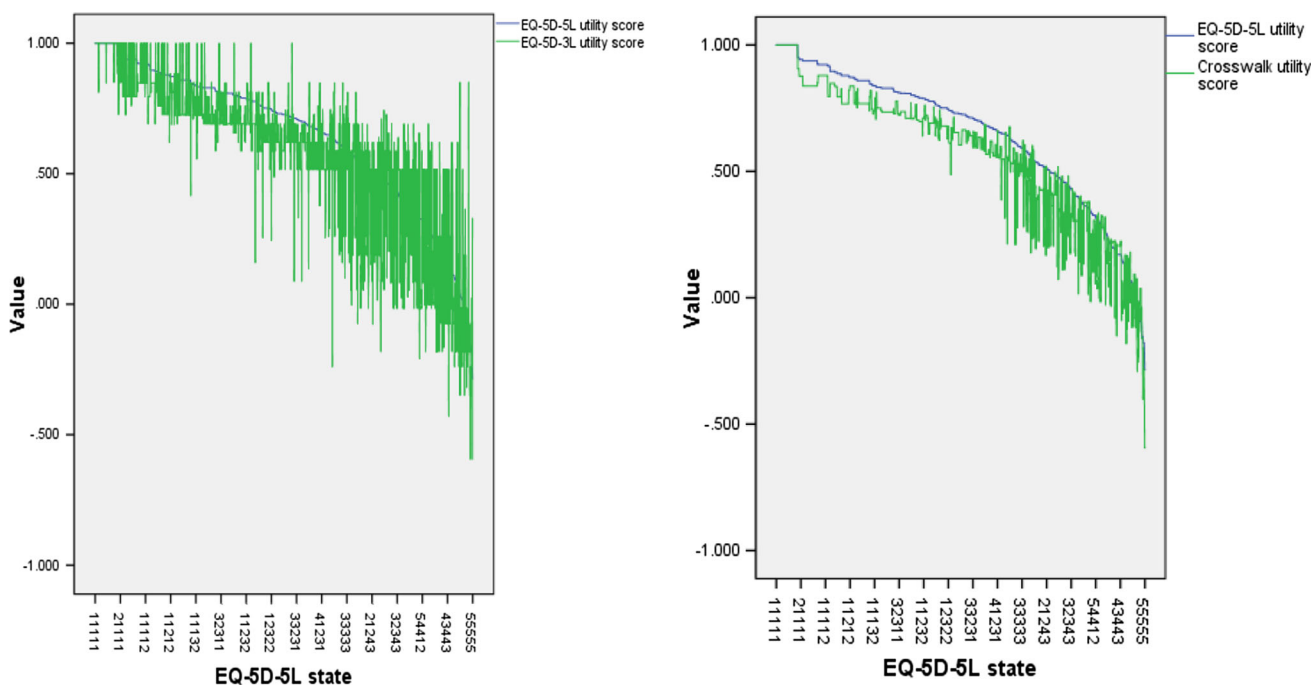
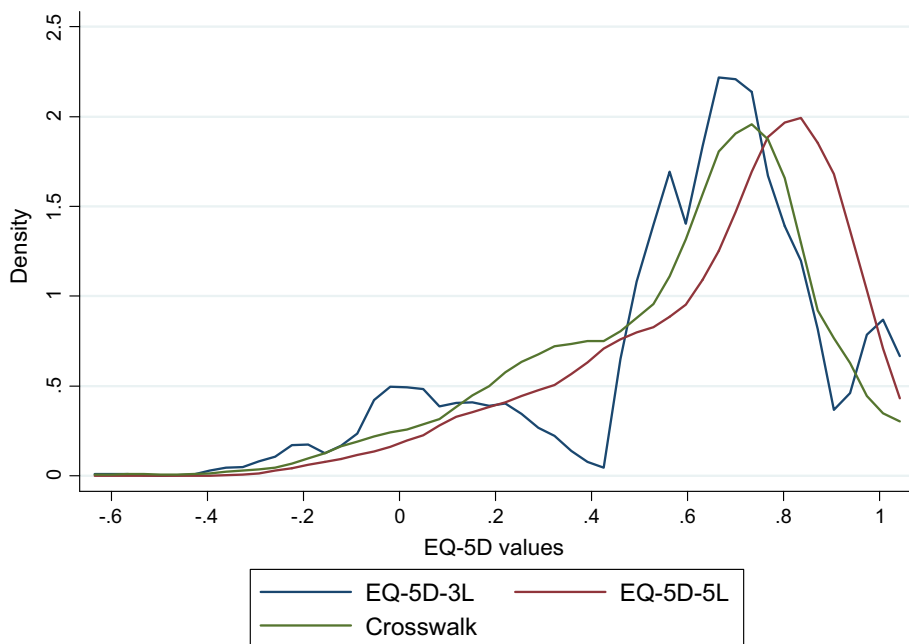
We have compared three EQ-5D value sets that can be used to support HTA in the UK. The comparison firstly investigated differences in the 'theoretical' values possible from the value sets for health states matched across the EQ-5D-3L and EQ-5D-5L descriptive systems and secondly compared values observed in patient data.

