

Department of Health & Social Care





Using Discrete choice experiments to help services and commissioners to improve the experience of people who need care, understand different groups and secure better sustainability: a case example with application to homelessness.

**Presenters:** 

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# Human choices drive demand for products and services.

**Transport**: mode choice, route choice, destination, car ownership, departure time, etc.

**Health decisions**: treatment choices, insurance, lifestyle, etc.

#### Social care decisions: resource

allocation, service design and improvement, elderly care, long-term care planning, etc.



#### **"Understanding" current behaviour**

Why does somebody choose a specific product in a given setting? , Used extensively in cost-benefit analysis, etc.

#### Predicting future behaviour

What would somebody choose in a future setting? , Used extensively in demand forecasting with new or reconfigured services or a changed population. Why preference data are needed in social care homelessness?
 and
 Who needs such information?

- Lack of robust information on how people who are homeless value services and other support to inform better service provision.
- Such data are needed by **service planners, commissioners and service providers** to:
  - Inform the **design**, **delivery**, **implementation** and **sustainability** of services.
  - Increase engagement and reach people.
  - And make sure that, if preferences are met, people do not go back to rough sleeping.

- Choice modelling is a flexible tool (originated from market research).
  - A commonly used method in health services, but they are less frequently used in social care.
  - It enables measurement of the strength of preferences between alternative scenarios or types of service provision.
  - A discrete choice experiment survey is used to collect data.

#### **2** A

#### Analysing DCE data you can measure:

 How preferences are influenced by each attribute of a service.

**Respondents value** who provide care **more than** location of care.



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#### 2

#### Analysing DCE data you can measure:

 How preferences are influenced by each attribute of a service.

<u>Time they are willing to wait for Service</u> <u>A is 3 days (vs. service B 10 days)</u>



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#### 2 Analysing DCE data you can measure:

• How preferences are influenced by each attribute of a service.

<u>85% prefer Service A</u> (vs. 15% Service B)



## **Applying DCEs to**

social care

(homelessness)

#### **Evaluation of the Out-of-Hospital Care Models (OOHCMs)** programme for people experiencing homelessness in England (2021-23)

- Funded by the Department of Health and Social Care (£16M).
- Currently underway in 17 local authority test sites across England.
- The evaluation aims to explore how support for homeless patients can be integrated as part of the new discharge from hospital operating model.

# DCEs are applied as part of a larger programme of research:

- To explore which elements of OOHCMs people who are homeless most value.
- To model the probability of uptake for OOHCMs.

# Stage 1: Identify attributes important to people, their levels and choice format



## Stage 2 & 3: Choice Development and Testing



- Development done in co-production with stakeholders (commissioners, service providers and people who are homeless).
- Attributes and levels informed based on our previous research.
- Mathematical modelling was applied to create the set of questions for the DCE questionnaire.
- Feasibility was tested with people who are homeless.
- Piloting included the first 10 persons from one site. We tested data collection, cleaning processes and the model.



## **112 participants**

#### Sociodemographic features

Male Female

Average age: 48

People experienced better health91% at exit (compared to when they entered services)

#### Location and experience of care

- 65% Outside vs 35% within London
- 57% Pathway 1 vs 43% Pathway 2

#### Hospital (D2A) discharge models implemented in England

- **Pathway 0**: Patient returns to usual place of residence (sleeping on the street).
- Pathway 1: Take home and 'settle-in' support
- <mark>(Reablement)</mark>.
- Pathway 2: Specialist (bedded) intermediate care and step-down houses.
- **Pathway 3:** For people who require bed-based 24-hour care (care home).

#### Preferred model, most valued attributes and their relative importance



MESSAGE 1: Respondents do not want to go back to rough sleeping

They prefer any alternative arrangement to going back to rough sleeping (when carer type, number rules, frequency of visits and period of care remain the same)

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#### **MESSAGE 2:**

All service characteristics are valued\* and the MAIN CARER (housing support worker) ranked first



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Housing support worker as MAIN CARER (vs. all other options)

#### No RULES

(vs. rules about behaviour where you live)



FREQUENCY OF VISITS (per week)

(they value less visits per week)

Own flat/house as LOCATION (vs. all other options)

PERIOD OF CARE (weeks) (they value a longer period)

\* MNL regression modelling, significance at 1% level

#### **Probability of services uptake (changing one attribute at a time)**



- **Alternative Service** You are in **your own home.** A **nurse** delivers most of your care. care. • You are seen **1 to 2 times per** week. week. Your care is provided for 4-5 weeks.
  - There are **no rules** about your • behaviour.



#### **Returning to rough sleeping\***

- You are rough sleeping.
- A **nurse** delivers most of your
- You are seen **1 to 2 times per**
- Your care is provided for 4-5 weeks.
- There are **no rules** about your behaviour.



#### Probability of services uptake (changing one attribute at a time)



**Alternative Service Returning to rough sleeping\*** • You are **rough sleeping**. • You are rough sleeping. • A **nurse** delivers most of your • A **nurse** delivers most of your care. care. • You are seen **1 to 2 times per** • You are seen **1 to 2 times per** week. week. Your care is provided for 4-5 Your care is provided for 4-5 weeks. weeks. There are **some rules** about • There are **no rules** about your your behaviour. behaviour. 31% 69%

#### Probability of services uptake (changing multiple attributes at a time)



You are in your own home.

