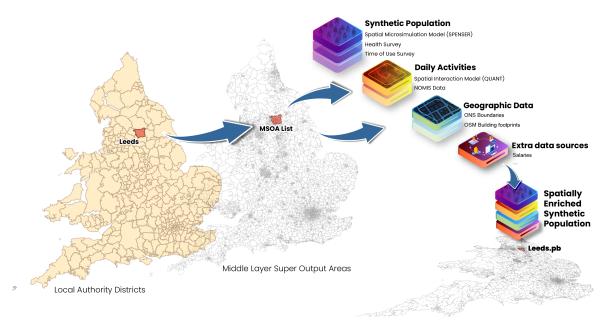
Synthetic Population Catalyst

Table of contents

1	Getting started	4
I	Using SPC	6
2	Outputs for England Counties 2.1 Versioning	7
3	Using the SPC output file 3.1 Python	15 15 15 16 16
4	Installation 4.1 Compiling SPC	18 18 18
5	Creating new study areas 5.1 Run SPC for the new area	19 19 19 20
H	Understanding SPC	21
6	Data schema6.1 Flows: modelling daily activites6.2 Flow weights	22 22 23
7	Modelling methods 7.1 Income data	24 25 25 26 20

8	Data sou 8.1 Cour		31 31
	8.2 Nati	ional data	33
Ш	Advanc	ed	35
9	Develope	er guide	36
	9.1 Code	le hygiene	36
	9.2 Som	ne tips for working with Rust	36
	9.3 Docl	ker	37
10	Code wal	lkthrough	38
	10.0	0.1 Split code into two stages	38
	10.0	2.2 Explicit data schema	38
	10.0	0.3 Type-safe IDs	39
	10.0	0.4 Idempotent data preparation	39
	10.0	0.5 Logging with structure	39
		0.6 Determinism	40
		tocol buffers	41
		example of the power of static type checking	41
11	Performa	ince	43

1 Getting started



The Synthetic Population Catalyst (SPC) makes it easier for researchers to work with synthetic population data in England. It combines a variety of data sources and outputs a single file in protocol buffer format, describing the population in a given study area. The data includes demographic, health, and daily activity data per person, and information about the venues where people conduct activities.

You can use SPC output to catalyze your own project. Rather than join together many raw data sources yourself and deal with missing and messy data, you can leverage SPC's effort and well-documented schema.

To get started:

- 1. Download sample data for a county in England
- 2. Explore how to use the data
- 3. If you need a different study area, build and then run SPC

You can also download this site as a PDF and find all code on Github.

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Part I Using SPC

2 Outputs for England Counties

You don't need to run SPC yourself. See config/ for the list of MSOAs covered by each study area. If you want to run SPC for a different list of MSOAs, see here.

- England/2012/bedfordshire.pb.gz
- England/2012/berkshire.pb.gz
- England/2012/bristol.pb.gz
- $\bullet \ \ England/2012/buckinghamshire.pb.gz$
- England/2012/cambridgeshire.pb.gz
- England/2012/cheshire.pb.gz
- England/2012/cornwall.pb.gz
- England/2012/cumbria.pb.gz
- England/2012/derbyshire.pb.gz
- England/2012/devon.pb.gz
- England/2012/durham.pb.gz
- England/2012/east-sussex.pb.gz
- England/2012/east-yorkshire-with-hull.pb.gz
- England/2012/essex.pb.gz
- England/2012/gloucestershire.pb.gz
- England/2012/greater-london.pb.gz
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- England/2012/herefordshire.pb.gz
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- England/2012/isle-of-wight.pb.gz
- England/2012/kent.pb.gz
- England/2012/lancashire.pb.gz
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- England/2039/bedfordshire.pb.gz
- England/2039/berkshire.pb.gz
- England/2039/bristol.pb.gz
- England/2039/buckinghamshire.pb.gz
- England/2039/cambridgeshire.pb.gz
- England/2039/cheshire.pb.gz
- England/2039/cornwall.pb.gz
- England/2039/cumbria.pb.gz
- England/2039/derbyshire.pb.gz
- England/2039/devon.pb.gz
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- England/2039/east-sussex.pb.gz
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- England/2039/west-sussex.pb.gz
- England/2039/west-yorkshire.pb.gz
- England/2039/wiltshire.pb.gz
- England/2039/worcestershire.pb.gz
- special/2012/northwest transpennine.pb.gz
- special/2020/northwest transpennine.pb.gz
- special/2022/northwest transpennine.pb.gz
- special/2032/northwest_transpennine.pb.gz
- special/2039/northwest transpennine.pb.gz
- Wales/2012/bridgend-and-neath-port-talbot.pb.gz
- Wales/2012/cardiff-and-vale-of-glamorgan.pb.gz
- Wales/2012/central-valleys.pb.gz
- Wales/2012/conwy-and-denbighshire.pb.gz
- Wales/2012/flintshire-and-wrexham.pb.gz
- \bullet Wales/2012/gwent-valleys.pb.gz
- Wales/2012/gwynedd.pb.gz
- Wales/2012/isle-of-anglesey.pb.gz
- Wales/2012/monmouthshire-and-newport.pb.gz
- Wales/2012/powys.pb.gz
- Wales/2012/south-west-wales.pb.gz
- Wales/2012/swansea.pb.gz