Regarding the theoretical values, the results demonstrate that there are differences between the EQ-5D-3L and EQ-5D-5L value sets, where the EQ-5D-5L values for matched states are higher, and the overall range and therefore change between adjacent states is smaller than for the EQ-5D-3L. The distribution of values also differs. There are similar differences between the EQ-5D-5L value set and the crosswalk tariff given that the latter is linked to the EQ-5D-3L value set. However, it is also worth noting that some underlying features of the preferences, and therefore utility scales, are similar. For example, the overall importance of each dimension to the overall value is similar, with only one difference (where the rank order of the dimensions is the same, apart from two dimensions, mobility and anxiety/depression, changing position in the ordering in the EQ-5D-5L value set), and the relative distance between the levels for different dimensions is similar.

Regarding the observed values from the patient data, the EQ-5D-5L value set produces higher values overall and across all of the conditions included, and the differences are generally significant. This is expected given the overall increase in the values of matched states and reduction in the overall utility scale. There is some evidence that the value sets rank different health conditions in a similar order, particularly the most and least severe conditions as measured by the descriptive systems. However, this requires further exploration across a larger range of conditions.

There are a number of possible reasons why the EQ-5D-3L and EQ-5D-5L value sets differ. These include differences in the samples used in terms of demographics and country. The EQ-5D-3L value set was based on a representative sample of England, Scotland, and Wales, whereas the EQ-5D-5L was based on just an English sample. This may have implications for decision making in the jurisdictions that are not represented. However, the project team has since collected EQ-5D-5L valuation data for the other countries in the UK so will be able to compare using a more representative sample (albeit one that is smaller than that used for the EQ-5D-3L). Potential changes in population demographics and preferences over time (from 1993 to 2013) are another possible reason why the value sets demonstrate differences. For example, the population is getting older [24], and this might impact on preferences for

