

ICS Stories: Closing the dataaction gap for patient and population health in Cheshire and Merseyside

8th July 2022 at 13:30

Speakers: Wesam Baker, Mersey Care NHS Foundation Trust and Professor Iain Buchan, University of Liverpool







Closing the Data-Action Gap for Patient Care & Population Health

Webinar, 8th July 2022

Wes Baker
Director of Strategic Analytics, Economics & Population Health Management
Mersey Care NHS Foundation Trust

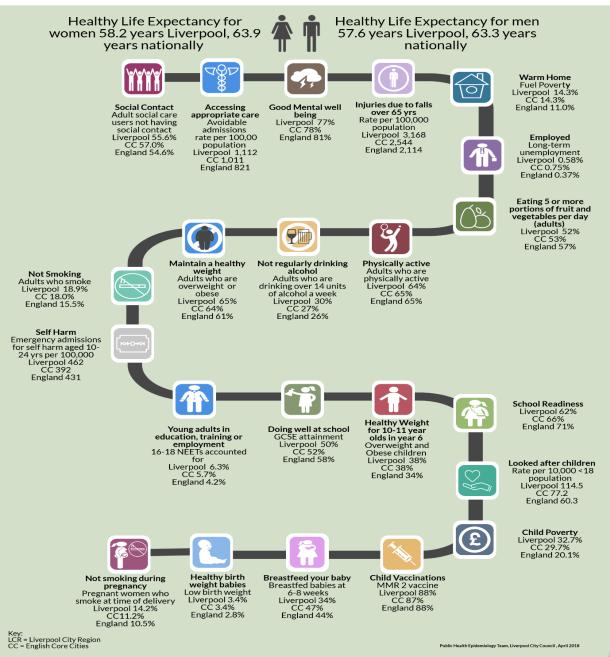
Iain Buchan
Associate Pro Vice Chancellor for Innovation; Professor of Public Health & Clinical Informatics
University of Liverpool

The Need for System-P

"Enabling the NHS to programme equity, rather than just describe inequality"

'The long & winding road to healthy life expectancy'





NHS as civic partner helping communities escape poverty while providing care equitably

Public Health perspective: Liverpool c.f. England

- 4th most deprived upper tier local authority
- Smoking in early pregnancy: 21.5% c.f. 12.8%
- 28% children living in poverty
 (~75% mental health problems arise in childhood)
- 16x more children in care
- School pupils with mental health issue 3.7% c.f. 2.8%
- Common mental disorders: 193 c.f. 156 per 100k
- Cheshire & Mersey: half of 6.3% using mental health services from fifth most deprived neighbourhoods
- Obese at 11 years 25.7% c.f. 21%
- Healthy life expectancy for males: 58.3 c.f. 63.1
- Healthy life expectancy for females: 57.9 c.f. 63.9

https://fingertips.phe.org.uk 2020/1 data

Connected World Disconnected Health Statistics





Poor Antenatal Health



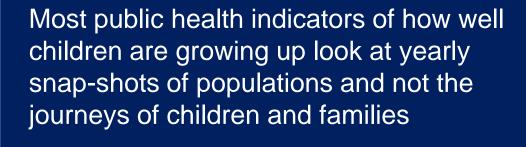
Poor Development



Poor **Child Health**



Poor Learning







Children

in Liverpool



Adult Social and Health Disadvantage

UK's first birth cohort in 20y will study the 'digital twins' of 10k children growing up in Liverpool using NHS and local authority linked data from www.cipha.nhs.uk

Mobilising Data To Support Children and Families

what's your problem? liverpool city region



We asked leaders in the public sector about their stickiest issues, about what opportunities that come from these problems and the barriers they are facing (whether that's adoption, integration, or growth).

We invited SMEs to submit project proposals to tackle three challenges:

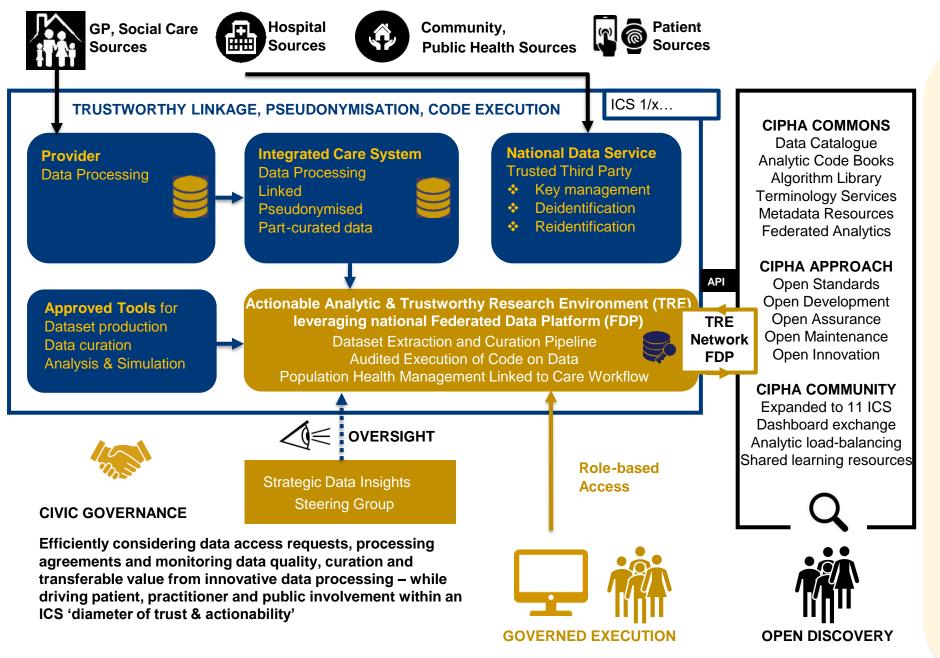
- Increasing family support to stop children entering care
- Reducing childhood asthma through better environments and awareness
- Making digital feel doable in local authority social care teams



We funded **Koala NW** to coproduce a framework of support for families at risk of being separated.