- Wales/2020/bridgend-and-neath-port-talbot.pb.gz
- Wales/2020/cardiff-and-vale-of-glamorgan.pb.gz
- Wales/2020/central-valleys.pb.gz
- Wales/2020/conwy-and-denbighshire.pb.gz
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- Wales/2032/gwent-valleys.pb.gz
- Wales/2032/gwynedd.pb.gz
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- $\bullet \ \ Wales/2039/bridgend-and-neath-port-talbot.pb.gz$
- Wales/2039/cardiff-and-vale-of-glamorgan.pb.gz
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- Wales/2039/monmouthshire-and-newport.pb.gz
- Wales/2039/powys.pb.gz
- Wales/2039/south-west-wales.pb.gz
- Wales/2039/swansea.pb.gz

If you use SPC code or data in your work, please cite using the Zenodo DOI (using the bottom-right tool to generate the citation).

2.1 Versioning

Over time, we may add more data to SPC or change the schema. Protocol buffers are designed to let combinations of new/old code and data files work together, but we don't intend to use this feature. We may make breaking changes, like deleting fields. We'll release a new version of the schema and output data every time and document it here. You should depend on a specific version of the data output in your code, so new releases don't affect you until you decide to update.

- v1: released 25/04/2022, schema
- v1.1, released 27/05/2022, schema
 - added pwkstat, salary_hourly, salary_yearly, and idp
 - reorganized Identifiers and Employment attributes
 - non-breaking change added 02/08/2022: added bmi_new field
- v1.2, released 29/12/2022, schema
 - switched to proto2 and made some fields optional
 - adjusted some numeric enum values to match ONS
- v2, released 09/03/2023, schema
 - new per-person and per-household fields
 - various changes to existing fields (adjusting enum number, removing the BMI enum, etc)
 - adding time-use diaries
 - expanding to Wales
 - adding multiple years of output

3 Using the SPC output file

Once you download or generate an SPC output file for your study area, how do you use it? Each study area consists of one .pb or protocol buffer file. This file efficiently encodes data following this schema. Read more about what data is contained in the output.

You can read the "protobuf" (shorthand for a protocol buffer file) in any supported language, and then extract and transform just the parts of the data you want for your model.

We have examples for Python below, but feel free to request other languages.

We have a web app using Svelte to interactively explore SPC data. Its source code is great reference for how to use the proto output.

3.1 Python

To work with SPC protobufs in Python, you need two dependencies setup:

- The protobuf library
 - You can install system-wide with pip install protobuf
 - Or add as a dependency to a conda, poetry, etc environment
- The generated Python library, synthpop_pb2.py
 - You can download a copy of this file into your codebase, then import synthpop_pb2
 - You can also generate the file yourself, following the docs: protoc --python_out=python/ synthpop.proto

3.1.1 Converting to Pandas data-frames and CSV

The schema expresses relationships between people, households, and venues that can't all be captured by a simple 2D table. Nevertheless, you can extract per-person information and express as a dataframe or CSV file. See this example Python script for inspiration. You can try it out:

```
# Download a file
wget https://rampOstorage.blob.core.windows.net/spc-output/v1/rutland.pb.gz
```

```
# Uncompress
gunzip rutland.pb.gz
# Convert the .pb to JSON
python3 python/protobuf_to_csv.py --input_path data/output/rutland.pb
# View the output
less people.csv
```

3.1.2 Converting .pb file to JSON format

To interactively explore the data, viewing JSON is much easier. It shows the same structure as the protobuf, but in a human-readable text format. The example below uses a small Python script:

```
# Download a file
wget https://rampOstorage.blob.core.windows.net/spc-output/v1/rutland.pb.gz
# Uncompress
gunzip rutland.pb.gz
# Convert the .pb to JSON
python3 python/protobuf_to_json.py data/output/rutland.pb > rutland.json
# View the output
less rutland.json
```

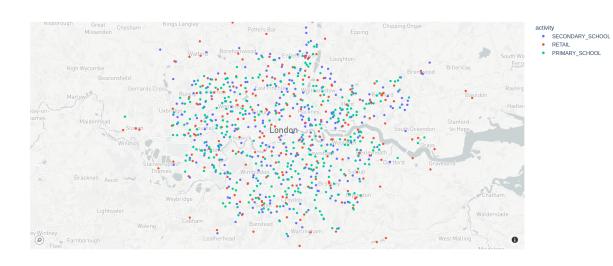
3.1.3 Converting to numpy arrays

The ASPICS project simulates the spread of COVID through a population. The code uses numpy, and this script converts the protobuf to a bunch of different numpy arrays.

Note the ASPICS code doesn't keep using the generated Python protobuf classes for the rest of the pipeline. Data frames and numpy arrays may be more familiar and appropriate. The protobuf is a format optimized for reading and writing; you don't need to use it throughout all of your model code.

3.1.4 Visualizing venues

Use this script to read a protobuf file, then draws a dot for every venue, color-coded by activity.



4 Installation

You only need to compile SPC to run for a custom set of MSOAs. Just download existing output if your study area matches what we provide.

• Rust: The latest stable version of Rust: https://www.rust-lang.org/tools/install

4.1 Compiling SPC

```
git clone https://github.com/alan-turing-institute/uatk-spc/
cd uatk-spc
# The next command will take a few minutes the first time you do it, to build external dep
cargo build --release
```

4.2 Troubleshooting downloading

If you get an error No such file or directory (os error 2) it might be because a previous attempt to run SPC failed, and some necessary files were not fully downloaded. In these cases you could try deleting the data/raw_data directory and then running SPC again. It should automatically try to download the big files again.

