Digital Population Twins in JUNE

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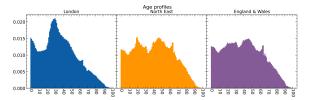
IDAS & IPPP

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F. Krauss Digital Population Twins in JUNE motivation: why granularity matters in epidemiology

impact of COVID=19 highly age-dependent → need geographical granularity for regional planning

(coincidence: Durham hosts & maintains England & Wales census data of past decades)



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IDAS & IPPP

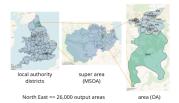
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inputs: demographics

• last census (2011)

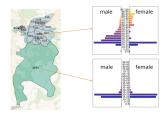
(data freely available from Office for National Statistics)

hierarchical data structure



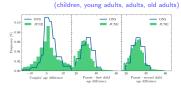
• OA's with ~ 250 residents, with similar characteristics

- build virtual population in OA: age, gender, ethnicity, deprivation index
- example: Durham



virtual households

- correct compositions important: primary place for infections
- household composition in 20 categories at OA level



• also: communal facilities

(carehomes)

- further test: interplay with social mixing
- example: North-East England

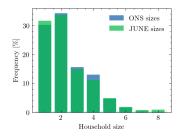
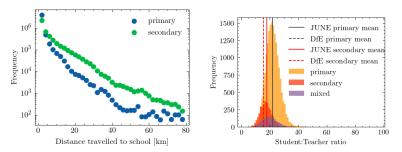


Image: Image:

virtual schools

- information about schools: age range of children and locations
- send kids to nearest age-appropriate structure
- could modulate this with school sizes, if necessary

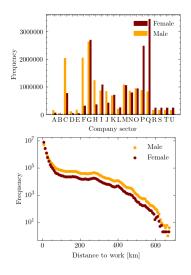


work & virtual companies

- workers and companies in ~ 20 macro-sectors at MSOA level
- know age/sex distribution in sectors nation wide
- distribute workers over companies (we know their sizes in bins)

(construct a big origin-destination matrix & optimise)

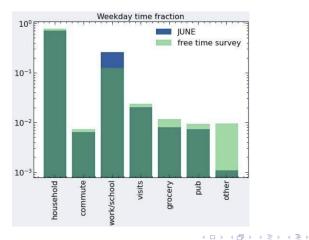
 information about commute mode: public vs. private



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daily activities: average time budget

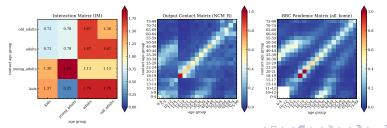
- compare our simulation with data from ONS
- average time spent per day in different activities



social mixing matrices

- denote number of contacts of person with age *i* with person of age *j*
- somewhat tricky format: averages over full population
 - \longrightarrow have to normalise it to our fractured social environment
 - \rightarrow interplay with household sizes etc. (may have to be fitted?)
- usually used in compartment models (SEIR)
 - → compartments often organised by age/sex with limited geographic or sociological granularity
- example: household interactions vs. BBC pandemics project

(census has 4 categories of residents: kids, young adults, adults, old adults)



summary

- constructed an individual-based model with supreme granularity: demography, geography, sociology
- model informed operational planning of NHS:
 - $\rightarrow\,$ early warning of second wave
 - $\rightarrow\,$ projections for school re-opening, Delta and Omicron waves
 - $\rightarrow\,$ understanding of transmission sociology
- code is highly flexible:
 - $\rightarrow\,$ addition of new effects & policies relatively painless
 - $\rightarrow\,$ adaptation to new environments: Cox's Bazaar
 - $\rightarrow\,$ adaptation to New Zealand underway
- challenge to widespread perception in computational sociology: more and better detail often helps