Koala NW will work with vulnerable parents with social care involvement to find out what barriers stopped them asking for help earlier. In coproduction with the parents Koala will produce a set of key indicators to be used as a framework to identify and improve pathways for parents seek help sooner.

Aim: to normalise families asking for help in the first 1001 days of a baby's life to prevent children from going into care settings.



- 3 key components for closing the data-action gap
 - People-to-data:

Spring 2019
Liverpool City Region
Civic Data Cooperative
civicdatacooperative.com

Data-to-analysis:

Summer 2020

NHS/social care/public health record for 2.7m stood up in 90 days "Combined Intelligence for Population Health Action" cipha.nhs.uk

Analysis-to-action:

Autumn 2020 World's first universal voluntary Covid-19 rapid antigen testing

2021/2...

System-P brings public health actionable analytics on CIPHA into the build of Cheshire & Mersey ICS

- Ainsworth J, Buchan I. Combining Health Data Uses to Ignite Health System Learning. Methods Inf Med. 2015;54(6):479-87
- Buchan I, National Grid of Civic Data Cooperatives for Health in The Health of the Nation February 2020
- www.liverpool.ac.uk/coronavirus/research-and-analysis/covid-smart-pilot/

COVID-19 Wake-up Call Need for Real-time Joined-up Intelligence

Tests & Cases

All Cheshire & Merseyside (C&M) residents tested at any Pillar 2 test site and non-C&M residents tested at a C&M test site

Note: this report does not include Pillar 1 data.

Note: positivity rate calculations do not follow PHE methodology

Show Filters

Clear Filters

Filters Applied



17,822,321

Tests Completed (LFT+PCR)

2,423,299 Individuals Tested

(LFT+PCR)

777.092 LFT Case Positivity

Positive (LFT+PCR)

PCR Case Positivity

Rate (Symptomatic)

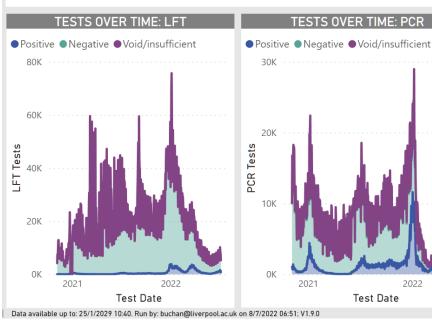
15.07%

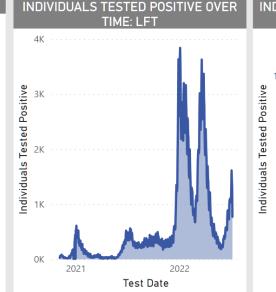
TEST COUNTS						
Test Kit	Tests Completed	Positive Tests	Negative Tests	Void / Insufficient Tests		
	366,368	19,828	333,668	12,872		
LFT	12,994,470	382,103	12,597,464	14,903		
PCR	4,461,483	659,097	3,715,779	86,607		
Total	17,822,321	1,061,028	16,646,911	114,382		

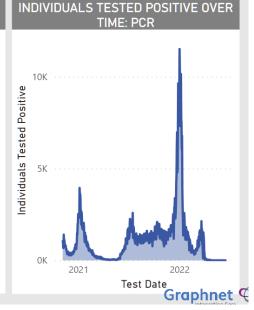
INDIVIDUALS TESTED POSITIVE						
Test Kit	Individuals Tested	Individuals Tested Positive	Case Positivity Rate (Individuals)			
	110,137	19,362	5.61%			
LFT	1,644,510	286,236	2.94%			
PCR	1,584,148	609,883	15.07%			
Total	2,423,299	777,092	5.99%			

2.94%

Rate (Asymptomatic)









Persistent longitudinal health & care record with real-time population analytics linkable to care workflows / notifications for Cheshire & Mersey 2.7m ICS population



Putting Data to Work for Residents at Pandemic Pace









World's first city-scale voluntary SARS-CoV-2 rapid antigen testing:
Cases fell by a 21% and hospitalization by 43% then 25% overall even after reopening in lower Tier



<u>First reopening of mass gatherings</u> <u>regional cluster in N. Hemisphere</u>: only 12 cases / >13k eventgoers

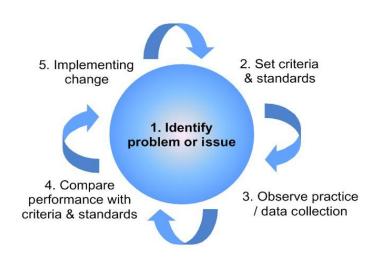




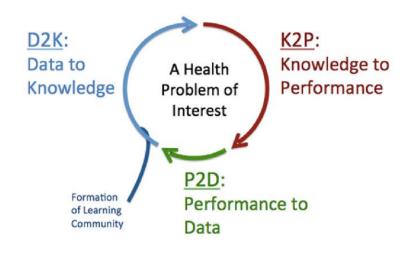
UK's first <u>Civic Data Cooperative</u>, and <u>CIVC NHS Combined Intelligence</u>

From Nice- to Must-have System Resilience

Clinical Audit \rightarrow Governance (1980s/90s \rightarrow 00s)



Learning Health Systems (10s)