**Fig. 4** Comparison of all EQ-5D-3L, EQ-5D-5L and EQ-5D-5L crosswalk values

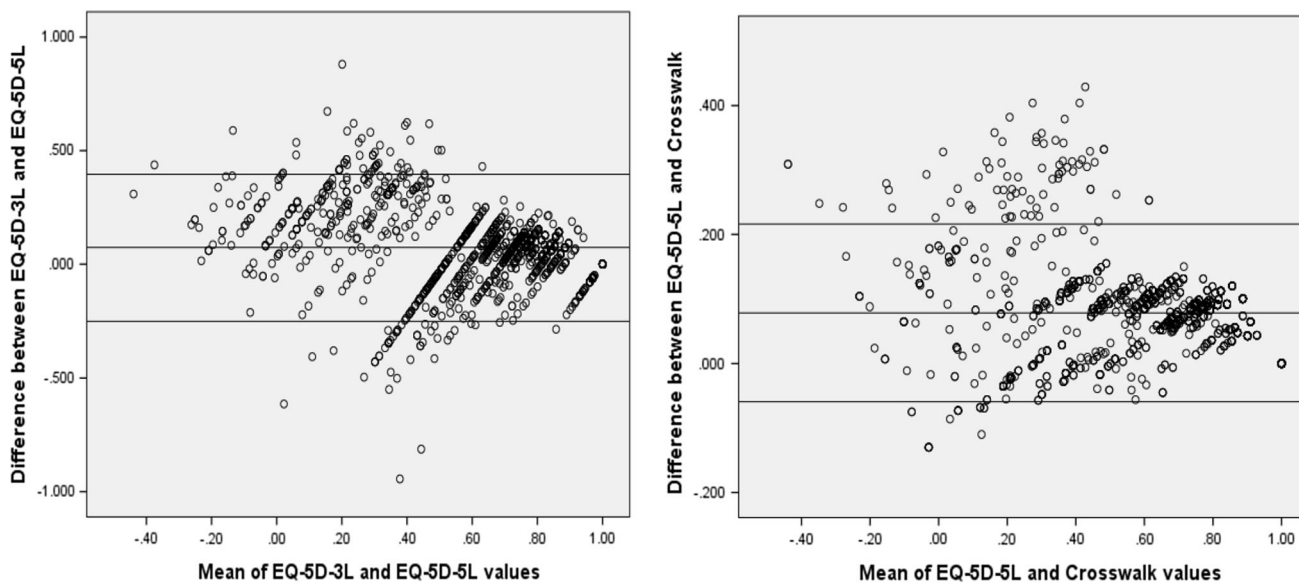


**Fig. 5** EQ-5D-3L and crosswalk patient values ordered by EQ-5D-5L

different health dimensions. One indication of change in preferences over time might be the increased magnitude of the anxiety/depression dimension given increased focus on the detrimental aspects of mental health conditions in policy [25] and reduction in stigma surrounding conditions such as depression [26]. Even without the development of the EQ-5D-5L, the currently used EQ-5D-3L value set is outdated and therefore would require updating anyway. Overall, the dimension preference structure between the

EQ-5D-3L and EQ-5D-5L is similar, with only one inversion (anxiety/depression and mobility), which is encouraging given the differences between the studies. This may demonstrate that the order of preferences for the five areas of health described by the EQ-5D may be generally consistent over time.

Other reasons why the value sets may differ relate to the descriptive system and the valuation method used. Firstly, regarding the descriptive system, the EQ-5D-5L uses more



**Fig. 6** Bland–Altman plots comparing EQ-5D-3L and EQ-5D-5L, and EQ-5D-5L and crosswalk scores

consistent wording, particularly for the more severe levels, and it is possible that the change in labelling of the mobility dimension (from ‘confined to bed’ to ‘unable to walk about’) has impacted the values, where mobility has a smaller weighting in the EQ-5D-5L than in the EQ-5D-3L. The increase in levels and associated sensitivity also may impact the magnitude of the difference and transition between the intermediate levels and therefore the overall value set.

Secondly, the valuation method differs, particularly regarding the process used to value states worse than dead, which was problematic for the EQ-5D-3L [11]. The methodological change to a new approach to eliciting values  $<0$ , the lead time TTO, meant that the lowest possible value for an EQ-5D-5L health state in the protocol used was  $-1$  [15, 27]. In contrast, the minimum value was  $-39$  in the Dolan study [3], which was rescaled to  $-1$ . This therefore led to a reduction in the overall scale. The inclusion of DCE tasks in the EQ-5D-5L valuation also provides a different type of valuation data focusing on the choices between states rather than measuring direct values for states, as is the case with TTO. The development of innovative modelling methods combining TTO and DCE data in one model [28, 29] provides further reasons for differences in the value sets. The modelling process for the EQ-5D-5L data also developed heterogeneous models for the TTO data only [19], and further work is underway to model the EQ-5D-3L valuation data applying the methods developed for the EQ-5D-5L [30]. It is also worth noting that a partial replication of the original EQ-5D-3L valuation study was carried out by Macran and Kind [31]. In this study, the authors used a smaller health state design, but a

similar TTO process to Dolan [3] and estimated an EQ-5D-3L value set with quite different characteristics. For example, the value for the worst state was substantially higher ( $-0.126$  vs  $-0.594$ ), and the amount of negative states was substantially lower (12.3 vs 34.6%). This value set is more in line with other EQ-5D-3L value sets developed internationally [32], and provides a useful counterpoint for comparisons between the value sets included in this study.