If you have trouble downloading any of the large files, you can download them manually. The logs will contain a line such as Downloading https://rampOstorage.blob.core.windows.net/nationaldata/to data/raw_data/nationaldata/QUANT_RAMP_spc.tar.gz. This tells you the URL to retrieve, and where to put the output file. Note that SPC won't attempt to download files if they already exist, so if you wind up with a partially downloaded file, you have to manually remove it.

5 Creating new study areas

If the area you want to model isn't already generated, then you can follow this guide to run SPC on a custom area. You must first compile SPC.

SPC takes a newline-separated list of MSOAs in the config/ directory as input, like this. You can generate this list from a LAD (local authority district). From the main SPC directory, run python scripts/select_msoas.py. Refer to data/raw_data/referencedata/lookUp.csv (only available after running SPC once) for all geographies available.

This script will create a new file, config/your_region.txt.

5.1 Run SPC for the new area

From the main directory, just run:

```
cargo run --release -- config/your_region.txt
```

This will download some large files the first time. You'll wind up with data/output/your_region.pb as output, as well as lots of intermediate files in data/raw_data/. The next time you run this command (even on a different study area), it should go much faster.

5.2 (Optional) run SPC for lots of areas

If you want to run the program over lots of areas at once and are using Mac/Linux, you can use a for loop in a terminal to repeatedly run SPC over all files in the config directory. For example, this will run SPC on all .txt files in the config directory:

```
for file in config/*.csv; do cargo run --release -- config/$file; done
```

5.3 Using the output

After you generate the files, see here for how to use them in your project.

If you use SPC code or data in your work, please cite using the Zenodo DOI (using the bottom-right tool to generate the citation).

Part II Understanding SPC

6 Data schema

Here are some helpful tips for understanding the schema.

Each .pb file contains exactly one Population message. In contrast to datasets consisting of multiple .csv files, just a single file contains everything. Some of the fields in Population are lists (of people and households) or maps (of venues keyed by activity, or of MSOAs). Unlike a flat .csv table, there may be more lists embedded later. Each Household has a list of members, for example.

The different objects refer to each other, forming a graph structure. The protobuf uses uint64 IDs to index into other lists. For example, if some household has members = [3, 10], then those two people can be found at population.people[3] and population.people[10]. Each of them will have the same household ID, pointing back to something in the population.households list.

6.1 Flows: modelling daily activites

SPC models daily travel behavior of people as "flows." Flows are broken down by by an activity – shopping/retail, attending primary or secondary school, working, or staying at home. For each activity type, a person has a list of venues where they may do that activity, weighted by a probability of going to that particular venue.

Note that flows_per_activity is stored in InfoPerMSOA, not Person. The flows for retail and school are only known at the MSOA level, not individually. So given a particular Person object, you first look up their household's MSOA — msoa = population.households[person.household].msoa and then look up flows for that MSOA — population.info_per_msoa[msoa].flows_per_activity.

Each person has exactly 1 flow for home – it's just person.household with probability 1. A person has 0 or 1 flows to work, based on the value of person.workplace.

This doesn't mean that all people in the same MSOA share the same travel behavior. Each person has their own activity_durations field, based on time-use survey data. Even if two people share the same set of places where they may go shopping, one person may spend much more time on that activity than another.

See the ASPICS conversion script for all of this in action – it has a function to collapse a person's flows down into a single weighted list.

Note that per MSOA, very few venues are represented as destinations – 10 for retail and 5 for school. Only the most likely venues from QUANT are used.

6.2 Flow weights

How do you interpret the probabilities/weights for flows? If your model needs people to visit specific places each day, you could randomly sample a venue from the flows, weighting them appropriately. For retail, you may want to repeat this sampling every day of the simulation, so they visit different venues. For primary and secondary school, it may be more appropriate to sample once and store that for the simulation – a student probably doesn't switch schools daily.

Alternatively, you can follow what ASPICS does. Every day, each person logically visits all possible venues, but their interaction there (possibly receiving or transmitting COVID) is weighted by the probability of each venue.

7 Modelling methods

The principles behind the generation of the enriched SPENSER population data and behind the modelling of trips to schools and retail from QUANT are detailed in

Spooner F et al. A dynamic microsimulation model for epidemics. Soc Sci Med., 291:114461 (2021). (DOI)

Lomax N et al. An Open-Source Model for Projecting Small Area Demographic and Land-Use Change. Geographical analysis, 54(3), 599-622 (2022). (DOI)

In order to distribute each individual of the population to a unique physical workplace, we first created a population of all individual workplaces in England, based on a combination of the Nomis UK Business Counts 2020 dataset and the Nomis Business register and Employment Survey 2015 (see Data sources). The first dataset gives the number of individual workplace counts per industry, using the SIC 2007 industry classification, with imprecise size (i.e. number of employees) bands at MSOA level. The second dataset gives the total number of jobs available at LSOA level per SIC 2007 industry category. We found that the distribution of workplace sizes follows closely a simple 1/x distribution, allowing us to draw for each workplace a size within their band, with sum constraints given by the total number of jobs available, according to the second dataset.