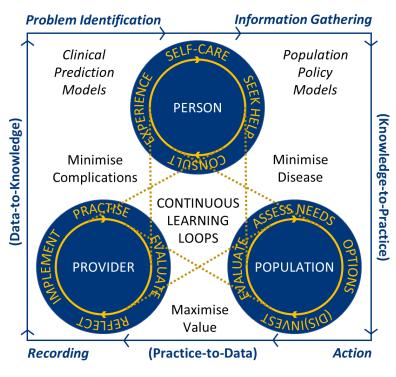


Integrated care record data

Fuller longitudinal prediction (difficulty learning models from data)

2-way translation aim (more practice-based evidence)

Pandemic-resilient Systems (20s)



Wider health & care data

Multi-outcome prediction (system dynamics; causal, counterfactual AI)

Combined intelligence (system resilience; embedded trials)

Specialist registry data

Crude predictive models (baseline risk factor loaded)

1-way translation (trials evidence into practice)

System-P into Action

Deploying in Cheshire and Merseyside Integrated Care System, a preventative, predictive, precise approach to population, patient and person in a joined-up intelligence-led way

What is **System-P**?











PAYMENT



System P

DATA

ANALYTICS

CARE

nme

Enabling us to programme equity, rather than tackle inequality

genuinely change public services in order to manage population health more effectively.

Data – Flow and integrate all data necessary to understand and improve residents' health and care journeys.

Analytics – Network and invest in the data analysts and technologies

delivering a world-leading combined health intelligence cooperative.

Care – Optimise care processes to improve outcomes and reduce cost as an integrated, health and care system.

Payment – ensure providers are being paid fairly for the value they provide in terms of individual patient and population care.

At a glance : At a glance : https://youtu.be/tz1Do_tUTpl

System P is the whole system approach to addressing multiagency, multisector challenges that negatively impact population health and will deliver transformational change in service provision through collaborative working.

It is based on four 'DACP' pillars of work – **data** transformation **analytical** transformation, **care** transformation and **payment** transformation – to reflect our belief that transformation in all four areas is required if we are to

System-P: Hackathon

- 1. Hackathon 10th of November 2021
- 2. Hackathon 27th April 2022





How is the Complex Lives Segment Defined?

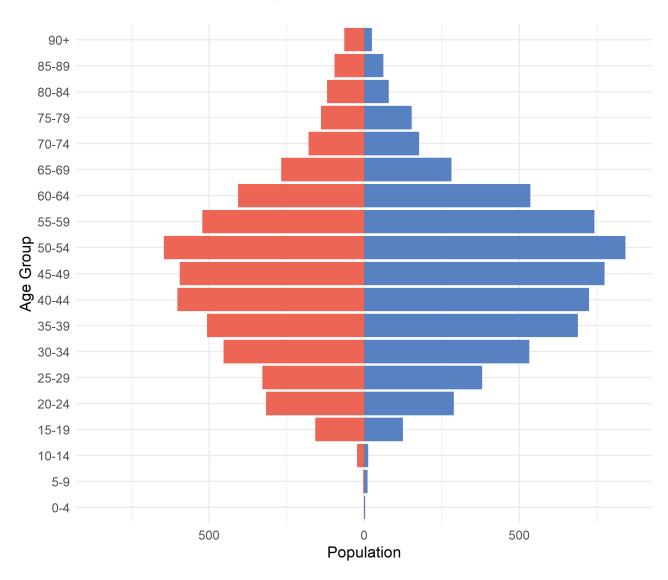


In Cheshire & Merseyside ICS 11,857 individuals (0.6% of the population) were identified as belonging to the Complex Lives segment. In this pack we describe the characteristics of people in this segment, before moving on to describe their other healthcare issues and how they use services.

Patient Characteristics: Age & Gender

For the Complex Lives segment the mean average age of these patients is 48 (interquartile range from 36 to 58).

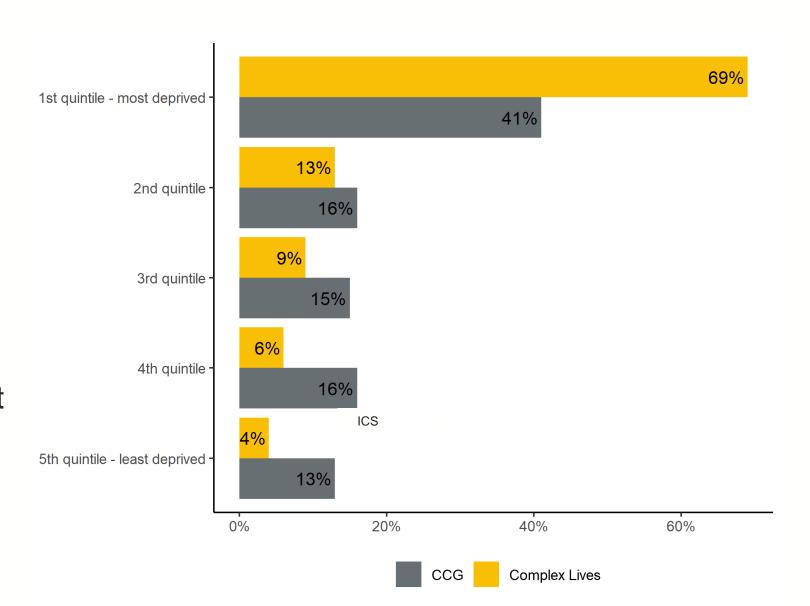
Gender splits within the segment are 54% male and 46% female.



Patient Characteristics: Deprivation

Those with Complex Lives are more likely to reside in areas of higher deprivation. 69% of the segment live in the most deprived quintile.

Proportionately, this is 1.7 times the share of the CCG population living in the most deprived quintile.