**Alternative Service** 

- A **housing support worker** delivers most of your care.
- You are seen 3 to 4 times per week.
- Your care is provided for 10-12 weeks.
- There are **no rules** about your behaviour.



**MESSAGE 3:** preferred model of care

**Returning to rough sleeping\*** 

- You are **rough sleeping**.
- A **nurse** delivers most of your care.
- You are seen **1 to 2 times per** week.
- Your care is provided for **4-5** weeks.
- There are **no rules** about your behaviour.



### Comments

- DCE survey was well received by study participants.
- Response rate 37% of the total cohort of 300 people enrolled from the 10 sites participating in the DCE survey.

#### **Analysis of DCE data indicates:**

- Respondents do not want to go back to rough sleeping.
- All service characteristics are valued and the three top characteristics are:
  1<sup>st</sup>: Main carer (housing support worker)
  2<sup>nd</sup>: No rules and
  3<sup>rd</sup>: Decreased frequency of visits.

#### The preferred service model would offer:

- Housing support worker comes to their home
- 3 to 4 times per week
- Care for 10 -12 weeks
- No rules about behaviour.



A key aim of the 'OOHCM Programme was to support individual test sites to develop a 'dashboard of key indicators' that could then be used in their business plans to make the strongest case possible for future funding. But also, to contribute to the broader landscape of national decision-making.

#### A management tool for national and local stakeholders to monitor progress and play a pivotal role in driving long-term service improvements.

#### **Key Metrics**

- Investment and budget utilisation
- People demographics
- Process outcomes (e.g., the flow of people in and out of care, staff composition, workload and more)
- Economic outcomes (for NHS and broader public budgets)
- Health outcomes (Quality-Adjusted Life Years, QALYs)
- Housing outcomes
- Care experiences
- Preferences for different models of care





## **Oxfordshire Out of Hospital Care Model**

- 2021 initial funding from DHSC's Shared Outcomes scheme (15 months)
- 17 test sites across the county innovative, partnership working
- Aims:
  - Planned, safe discharges from hospital avoid discharge to street
  - Increase access to services in community avoid (re)admissions and reduce inequalities
  - Prevent rough sleeping and homelessness
- Project evaluated by King's College London and London School of Economics
- 2022 further short-term funding secured; model expanded
- 2023 Two-year funding secured (BCF and ICB)



Preventative Step-up services

- Social Workers
- Clinical Psychologists / Psychiatrist
- Mental Health Practitioner
- Occupational Therapist
- Step Up accommodation

Avoid unnecessary

admissions

Acute General or Mental Health Hospital

- Housing Options Officers
- Dual Diagnosis service
- MH Health Support Workers

Person facing homelessness • Peer assessors

• LEAF

Step-down accommodation

- Up to 6 weeks free of charge
- Input from OOHC team and Primary Care

Oxfordshire Out of Hospital Care Team

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 Ongoing housing
 Transitional support from clinical roles and EMHWs

## **Key metrics**



#### **Step Down Hospital Data**

#### **Performance April 2021 – December 2022**

- Over **200 planned discharges** from hospital (50% from Mental Health wards)
- Average length of stay in our Step Down service is 28 days
- 22% of people in Step Down rough sleeping prior to hospital; only one person returned to rough sleeping
- Hospital data
  - 24% reduction in emergency admissions
  - 56% reduction in presentations to ED
  - 155% increase in Outpatient visits

#### **My Role and experience**

#### **Implementing OOHC in Oxfordshire**

- Recruitment and team development cohesion and values
- Connect with key stakeholders identify need/shape model
- Establish a place in the system trust and commitment from partners
- Collect data and build evidence base identify opportunities
- Secure ongoing funding most projects stopped after 12 months

#### **Collaborating with the evaluation team**

- Supportive, responsive, knowledgeable, easy to work with
- Appointed too late to contribute to the initial business case
- Clear and credible data that stood up to scrutiny
- LEAF / EBE valuable insights, authenticity
- Great support from NHS data team

## Use of the data and plans for the future

#### **Data and Evaluation**

- Adapted our Step Down service
  - Pre-move in paperwork and move in protocols
  - Consistent information and communication
  - Challenges: House Rules, length of stay (cp to DCE feedback)
- Recruitment LEAF on interview panel and part of MEAM

#### What next?

- Longer-term outcomes LEAF
- Storytelling qualitative, experience and preference (alongside economic/outcome data)
- Adopt dashboard as management tool
- Use DCE data and dashboard to design model
  - Best scenario of care
  - Predictive service uptake and number of beds required
- Embed in system
  - A way of working (cp to MEAM)
  - Ongoing funding business as usual







#### The team:

Michelle Cornes (co-PI) and Va Michela Tinelli (co-PI), Kyann Z Joanne Madridejos (**London S** Jess Harris, Joanne Coombes Stan Burridge (**Expert Focus**)

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