There are also large differences in the proportion of states valued as worse than dead (i.e. with a negative value) and the associated values assigned to these states, which has resulted in a smaller range for the EQ-5D-5L. One of the key criticisms of the EQ-5D-3L value set was the process used to value and subsequently model states worse than dead, which led to the large range observed [11], which may not realistically reflect population preferences. The protocol for the development of the EQ-5D-5L value set introduced a new method for the valuation of states worse than dead, which bounded all observed values on a  $-1$  to  $1$  scale [15, 17]. This has reduced the overall proportion of negative values and moved the anchor value of  $0$  (i.e. the state equivalent to dead). Further work could compare the characteristics of the health states that have values close to zero across different value sets.

However, the impact of the change in negative values on HTA is unclear, as it is not well established how often states that are worse than dead actually appear in cost-effectiveness models. There are differences in the proportions of negative states in different conditions, where the proportion is similar across the EQ-5D-3L and EQ-5D-5L in Parkinson’s disease, but quite different for multiple

**Table 6** Comparing value set scores overall and across different conditions

Demographic	N	EQ-5D-3L			Crosswalk			EQ-5D-5L			Significance		
		Mean (SD)	Range	% SWD	Mean (SD)	Range	% SWD	Mean (SD)	Range	% SWD	3L-5L	5L-Xwalk	
Overall	1501	0.577 (0.31)	1 to -0.594	8.6	0.571 (0.28)	1 to -0.594	4.5	0.650 (0.27)	1 to -0.285	2.3	<0.001	<0.001	
Condition													
COPD	320	0.546 (0.32)	1 to -0.349	11.3	0.558 (0.29)	1 to -0.292	1.9	0.629 (0.28)	1 to -0.179	1.9	0.001	0.003	
Heart problems	251	0.567 (0.32)	1 to -0.429	6.8	0.559 (0.29)	1 to -0.594	4.8	0.638 (0.27)	1 to -0.285	1.2	0.007	0.003	
Arthritis	250	0.636 (0.23)	1 to -0.181	5.6	0.618 (0.21)	1 to -0.134	0.8	0.704 (0.21)	1 to -0.069	0.4	0.001	<0.001	
Depression	250	0.643 (0.30)	1 to -0.349	6.8	0.640 (0.27)	1 to -0.160	3.6	0.718 (0.25)	1 to -0.153	2.0	0.002	0.001	
Rheumatoid arthritis	87	0.480 (0.34)	1 to -0.239	14.9	0.455 (0.31)	1 to -0.353	8.0	0.531 (0.31)	1 to -0.187	9.2	0.292	0.118	
Stroke	85	0.521 (0.32)	1 to -0.074	9.4	0.523 (0.29)	1 to -0.122	9.4	0.606 (0.27)	1 to -0.036	1.2	0.052	0.058	
Back pain	70	0.475 (0.28)	1 to -0.319	10.0	0.466 (0.28)	1 to -0.472	8.6	0.563 (0.26)	1 to -0.224	4.3	0.059	0.037	
ADHD	69	0.587 (0.33)	1 to -0.349	8.7	0.571 (0.28)	1 to -0.116	2.9	0.650 (0.27)	1 to -0.116	2.9	0.210	0.117	
Diabetes	45	0.723 (0.25)	1 to -0.016	2.2	0.707 (0.24)	1 to 0.093	0	0.773 (0.22)	1 to 0.020	0	n/a	n/a	
Parkinson's	37	0.431 (0.43)	1 to -0.594	18.9	0.410 (0.36)	1 to -0.594	13.5	0.487 (0.36)	1 to -0.285	16.2	n/a	n/a	
Breathing problems	22	0.616 (0.26)	1 to -0.181	4.5	0.616 (0.22)	1 to 0.206	0	0.688 (0.22)	1 to 0.171	0	n/a	n/a	
Multiple sclerosis	15	0.474 (0.37)	1 to -0.074	20.0	0.533 (0.31)	1 to -0.104	6.7	0.638 (0.26)	1 to 0.032	0	n/a	n/a	

ADHA attention deficit hyperactivity disorder, COPD chronic obstructive pulmonary disease, SD standard deviation, Sev order most severe mean value for each condition (where 1 is the most severe), % SWD percentage of states worse than dead, Xwalk crosswalk

sclerosis and COPD, for example. This might be due to changes in the magnitude of the decrement associated with the key dimensions for each condition. As the overall range of EQ-5D-5L values is smaller, the change in QALYs (for estimates generated from quality of life changes) might be reduced across the whole scale for states both better and worse than dead. This depends on the descriptive data, where respondents could show no change on the 3L (i.e. 'some' problems both before and after) whilst showing a change on the 5L (move from "moderate" to "slight"), leading to higher QALY gains.