The workplace 'population' and individual population are then levelled for each SIC 2007 category by removing the exceeding part of whichever dataset lists more items. This takes into account that people and business companies are likely to over-report their working availability (e.g. part time and seasonal contracts are not counted differently than full time contracts, jobseekers or people on maternity leave might report the SIC of their last job). This process can be controlled by a threshold in the parameter file that defines the maximal total proportion of workers or jobs that can be removed. If the two datasets cannot be levelled accordingly, the categories are dropped and the datasets are levelled globally. Tests in the West Yorkshire area have shown than when the level 1 SIC, containing 21 unique categories, is used, 90% of the volume of commuting flows were recovered compared to the Nomis commuting OD matrices at MSOA level.

The employees for each workplace are drawn according to the 'universal law of visitation', see

Schläpfer M et al. The universal visitation law of human mobility. Nature 593, 522–527 (2021). (DOI)

This framework predicts that visitors to any destination follow a simple

$$(r,f) = K / (rf)2$$

distribution, where (r,f) is the density of visitors coming from a distance r with frequency f and K is a balancing constant depending on the specific area. In the context of commuting, it can be assumed that f=1. Additionally, we only need to weigh potential employees against each other, which removes the necessity to compute explicitly K. In the West Yorkshire test, we found a Pearson coefficient of 0.7 between the predicted flows when aggregated at MSOA level and the OD matrix at MSOA level available from Nomis.

7.1 Income data

This modelling is mainly based on the 2020 revised edition of the Earnings and hours worked, region by occupation by four-digit SOC: ASHE Table 15 database from ONS. Some percentiles for employees' gross hourly salaries are provided for each full-time and part-time job according to their four-digit SOC classification per region, and separated by sex.

7.1.1 Methods

The data are far from complete (only about 15% of all possible values), especially for the highest deciles. We found that an order 3 polynomial fit was satisfactory for most categories (93.11%) to complete the partially filled SOCs. SOCs with too many missing values are given the value for the category that is immediately higher in the SOC hierarchy. Some jobs appear to have a 'ceiling' for the highest percentiles, making the polynomial fit fail. In that case, we have replaced the unknown values by the highest known value in the raw data (as there is no clear and systemic fit for these special cases). In addition, there is no information for the highest decile in all cases, which means that the highest salaries are underestimated (and exceptionally high salaries cannot be obtained). The result of this phase is four tables {male full-time, male part-time, female full-time, female part-time} containing the coefficients of the fitted order 3 polynomial, with an optional ceiling percentile when relevant.

A percentile is chosen randomly (uniformly) for each individual, and the salary is then deduced according to their full-time/part-time status, region, sex and SOC category. A basic hourly salary column is added to the unprocessed SPC data, as well as a corresponding annual salary based on their estimated hours worked per day, according to the Time Use Survey matching. In addition, we repeat this process for all individuals that are categorised as 'Self-employed' or 'Employee unspecified' by the Time Use Survey matching, as if they were full time employees. These values are recorded in the columns IncomeHAsIF and IncomeYAsIf. We noticed that a high number of employees were given no worked hours by the Time Use Survey. We have added to the IncomeYAsIf column an estimation of their annual salary based on Table 15.9a:

Paid hours worked - Total 2020, and also depending on the same four variables as above (full-time/part-time status, region, sex and SOC category).

In addition, age data are made available by ONS. Part of the differences that can be observed between different age groups are already taken into account through the fact that the SOC category can evolve during a career. To take into account that dependence, we first run the above method without weighing by age. The results are shown in the age validation section below. The residual impact of age alone is then added to the model in the following way. When the percentile is drawn for a specific individual, it is morphed to fit within the usual percentage range accessible to that age category. The function that operates this morphing is inferred beforehand and takes into account the salary distribution per age computed by the previous non-age weighted iteration of the modelling (see figure - TBA - for a more detailed description of this function).

The R codes for this modelling are here.

The methods are validated in the next section. Since it is not possible to optimise every criterion at once, this next secton can also be used as a reference to re-adjust some values to match exactly the ONS estimated means for one particular criterion of interest.

7.1.2 Comparison to reference values from ONS

We compare the results of the modelling to the raw datasets from ONS.

- Mod for modelled
- M for male
- F for female
- H for hourly gross salary
- Y for annual gross salary
- FT for full-Time
- PT for part-Time
- Only individuals recorded as employees (i.e. not self-employed) are taken into account in this section.

Number of employees per sex and full-time/part-time classification

The numbers given by ONS vary from dataset to dataset and are reported by ONS as indicative only. For the modelled values, we give the total number of individuals with a non-zero salary in each category.

					M				
Variable	All	FT	PT	M	FT	M PT	F	F FT	F PT
ONS tot	22-26k	16-19k	6-8k	11-13k	9-11k	1.5-2k	11-13k	6.5-7.5k	4.5-5.5k

					M				
Variable	All	FT	PT	M	FT	M PT	F	F FT	F PT
Mod tot H	23.1k	18.5k	4.6k	11.8k	11k	0.8k	11.3k	7.5k	3.8k
Mod tot Y	17.6k	14.8k	2.8k	9.4k	8.9k	0.5k	8.2k	5.9k	2.3k

A significant number of individuals listed as working either full or part time have 0 effective worked hours per day according to the Time Use Survey matching. In those cases, an hourly salary is modelled depending on their SOC, region and sex, as for any other employee, but the annual salary will be displayed as 0. It is possible to estimate their likely true number of hours worked from the same ONS dataset (Table 15.9a: Paid hours worked - Total 2020), also depending on their sex, soc and region. This calculation has been added to the "As If" column.