Patient Characteristics: Living Arrangements

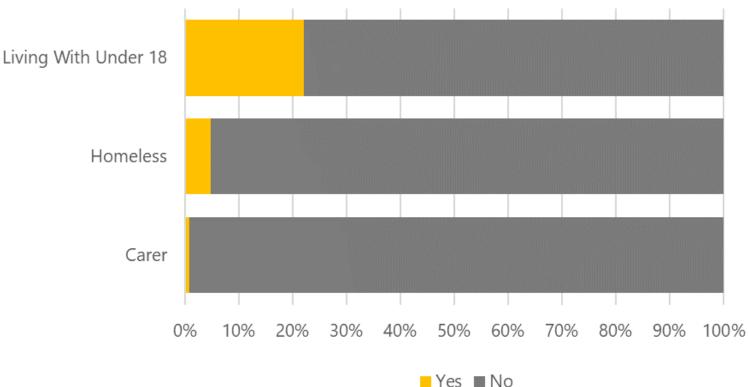
Those in the Complex Lives segment are found to be living with, on average, 1.6 other people. Therefore, beyond the Complex Lives individuals already identified, an additional estimated 18,473 people are also affected by Complex Lives. A number of these will be children.

For the Complex Lives segment 2,424 people (22%) are identified as living in a household with someone

under 18.

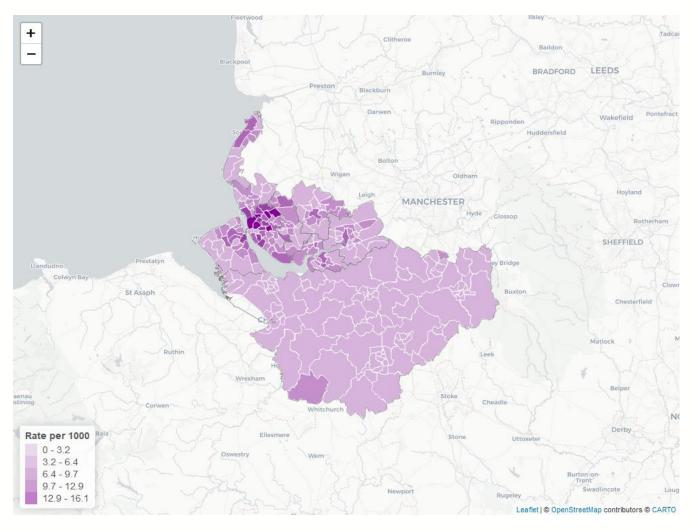
5% of the segment have experienced homelessness in the last 2 years and 1% have caring responsibilities.

44 people were found with a status of asylum seeker.



Patient Characteristics: Geography

The map shows, for wards within the ICS, The rate of Complex Lives individuals per 1,000 population.



Areas with some of the highest density for Complex Lives are:

- Norris Green (Liverpool)
- Everton (Liverpool)
- Kirkdale (Liverpool)

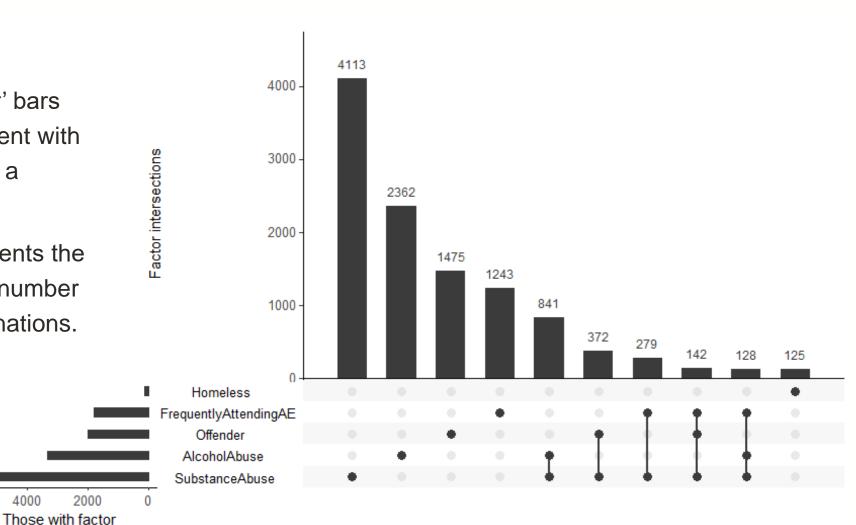
Patient Characteristics: Complex Factors

2000

The definition for an individual to be assigned to the Complex Lives segment relates to certain factors about that person. For those in the Complex Lives segment the largest factor is substance abuse.

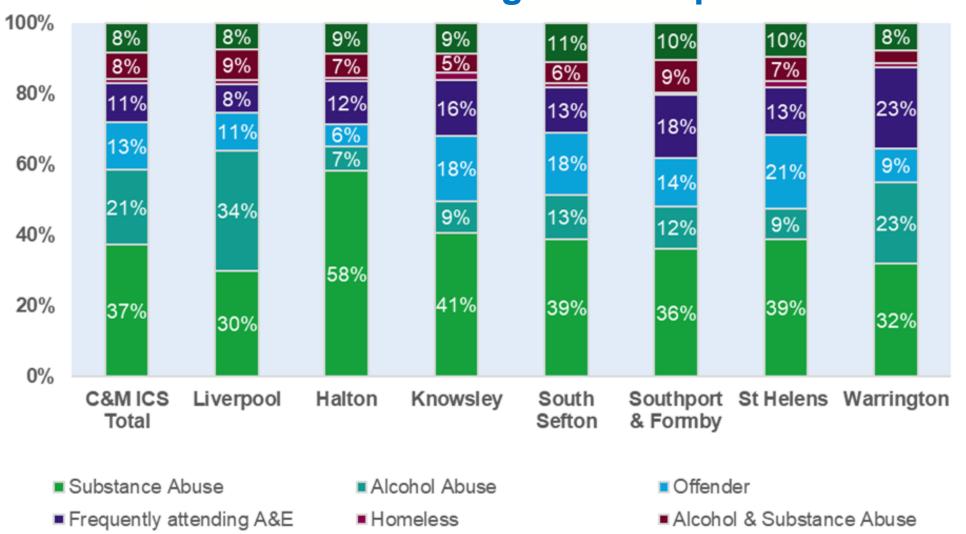
The 'Those with complex factor' bars represent all those in the segment with those factors or components of a 'complex life'.

