National Volunteer Management Conference



Challenges with indentifying the effects of volunteering

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Why is Volunteer Impact Important?

'Volunteer Impact' → the effects of volunteering

- Volunteering always has many effects even if we're only interested in a few
- There are different types of effect \rightarrow direct vs. indirect, volunteer vs. community
- Different effects are important for different people

The effects of volunteering are important for many reasons, including...

- Help us understand the value of volunteering & find ways to improve
- Provide evidence to recruit volunteers and/or attract funding
- Volunteering isn't an unqualified good: we need to **mitigate the downsides**

Anybody here for another reason?

Volunteer Impact Masterclass

How is impact typically measured?

Major challenges confronting research about volunteer impact

- · Defining & measuring impact
- Measuring volunteering
- Causality conundrum

Addressing the challenges: research designs

What can we do?

How are volunteer impacts typically measured?



Challenges: Defining & Measuring Impact

Major challenge to much research stems from failure to clearly **define** & sometimes even **measure** impact – or **volunteering**!

• Bekkers and Verkaik (2015) reviewed 400 articles mentioning 'effects of volunteering'; fewer than 10% contained both the measure of volunteering & it's theorised impact!

Must also be clear about the **nature** of the impact we're interested in – different types need to be measured in different ways:

- Direct: deliberate or intended consequence of volunteering
- Indirect: impacts besides the deliberate/targeted, potentially unintended

A pre-requisite to valid identification of volunteer impact is a clear definition & measure of it, and of the volunteering thought to cause it!

'Volunteering' is multi-faceted & complex: it is impossible to measure perfectly

We aren't aiming for perfection: we're aiming **to learn something important** & **sacrifice complexity** to achieve it

Key is to be **clear about those sacrifices** & their **implications for our research**

In essence, researchers take 3 approaches to measuring 'volunteering':

- 1) Instance specific: defined by participation in program, activity, event
- 2) Generalised: defined by researcher for use in broad reach data collection
- **3) Membership count**: defined by 'membership' of organisation, programme etc.

Most volunteer impact research uses **instance specific** from a sample of volunteers defined by participation in a specific activity, organisation, event

□ E.g., evaluation study, interview with participants in European Solidarity Corps

Evidence of impact from opinions, experiences, recollections of volunteers

□ Interviews, focus groups, diaries, online forms, volunteer-surveys

These approaches have numerous advantages:

- Clarity of definition: what is meant by 'volunteering' can be **clearly defined** and **accurately recorded** in a way that **captures diversity** of activities, roles, contexts etc.
- Efficiency: certain data is collected from relevant volunteers & they are easy to recruit
- Causal validity: able to prompt and/or draw clear links with impact of interest

The **drawbacks** include:

- Difficulty of comparison with **non-volunteers** or volunteers from other events, orgs etc.
- Data analysis limited to whatever was directly collected
- Numerous sources of bias which threatens accuracy and/or generalisability
 - □Selection: sample has something in common that might not apply to other people
 - Volunteers are (statistically speaking, at least) weird most people don't volunteer!
 - □ **Recall:** ability to remember what we did, how it affected etc. is flawed in varying ways
 - Age, how important volunteering was, mental health
 - □ Response or social desirability: responses shaped by desire to please
 - Someone who's just volunteered is unlikely to tell the organiser it was rubbish
 - □ **Attribution:** people focus on immediate or prompted potential cause of something
 - 'Do you think this volunteering programme has improved your self confidence?'

Most common alternative for academics are **generalised measures**

☐ Most common form: 'Have you volunteered for an organisation in last 12 months?'

Impact identified based on associations between characteristics in data

□ Correlation between volunteering & voting in elections

Main advantage: used in large surveys of nationally representative samples

- Allows generalisation from the survey data to larger populations
- Allows **comparisons** that can identify impacts e.g., with non-volunteers, volunteers for different types of organisations, volunteers who differ by age, gender, health
- Easier to study **indirect impacts** of volunteering or those **not experienced by volunteer**
- Less susceptible to **attribution** & **response** bias

Drawbacks:

- National surveys suffer from **social desirability** & **selection** bias
- 'Imposes homogeneity on diversity' (Mohan 2024): requires considerable simplification of volunteer experiences, motivations, activities, contexts etc.
- That simplification assumes common understanding of 'volunteering'...
 - ☐ Most people don't use the word 'volunteering' to describe their activities
 - ☐ Common understandings of 'volunteering' vary between cultures & over time
- ...and how we choose to do it can **substantially affect results**
 - ☐ Higher volunteering rates if you prompt specific activities

Most common for economists & civil servants is 'membership count'

□ E.g., How many members of Action Aid Ireland + Birdwatch Ireland etc.

Impact evidenced by **association between aggregated statistics**

□ Correlation volunteer membership & crime; volunteer hours as paid labour

Advantages include:

- It is (relatively) **easy & cheap** to access and analyse such data
- Enables study of volunteering & impacts using statistics valued by policymakers
- Can recognise **membership of different organisations**
- Easier to study **indirect impacts** of volunteering or those **not experienced by volunteer**
- Not at all affected by recall, attribution or social desirability bias

Drawbacks:

- Simplification: will not capture diversity of experiences within organisations
- Accuracy: not every organisation **defines 'members'** in the same way; not every member will be **active** & many people **join more than one** organisation
- Research limited to fields about which other aggregated statistics are available –
 which isn't true for many impacts we are interested in
- **Ecological Fallacy:** relationship at aggregate level doesn't necessarily indicate causal effect of volunteering at individual level

Challenges: Causality Conundrum

Final challenges relate to how we can **identify the causal effect of volunteering**, i.e., how do we know volunteering **caused** the impact we identify?