Hourly gross salary per sex and full-time/part-time classification

Variable	All	FT	PT	M	M FT	M PT	F	F FT	F PT
ONS mean	17.63	18.32	13.93	18.81	19.12	14.69	16.19	17.08	13.68
ONS median	13.71	15.15	10.38	14.84	15.58	10.12	12.58	14.42	10.47
Mod mean	16.45	17.19	13.45	17.50	17.84	12.75	15.35	16.23	13.60
Mod median	13.55	14.46	10.23	14.27	14.72	9.16	12.79	14.12	10.51

The median values are quite close to the ONS values, but the mean values are always lower. This is expected, see the description of the modelling above.

Annual gross salary per sex and full-time/part-time classification

Only values > 0 are retained for these calculations.

Variable	All	FT	PT	M	M FT	M PT	F	F FT	F PT
ONS mean ONS	31,646 25,886	38,552 31,487	13,819 11,240	38,421 31,393	42,072 33,915	14,796 10,883	24,871 20,614	33,253 28,002	13,512 4,743
median Mod mean Mod median	34,317 28,713	36,595 30,942	22,257 17,928	37,574 31,404	38,496 32,382	20,698 17,382	30,594 25,875	33,729 29,028	22,585 18,137

The average salary for part-time employees is correct when values equal to 0 are taken into account. This suggests that the total number of hours worked for part-time employees is

correct, but the way they are distributed among individuals is not. It could be due to the TUS taking a snapshot of the situation during a particular week, rather than averaging their data over the year. It appears that the TUS matching also overestimates the average number of hours worked for female employees.

Regional differences (hourly gross salary)

Region	East Mid- East lands	North Londo L ast	North West	South East	South West	West Mid- lands	Yorkshire and The Humber
ONS	16.74 15.87	23.78 15.69	16.36	17.88	16.36	16.34	15.76
mean ONS me-	13.28 12.65	18.30 12.40	12.90	14.33	12.74	12.92	12.46
dian Mod	16.67 15.29	19.39 15.05	15.22	17.34	15.92	15.47	14.41
mean Mod me- dian	13.69 12.79	16.25 12.42	12.44	14.84	13.35	12.64	12.44

The pearson correlations for mean and median between the modelled and raw values are 0.92 and and 0.93.

Hourly gross salary per one-digit SOC

1d SOC	1	2	3	4	5	6	7	8	9
ONS mean	26.77	23.38	18.29	13.42	13.35	10.87	10.94	12.23	10.77
ONS median	20.96	21.34	15.66	11.54	12.04	10.08	9.52	10.93	9.22
Mod mean	21.52	22.14	16.00	12.76	12.55	10.49	10.50	12.05	9.87
Mod median	17.22	20.66	14.12	11.46	11.34	9.71	9.59	10.82	9.12

- 1. Managers, directors and senior officials
- 2. Professional occupations
- 3. Associate professional and technical occupations
- 4. Administrative and secretarial occupations
- 5. Skilled trades occupations
- 6. Caring, leisure and other service occupations
- 7. Sales and customer service occupations
- 8. Process, plant and machine operatives
- 9. Elementary occupations.

The pearson correlations for mean and median between the modelled and raw values are 0.98 and 0.98.

Hourly gross salary per age

The reference for this table is: Table 6.5a Hourly pay - Gross 2020

Table before weighting by age:

Age	16-17	18-21	22-29	30-39	40-49	50-59	60+
ONS mean	7.21	9.59	14.09	18.13	20.04	19.12	16.32
ONS median	6.36	9.00	12.26	15.08	15.89	14.39	12.17
Mod mean	12.77	14.96	16.33	16.93	16.83	16.66	16.29
Mod median	10.93	12.71	13.88	14.02	13.96	13.85	13.65

The pearson correlations for mean and median between the modelled and raw values are 0.92 and 0.92.

Table after weighting by age:

Age	16-17	18-21	22-29	30-39	40-49	50-59	60+
ONS mean	7.21	9.59	14.09	18.13	20.04	19.12	16.32
ONS median	6.36	9.00	12.26	15.08	15.89	14.39	12.17
Mod mean	9.05	11.15	14.87	17.35	17.96	17.47	15.41
Mod median	8.20	9.51	12.86	14.41	14.78	14.43	12.56

The pearson correlations for mean and median between the modelled and raw values are 0.99 and 0.99.

7.2 BMI data

Body Max Index (BMI) is calculated for each individual from the Health Survey for England 2019 (access needs to be requested to the UK Data Service). This calculation is completely indepedent from the PSM to the HSE 2017, and therefore the new BMI values will not fit within the categories indicated by this earlier PSM. As the BMI variable is not necessarily independent from the other health variables (diabetes etc.), the new variable should only be used for studies where all other variables are considered equal. The new variable is continuous (a float).

According to the HSE 2019, the distribution of BMI values should follow figure 1. Socio-economic category was discarded for the modelling as it is not independent from the other variables and "mixed" and "other" ethnicities have been merged due to small sample sizes.

Figure 1. BMI per age. Columns represent ethnicity (White, Black, Asian, Other), and the rows sex (female, male).