The 'Factor intersection' represents the combination of factors and the number of individuals with those combinations.



Characteristics: Contributing Factors

Factors Contributing to a Complex Life



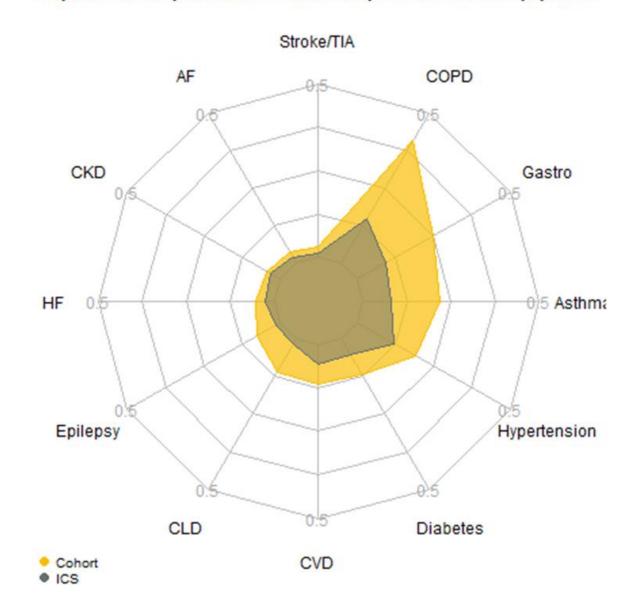
■ Any other combinations

Long Term Conditions in the Population

Proportion of LTC prevalence in cohort compared to overall ICS population

For the specified long term conditions a comparison of prevalence rates is made between those in the Complex Lives segment and the total population (aged 15+). This indicates the scale of the difference in these disease areas between the segment and the total population.

The scale shows the rate per person so 0.5 represents prevalence of 50% of people.



Health & Care Use: A&E Services

Those in the Complex Lives segment attend A&E services on average **2.9** times per person, per year. This is much higher than the total population who attend A&E services **0.4** times per person, per year. Emergency Departments are the most used A&E service and also the service where there is the greatest disparity in use between segment and total population. In a year **61%** of people in the Complex Lives segment attend an A&E service. For the total population the same figure is **22%**.

When attending A&E services the average cost per attendance is £153 for those in the Complex Lives segment. This is 10% higher than the average cost per attendance for the total population.

Where a clinical reason for attending A&E has been recorded this identifies that for the Complex Lives segment common reasons for attending A&E relate to *Trauma / musculoskeletal* or *Psychosocial / Behaviour change* problems.

Health & Care Use: Emergency Admissions

On average those in the Complex Lives segment have 1.1 emergency admissions per person, per year. This is again much higher than the total population who have on average **0.1** emergency admissions per year.

40% of people in Complex Lives segment have an emergency admission in a year. For the total population the same figure is lower at **7%**.

The average emergency admission cost is £2,152 for Complex Lives compared to £2,141 for the total population.

When those in the Complex Lives segment are admitted as an emergency common reasons for admission relate to *Poisoning Toxic Effects Special Examinations Screening and Other Healthcare Contacts* and *Treatment of Mental Health Patients by Non-Mental Health Service Providers*.

Health & Care Use: Planned Care

Those in the Complex Lives segment use more planned admissions, both as electives and as daycases, on average per person, per year. They also use more outpatient attendances.

In a year **15**% of people in Complex Lives segment have a planned admission. For the total population the same figure is **7**%. For outpatient attendances **63**% of those in the Complex Lives segment attended at least one outpatient appointment in a year compared to **35**% for the total population.

For elective planned care the Complex Lives segment are often admitted with a reason of *Orthopaedic Non-Trauma Procedures*. In daycase admissions their main reason for admission is related to *Digestive System Endoscopic Procedures*. For outpatients their most common clinical specialty is *Gastroenterology* (first attendances) and also *Gastroenterology* (follow-up attendances).

Health & Care Use: Mental Health

On average those in the Complex Lives segment have **7.7** mental health contacts per person, per year. This is much higher than the total population who on average have **0.3** contacts with mental health services per year.

53% of people in the Complex Lives segment have an contact with mental health services in a year. For the total population the same figure is only **4%**.

When those in the Complex Lives segment are in contact with mental health this most commonly involves contacts with *Community Mental Health Team – Functional* or *Crisis Resolution Team/Home Treatment Service* teams.

Health & Care Use: Community Services

On average those in the Complex Lives segment have **7.2** contacts with community services per person, per year. This is higher than the total population who, on the same basis, have **1.7**.

30% of people in the Complex Lives segment are in contact with community services in a year. For the total population the figure is **21%**.

The most used community service for the Complex Lives segment is District Nursing Service.