It is also useful to compare the scale of the English EQ-5D-5L value set with those from other countries that were developed using the same valuation protocol [15]. For example, the Dutch value set has a minimum value of  $-0.446$ , with around 15% of states valued negatively [33]. The Spanish EQ-5D-5L value set has a minimum value of  $-0.224$  [34]. Differences between countries could be due to cultural differences in preferences as well as the use of different modelling approaches. Further work should compare EQ-5D-5L value sets from different countries in more detail.

It is unclear how the differences between the value sets indicated in both analysis of the estimates and patient data will impact the HTA process. This is because the utility values will be applied to both treatments and their comparators, and therefore to some extent the differences may be even, and the estimates of improvements in quality of life between arms of a clinical trial could be similar using the EQ-5D-3L or EQ-5D-5L value sets. The increased sensitivity of the EQ-5D-5L in terms of the addition of two extra response levels, and the change possible across the levels may also favour QALY gains even if the changes in utility are smaller. An added complexity is whether the gain is linked to improving quality of life or extending length of life, and the interaction between the two. This requires further investigation on clinical trial data, which is a key part of this programme of research, and has also been investigated by other researchers, who found different cost-effectiveness estimates based on the value set used [20].

There are also implications for the NICE reference case and further decision making based on their recently released position statement regarding the use of the EQ-5D-5L. The improvement in the methods used to both collect and model the valuation data and the increased use of the improved descriptive system make a strong case for the use of the new EQ-5D-5L value set. The EQ-5D-3L value set has benefits if the instrument is still being used in trials and other settings, but is based on societal preferences from decades ago. The crosswalk draws on the EQ-5D-3L values so is prone to the same issues as that value set. There is also the potential for 'gaming' where the crosswalk may be used instead of the EQ-5D-5L value set

to potentially inflate QALY gains (as the utility range, and therefore change between states, is larger). One important point is how to compare results of cost-utility analyses using the EQ-5D-5L against those using the EQ-5D-3L and establish the cost per QALY thresholds that should be used. Further work is required to explore this.

The main limitation of this study is that we have not tested the impact of the value sets on any clinical trial data, which would have enabled us to directly compare QALY estimations. This would allow us to test some of the issues raised in data previously used for cost-utility analysis, and is the next planned stage of this programme of research. It will also be important to compare the psychometric performance, and impact on cost-utility analysis, of the EQ-5D-5L descriptive system and value set with those of other widely used generic measures. In particular, comparisons with version two of the SF-6D (SF-6Dv2) [35], which has been valued using DCE with duration methods, would be useful.

## 5 Conclusions

In conclusion, we have demonstrated key differences in the theoretical and observed values from three EQ-5D value sets that can be used in HTA. The value sets will lead to differences, and the use of the EQ-5D-5L value set will have implications for the decision-making process carried out by NICE and may require revision to the guidelines used for the economic evaluation of health technologies.

**Data Availability Statement** The crosswalk data used in this study were funded and are owned by the EuroQol Research Foundation. The data are not publicly available as analysis is ongoing, but requests for data sharing can be made to the EuroQol Research Foundation.

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**Author contribution statement** BM led the data analysis and interpretation and was primarily responsible for drafting the manuscript. YF, KS, MH, and ND supported the data analysis and interpretation, and commented on and amended the draft manuscript. MFJ and BvH were involved in the collection of the data underpinning the study, the interpretation of the data analysis, and revising the manuscript. BM acts as overall guarantor of the work.

### Compliance with Ethical Standards

**Conflict of interest** All authors (BM, KS, YF, BvH, MFJ, MH, and ND) are members of the EuroQol Research Foundation (the copyright holders of the EQ-5D-3L and EQ-5D-5L). No funding was received for this study, but the crosswalk study data collection was funded by the EuroQol Research Foundation.



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