- Does volunteering make people happier, or are happier people more likely to volunteer – or both?
- Does having more volunteers reduce crime, or do people volunteer in neighbourhoods with low crime? Or is it both?

'Proving' causality is extremely difficult – **especially for volunteering** – and this issue is

ignored in most research

Challenges: Causality Conundrum

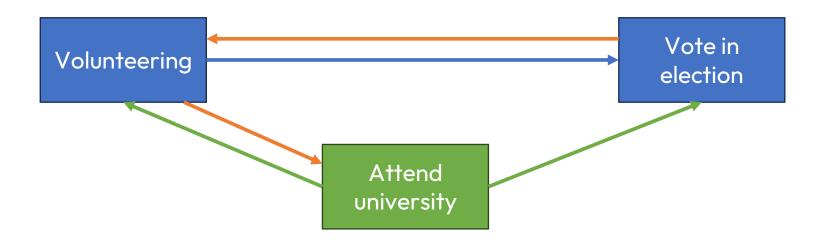
Many characteristics identified as **benefits** of volunteering also **encourage** it

- □Volunteering associated with improved **mental health**, **educational qualifications**, **income**, **political engagement** and **social capital**
- □ Superior mental health, educational qualifications, income, political engagement and social capital are all associated with increased likelihood of volunteering



Challenges: Causality Conundrum

Many of the drivers and impacts of volunteering are also related to each other



How can we be confident that **volunteering caused the impact**, rather than **both volunteering and the impact** being caused by **something else**?

Addressing the Challenges: Key Principles

Don't aim for perfection. The purpose of research is to tell us something important we wouldn't otherwise know

Be explicit about the **specific impact** you are investigating & make certain you **observe/measure it directly**

Be explicit about the **way you measure volunteering** and how **sacrifices involved** affect your research & findings

Exploit **advantages of different ways** of measuring volunteering & impact

Use **lessons from previous research**: how they measured volunteering, minimised selection bias, overcame causality conundrum etc...

Research design: strategy for collecting, organising & analysing data

Many challenges – causality conundrum in particular – can't be 'solved' but are being mitigated by researchers using different research designs

Panel Studies

- Collection of same data from same individuals over time
- Provides far greater confidence about volunteer effects by directly observing change in the characteristic of interest within the same people
 - □ In my research, I compared the interest in politics of the same young people before and after volunteering to identify its effect on their political engagement
- Can be used with national surveys, volunteer reporting, interviews almost anything!

Comparative designs

- Comparative research uses **comparison to attribute (potential) causation**
 - □ E.g., Compare mental health of volunteers with non-volunteers
- Commonly employed in quantitative & survey-based research but less so in interviews & very rarely used in volunteer evaluation studies
 - □ Could interview non-volunteers & compare with volunteers to identify impact
- Combined with **panel design** can be powerful way of identifying volunteer impact & accounting for influence of confounding factors
 - □ E.g., Compare mental health of volunteers & non-volunteers at time1, time2, time3 & see how trajectories differ

Data Matching

- This one is very rare but has serious & growing potential
- Involves matching of similar individuals across different sources of data
- Can **compensate** for our inability to measure all the characteristics we're interested in, to compare over time or between non/volunteers, or small sample size
- Cost of **specificity & precision** about the volunteers we study or attribute impacts to
 - □ Interview a volunteer who is female, 24, with a degree, good mental health about their experience of participating in European Solidarity Corps
 - □ Identify 24-year-old women with degrees & good mental health who volunteered in Eurobarometer to learn other things about them e.g., income trajectory

Quasi-Experiments

- Barely used but most effective tool for overcoming causality conundrum
- Approximate experimental circumstances based on access to 'treatment'
- Builds on **comparative and panel design**: compare two otherwise (almost) identical people/communities who differ only in terms of volunteering ('treatment')
 - ☐ Taylor (2023) used quasi-experimental design to study the effect of the UK's National Citizen Service: matched young people based on gender, ethnicity, education, family socio-economic class, region, religion but who differed in terms of NCS participation
 - □ Found differences in political interest, knowledge, efficacy & anticipated participation stemmed purely from NCS

It's Easy for Me to Say...

Of course, there's another big challenge: **cost**Any of these research designs would increase the time, expertise & money required to identify impact

So, what can we do? Collaborate!

Academics & practitioners are rubbish when it comes to collaborating to study of volunteering, but we have shared goals & can reduce each others' costs



SO WE'RE NOT GOING TO DISCUSS IT?

It's Easy for Me to Say...

It's not straight-forward...

- Goals need to align: we need publications, you need results & the right results
- Timelines need to align: can take years to prepare a publication
- Resources need to align: skills, data, expertise

But we all have a shared interest in producing quality research on volunteering

What else can we do?



Challenges with Identifying the Effects of Volunteering

To find out more:

Email me: <u>S.Fox2@exeter.ac.uk</u>

Sign up to my newsletter (drop me an email)

Thank you for listening!