Health and Care Use: Social Care

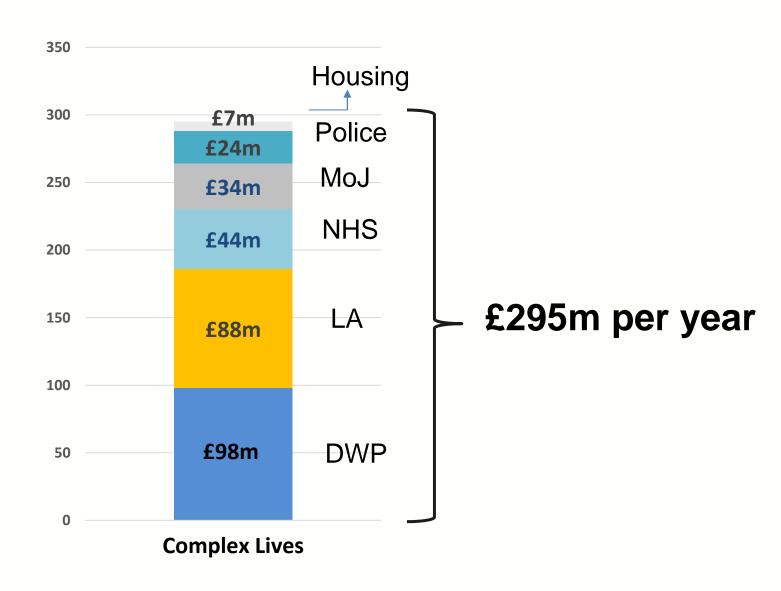
From available data there were on average **0.15** of the Complex Lives segment known to social services in the last year. More than the total population where the same figure is **0.02**. However, social services data should be viewed as indicative and treated with caution. A recent review of the data indicated concerns with the data and further work is already underway to improve the consistency and quality of social care data.

Social services data includes information collected by councils and does not include services purchased directly by patients or provided by the voluntary sector.

The CIPHA platform gives us the opportunity to network intelligence consistently across C&M.

System P provides the methodology and approach to using this practically to effect change.

Each year C&M ICS, we spend c.£295m to support c.11,857 people with complex lives



Data source: New Economy Manchester Unit Cost Database.

Frailty & Dementia Segment Definition

People are defined as belonging to the Frailty and Dementia segment if they have:

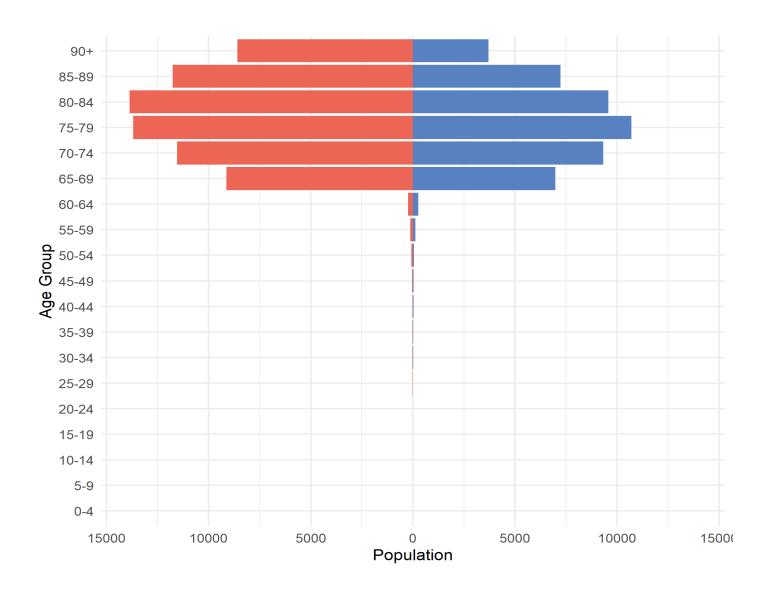


In Cheshire & Merseyside ICS 117,243 individuals (5.6% of the population) were identified as belonging to the Frailty and Dementia segment. In this pack we describe the characteristics of people in this segment, before moving on to describe their other healthcare issues and how they use services.

Patient Characteristics: Age & Gender

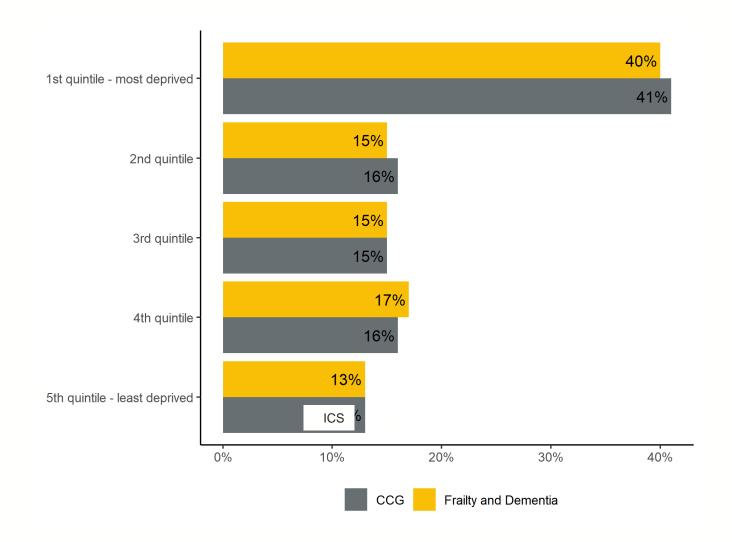
For the Frailty and Dementia segment the mean average age of these individuals is 79 (interquartile range from 73 to 85).

Gender splits within the segment are 41% male and 59% female.