The distribution for each age group is a gamma distribution. See figure 2.

Figure 2. Distribution of BMI values for white females aged 30-34.

Due to small sample sizes, the BMI is calculated for each individual depending on their age according to a gamma distribution whose mean is the mean for the corrresponding age, sex and ethnicity (thick line in figure 1), but whose variance is only determined by the total variance by sex and ethnicity. The resulting BMI where validated for Bedfordshire, and correlations of 0.93 and 0.97 were found between the mean and variance of the modelled data compared to those for the reference HSE 2019 data. See figure 3. The distribution per age, as in figure 1, were also validated.

Figure 3. Modelled mean and variance compared to the reference mean and variance from the HSE 2019 data for each of the eight categories of figure 1.

The R codes for this modelling are here.

8 Data sources

The data is sorted around the 2011 Middle-layer Super Output Area (MSOA) geographical unit. These units where created for census collection and are designed to be relatively homogeneous, with an average population size of 8000. Any list of MSOAs in England can be run, with the exception of the MSOAs forming the City of London (i.e the London borough called the City, not London as a whole).

The data from Open Street Map (OSM) is downloaded directly from https://www.openstreetmap.org. Everything else is hosted as local copies on one Azure repository that interacts automatically with the model, and divided into utilities, county level data and national data.

lookUp.csv

The look-up table links different geographies together. It is used internally by the model, but can also help the user define their own study area. MSOA11CD, MSOA11NM, LAD20CD, LAD20NM, ITL321CD, ITL321NM, ITL221CD, ITL121NM, ITL121CD, ITL121NM are all standard denominations fully compatible with ONS fields of the same name. They are based on ONS lookups. See ONS documentation for more details. CTY20NM and CCTY20NM are custom denominations for the counties of England (used to sort the county level population data) and the ceremonial counties of England respectively. Their spelling may vary in different data sources and the field CTY20NM is not compatible with the ONS field of the same name (which excludes all counties that are also unitary authorities). GoogleMob and OSM are different spellings for the counties of England used by Google and OSM for their data releases.

8.1 County level data

Contains 47 files, each representing the population in 2020 of one of the counties of England mentioned above, and named

pop_<county_name>.gz

This data is based on the 2011 UK census, the Time Use Survey 2014-15 and the Health Survey for England 2017. The SPENSER (Synthetic Population Estimation and Scenario Projection) microsimulation model (reference) distributes a synthetic population based on the census at MSOA scale and projects it to 2020 according to estimates from the Office for National Statistics (ONS). This information was enriched with some of the content of the other two datasets through propensity score matching (PSM) by Prof. Karyn Morrissey (Technical University of Denmark). The rest of the datasets can be added a posteriori from the identifiers provided.

The fields currently contained are:

- idp: a unique global individual identifier across all counties
- MSOA11CD: MSOA code where the individual lives
- hid: household identifier, includes communal establishments
- pid: identifier linking to the 2011 Census
- pid_tus: identifier linking to the Time Use Survey 2015
- pid_hse: identifier linking to the Health Survey for England 2017
- sex: 0 female; 1 male
- age: in years
- origin: 1 White; 2 Black; 3 Asian; 4 Mixed; 5 Other
- nssec5: National Statistics Socio-economic classification:
 - 1: Higher managerial, administrative and professional occupations
 - 2: Intermediate occupations
 - 3: Small employers and own account workers
 - 4: Lower supervisory and technical occupations
 - 5: Semi-routine and routine occupations
 - 0: Never worked and long-term unemployed
- soc2010: Previous version of the Standard Occupational Classification
- sic1d07: Standard Industrial Classification of Economic Activities 2007, 1st layer (number corresponding to the letter in alphabetical order)
- sic2d07: Standard Industrial Classification of Economic Activities 2007, 2nd layer
- pwkstat: Employment status according to the TUS
- Proportion of 24h spent doing different daily activities:
 - punknown + pwork + pschool + pshop + pservices + pleisure + pescort +
 ptransport = pnothome
 - phome + pworkhome = phometot
 - pnothome + phometot = 1
- IncomeX: hourly (X = "H") and annual (X = "Y") income for employees, see modelling methods for more details
- cvd: has a cardio-vascular disease (0 or 1)
- diabetes: has diabetes (0 or 1)

• bloodpressure: has high blood pressure (0 or 1)

• BMIvg6: Body Mass Index:

Not applicable

Underweight: less than 18.5
Normal: 18.5 to less than 25
Overweight: 25 to less than 30
Obese I: 30 to less than 35
Obese II: 35 to less than 40
Obese III: 40 or more

• bmiNew is a float estimated value for the BMI of the individual. These were drawn directly according to age, sex and ethnicity and are completely independent of the above values (see modelling methods).

lng: longitude of the MSOA11CD centroidlat: latitude of the MSOA11CD centroid

Some other fields were kept from and for other specific projects but are not from official sources and should generally not be used.

8.2 National data

businessRegistry.csv

Contains a breakdown of all business units (i.e. a single workplace) in England at LSOA scale (smaller than MSOA), estimated by the project contributors from two nomis datasets: UK Business Counts - local units by industry and employment size band 2020 and Business Register and Employment Survey 2015. Each item contains the size of the unit and its main sic1d07 code in reference to standard Industrial Classification of Economic Activities 2007 (number corresponding to the letter in alphabetical order). It is used to compute commuting flows.