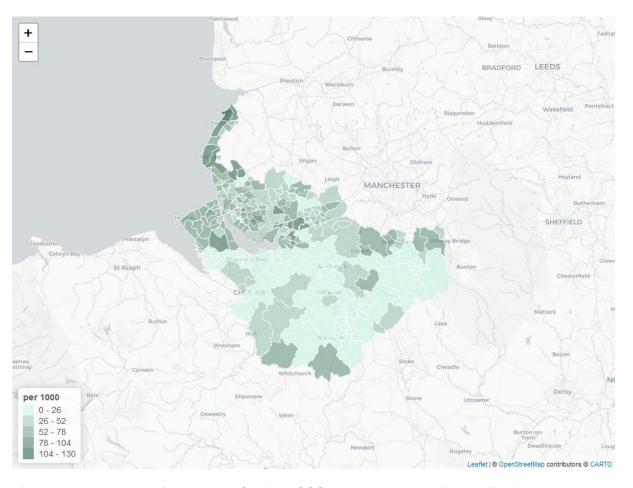
Patient Characteristics: Deprivation

Those with Frailty and Dementia are more likely to reside in areas of higher deprivation. 40% of the segment live in the most deprived quintile.



Patient Characteristics: Geography

The map shows, for wards within the ICS, the rate of Frailty and Dementia individuals per 1,000 population.



Areas with some of the highest density for Frailty and Dementia are:

- Bankfield (Halton)
- Duke's (Southport & Formby)
- Cambridge (Southport & Formby)

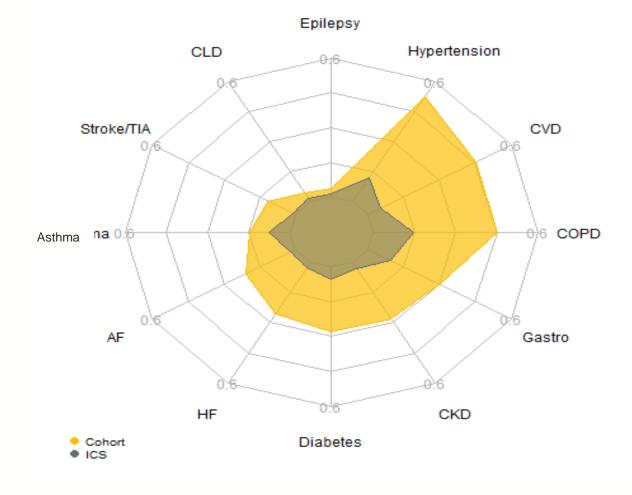
Due to the low sign up of practices in Cheshire CCG rates in this area will be artificially low.

Long-term Conditions in the Population

For the specified long term conditions a comparison of prevalence rates is made between those in the Frailty and Dementia segment and the total population (aged 15+). This indicates the scale of the difference in these disease areas between the segment and the general population.

The scale shows the rate per person so 0.6 represents prevalence of 60% of people.

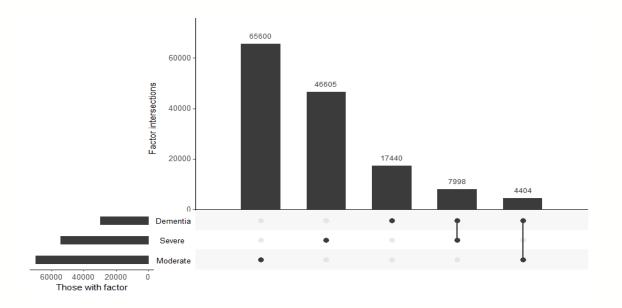
Proportion of LTC prevalence in cohort compared to overall ICS population



Patient Characteristics: Factors

For an individual to be assigned to the Frailty and Dementia segment depends on factors for the level of frailty (moderate or severe) or a clinical code for dementia. Numbers are larger frailty alone but some individuals do have both frailty and dementia.

The 'Those with factor' bars represent all those in the segment with those factors. The 'Factor intersection' represents the combination of factors and the number of individuals with those combinations.



Although not a factor, there was interest in identifying those prescribed an anticholinergic. In the last year 80% of those in the segment were prescribed these at least once, and where prescribed there were on average 2.4 different types.

Frailty & Dementia: Anticholinergic Burden

The cumulative effect of taking one or more medicine with anticholinergic properties (anticholinergic burden) increases the risk of:



Delirium and falls (shortterm adverse outcomes)



Hospitalisation



Dementia, loss of physical function, loss of independence (long-term adverse outcomes)



Mortality

System-P Action-research

Many researchers are highly-motivated by NHS problem-solving, and new AI solutions need careful academic partnership to make the most of ICS integrated care records

'Big Data' is Not Enough for Big Insight

Explaining Variation

Big Enough Data

Understanding Mechanisms

Causal inference and counterfactual prediction in machine learning for actionable healthcare | Nature Machine Intelligence

Why Most Published Research Findings Are False John P.A. Ioannidis

Summary

There is increasing concern that most current published research findings are false. The probability that a research claim is true may depend on study power and bias, the number of other studies on the same question, and, importantly, the ratio of true to no relationships among the relationships probed in each scientific field. In this framework, a research finding is less likely to be true when the studies conducted in a field are smaller; when effect sizes are smaller; when there is a greater number and lesser preselection of tested relationships; where there is greater flexibility in designs, definitions, outcomes, and analytical modes; when there is greater financial and other interest and prejudice; and when more teams are involved in a scientific field in chase of statistical significance.

Simulations show that for most study designs and settings, it is more likely for most likely for most likely for most likely for most likely for more likely for more

factors that influence this problem and some corollaries thereof.

Modeling the Framework for False Positive Findings

Several methodologists have pointed out [9–11] that the high rate of nonreplication (lack of confirmation) of research discoveries is a consequence of the convenient, yet ill-founded strategy of claiming conclusive research findings solely on the basis of a single study assessed by formal statistical significance, typically for a p-value less than 0.05. Research is not most appropriately represented and summarized by p-values, but, unfortunately, there is a widespread notion that medical research articles

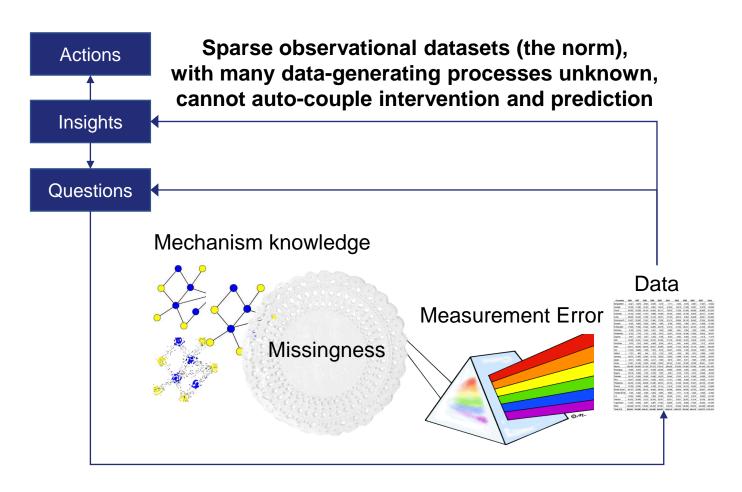
It can be proven that most claimed research findings are false.

is characteristic of the field and car vary a lot depending on whether the field targets highly likely relationships or searches for only one or a few true relationships among thousands and millions of hypotheses that may be postulated. Let us also consider, for computational simplicity, circumscribed fields where either there is only one true relationship (among many that can be hypothesized) or the power is similar to find any of the several existing true relationships. The pre-study probability of a relationship being true is R/(R+1). The probability of a study finding a true relationship reflects the power 1 - B (one minus the Type II error rate). The probability of claiming a relationship when none truly exists reflects the Type I error rate, a. Assuming that e relationships are being probed in the field, the expected values of the 2 x 2 table are given in Table 1. After a research finding has been claimed based on

Big data * small research → noise

Biology ∪ **Behaviour** ∪ **Environment**

Interactions may be emergent (don't rely on simulated biology)



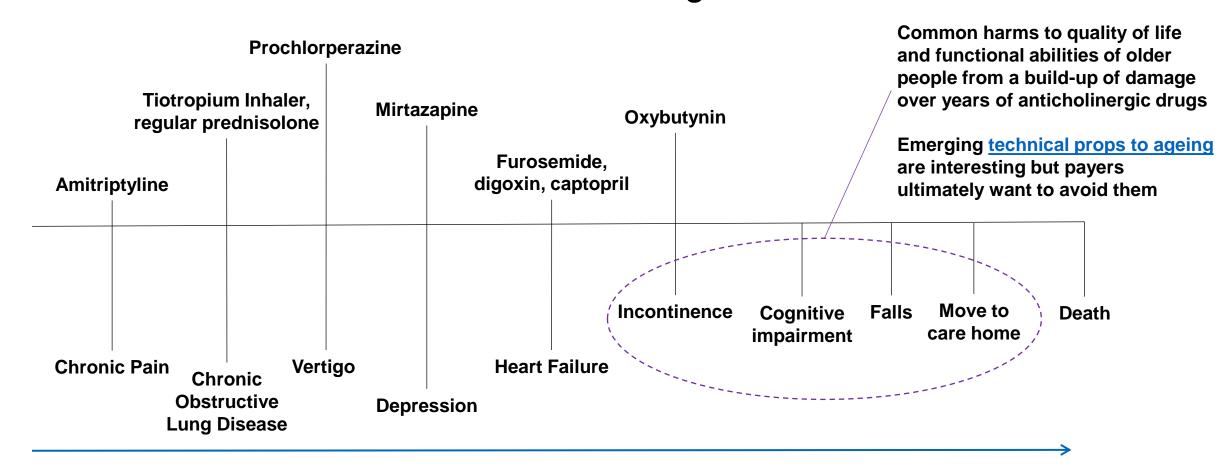
Computers Will Soon Learn to 'See' High-risk Patient Journeys Early

Some machine learning methods are becoming very useful for discovering causes of diseases and seeing 'statistical structure' in data in very high definition – especially when there are multiple observations over time per individual:

- TARNET (estimating individual treatment effects)
- CEVAE (autoencoders with causal structure)
- Dragonnet (doubly robust counterfactual prediction of treatment effects)
- PSSAM-GAN (generating synthetically balanced case-control populations)
- DCN-PD (deep counterfactual learning on propensity scores)
- To watch: treatment effects with <u>dynamic regimens</u> (~ g-estimation)
- <u>Transformers</u>: e.g. GatorTron (8.9bn parameter model: mining clinical text)

From Risk-stratification to Integrated Actions

Mental vs physical healthcare dissonance is pervasive, however, neither research nor care are integrated across conditions, disciplines and the life-course. Consider anti-cholinergic chaos...



Reducing Valproate Harms Population Insight to Patient Action on CIPHA

Valproate poses significant risk to the unborn child throughout pregnancy, including:

- 10% risk of major malformations on average and 25% at higher doses
- 30-40% of significant reduction in IQ

Withdrawal of valproate from the market would be harmful as it:

- Is the most effective treatment for idiopathic generalised epilepsy (about a third of epilepsy ~50,000 in Cheshire & Merseyside). Withdrawal would result in otherwise avoidable deaths and morbidity.
- Is an effective treatment in bipolar disorder. Withdrawal of valproate would deny access to an effective treatment.

System-P action on CIPHA being explored:

- Identify women of childbearing age taking valproate (and not attending secondary care)
- Provide secondary care with dashboards to invite affected women not seen recently to clinic
- Provide GPs with prescribing audit and feedback on women not receiving annual counselling
- Provide patients with the option of awareness raising text messages / reminders





Questions?