The R codes to compute this file are here.

MSOAS_shp.tar.gz

Is a simple shapefile taken from ONS boundaries.

QUANT_RAMP.tar.gz

See: Milton R, Batty M, Dennett A, dedicated RAMP Spatial Interaction Model GitHub repository. It is used to compute the flows towards schools and retail.

timeAtHomeIncreaseCTY.csv

This file is a subset from Google COVID-19 Community Mobility Reports, cropped to England. It describes the daily reduction in mobility, averaged at county level, due to lockdown and other COVID-19 restrictions between the 15th of February 2020 and 15th of April 2022. Missing values have been replaced by the national average. These values can be used directly to reduce pnothome and increase phometot (and their sub-categories) to simulate more accurately the period.

The R codes to process these data are here.

Part III Advanced

9 Developer guide

The site is built with Quarto. You can iterate on it locally: cd docs; quarto preview

9.1 Code hygiene

We use automated tools to format the code.

```
cargo fmt

# Format Markdown docs
prettier --write *.md
prettier --write docs/*.qmd --parser markdown
```

Install prettier for Markdown.

9.2 Some tips for working with Rust

There are two equivalent ways to rebuild and then run the code. First:

```
cargo run --release -- devon
```

The -- separates arguments to cargo, the Rust build tool, and arguments to the program itself. The second way:

```
cargo build --release
./target/release/aspics devon
```

You can build the code in two ways – **debug** and **release**. There's a simple tradeoff – debug mode is fast to build, but slow to run. Release mode is slow to build, but fast to run. For the ASPICS codebase, since the input data is so large and the codebase so small, I'd recommend always using --release. If you want to use debug mode, just omit the flag.

If you're working on the Rust code outside of an IDE like VSCode, then you can check if the code compiles much faster by doing cargo check.

9.3 Docker

We provide a Dockerfile in case it's helpful for running, but don't recommend using it. If you want to, then assuming you have Docker setup:

```
docker build -t spc .
docker run --mount type=bind,source="$(pwd)"/data,target=/spc/data -t spc /spc/target/rele
```

This will make the data directory in your directory available to the Docker image, where it'll download the large input files and produce the final output.

10 Code walkthrough

SPC is implemented in Rust, and its code can be found here. This is an unusual implementation choice in the data science world, so this page has some notes about it.

The code-base makes use of some techniques that may be generally applicable to other projects, independent of the language chosen.

10.0.1 Split code into two stages

Agent-based models and spatial interaction models require some kind of input. Often the effort to transform external data into this input can exceed that of the simulation component. Cleanly separating the two problems has some advantages:

- iterate on the simulation faster, without processing raw data every run
- reuse the prepared input for future projects
- force thinking about the data model needed by the simulation, and transform the external data into that form

SPC is exactly this first stage, originally split from ASPICS when further uses of the same population data were identified.

10.0.2 Explicit data schema

Dynamically typed languages like Python don't force you to explicitly list the shape of input data. It's common to read CSV files with pandas, filter and transform the data, and use that throughout the program. This can be quick to start prototyping, but is hard to maintain longer-term. Investing in the process of writing down types:

- makes it easier for somebody new to understand your system they can first focus on what you're modeling, instead of how that's built up from raw data sources
- clarifies what data actually matters to your system; you don't carry forward unnecessary input
- makes it impossible to express invalid states

- One example is here per person and activity, there's a list of venues the person may visit, along with a probability of going there. If the list of venues and list of probabilities are stored as separate lists or columns, then their length may not match.
- reuse the prepared input for future projects

There's a variety of techniques for expressing strongly typed data:

- protocol buffers or flatbuffers
- JSON schemas
- Python data classes and optional type hints
- statically typed languages like Rust

10.0.3 Type-safe IDs

Say your data model has many different objects, each with their own ID – people, households, venues, etc. You might store these in a list and use the index as an ID. This is fine, but nothing stops you from confusing IDs and accidentally passing in venue 5 to a function instead of household 5. In Rust, it's easy to create "wrapper types" like this and let the compiler prevent these mistakes.

This technique is also useful when preparing external data. GTFS data describing public transit routes and timetables contains many string IDs – shapes, trips, stops, routes. As soon as you read the raw input, you can store the strings in more precise types that prevent mixing up a stop ID and route ID.

10.0.4 Idempotent data preparation

If you're iterating on your initialisation pipeline's code, you probably don't want to download a 2GB external file every single run. A common approach is to first test if a file exists and don't download it again if so. In practice, you may also need to handle unzipping files, showing a progress bar while downloading, and printing clear error messages. This codebase has some common code for doing this in Rust. We intend to publish a separate library to more easily call in your own code.

10.0.5 Logging with structure

It's typical to print information as a complex pipeline runs, for the user to track progress and debug problems. But without any sort of organization, it's hard to follow what steps take a long time or encounter problems. What if your logs could show the logical structure of your pipeline and help you understand where time is spent?

```
Found 474,207 buildings from data/raw data/countydata/OSM/west-yorkshire-latest-free/gi
                     [get_info_per_msoa]
192.645] [get_Info_per_msoa] Found 474,207 buildings from data, ran_asta, sampless of the per_msoa] Matching after 1.8hp
192.70s] [get_info_per_msoa] Matching 474,207 points to 299 polygons. Building R-Tree...
194.22s] [calculate lockdown_per_day] Calculating per-day lockdown values
194.22s] [load events] Loading events data
194.25s] [initialisation] By the end, Memory usage: 1.53GiB
200.89s] [Writing snapshot] Merging flows for all activities
                                            initialisation WestYorkshireLarge
                                               grab_raw_data
creating population
read_individual_time_use_and_health_data
Reading "data/raw_data/countydata/tus_hse_west-yorkshire.csv'
            8.20s
                                                          Creating households
                                                     create_commuting_flows
setup_venue_flows Retail
             152.38s
                                                    setup_venue_flows Retail
Copying flows to people Retail
setup_venue_flows Nightclub
Copying flows to people Nightclub
setup_venue_flows PrimarySchool
Copying flows to people PrimarySchool
setup_venue_flows SecondarySchool
Copying flows to people SecondarySchool
                 6.58s
            7.00s
                6.50s
       2.03s
                                                 get_info_per_msoa
calculate_lockdown_per_day
        24.48ms
                                                load_events "model_parameters/eventDataConcerts.csv"
Writing population to "data/processed_data/WestYorkshireLarge/rust_cache.bin"
Writing snapshot
         251.20μs
       16.93s
```

The screenshot above shows a summary printed at the end of a long pipeline run. It's immediately obvious that the slowest step is creating commuting flows.

This codebase uses the tracing framework for logging, with a custom piece to draw the tree. (We'll publish this as a separate library once it's more polished.) The tracing framework is hard to understand, but the main conceptual leap over regular logging framworks is the concept of a **span**. When your code starts one logical step, you call a method to create a new span, and when it finishes, you close that span. Spans can be nested in any way – create_commuting_flows happens within the larger step of creating population.

10.0.6 Determinism

Given the same inputs, your code should always produce identical output, no matter where it's run or how many times. Otherwise, debugging problems becomes very tedious, and it's more difficult to make conclusions from results. Of course, many projects have a stochastic element – but this should be controlled by a random number generator (RNG) seed, which is part of the input. You vary the seed and repeat the program, then reason about the distribution of results.

Aside from organizing your code to let a single RNG seed influence everything, another possible source of non-determinism is iteration order. In Rust, a HashMap could have different order every time it's used, so we use a BTreeMap instead when this matters. In Python, dictionaries are ordered. Be sure to check for your language.

10.1 Protocol buffers

SPC uses protocol buffers v2 for output. This has some advantages explained the "explicit data schema" section above.

Note that we chose proto2 instead of proto3, because proto3 doesn't support required fields. This is done to allow schemas to evolve better over time, but this isn't a feature SPC makes use of. There's no need to have new code work with old data, or vice versa – if the schema is updated, downstream code should adapt accordingly and use the updated input files.

Note also that protocol buffers don't easily support type-safe wrappers around numeric IDs, so downstream code has to be careful not to mix up household, venue, and person IDs. For this reason, SPC internally doesn't use the auto-generated protobuf code until the very end of the pipeline. It's always possible to be more precise with native Rust types, and convert to the less strict types later.

10.2 An example of the power of static type checking

Imagine we want to add a new activity type to represent people going to university and higher education. SPC already has activities for primary and secondary school, so we'll probably want to follow those as a guide. In any language, we could search the codebase for relevant terms to get a sense of what to update. In languages like Python without an up-front compilation step, if we fail to update something or write blatantly incorrect code (such as making a typo in variable names or passing a list where a string was expected), we only find out when that code happens to run. In pipelines with many steps and large input files, it could be a while before we reach the problematic code.

Let's walk through the same exercise for SPC's Rust code. We start by adding a new University case to the Activity enum. If we try to compile the code here (with cargo check or an IDE), we immediately get 4 errors.

Three of the errors are in the QUANT module. The first is here. It's immediately clear that for retail and primary/secondary school, we read in two files from QUANT representing venues where these activities take place and the probability of going to each venue. Even if we were unfamiliar with this codebase, the compiler has told us one thing we'll need to figure out, and where to wire it up.

The other error is in the code that writes the protobul output. Similarly, we need a way to represent university activities in the protobul scheme.

Extending an unfamiliar code-base backed by compiler errors is a very guided experience. If you wanted to add more demographic attributes to people or energy use information to households, you don't need to guess all of the places in the code you'll need to update. You can just add the field, then let the compiler tell you all places where those objects get created.

11 Performance

The following tables summarizes the resources SPC needs to run in different areas.

year study_area	$\mathrm{num}_{_}$	_maoaa_ho	us ehooh dspe pþ l	efile_sincentim	e commuting	_mention@	_usage
2012 England/bedfordshire	74	245,166	647,272 256	6.83 10	3 seconds	849.02	
			Mil	B sec-		MiB	
				onds			
$2020\mathrm{England/bedfordshire}$	74	$272,\!875$	674,044 271	.65 9 sec-	3 seconds	922.88	
			Mil			MiB	
$2022 \mathrm{England/bedfordshire}$	74	309,706	703,582 277	7.74 9 sec-	3 seconds	929.80	
			Mil	B onds		MiB	
$2032 \mathrm{England/bedfordshire}$	74	309,706	703,582 277	7.74 9 sec-	3 seconds	929.80	
			Mil	B onds		MiB	
2039 England/bedfordshire	74	$329,\!061$	715,797 278	3.39 11	3 seconds	927.77	
			Mil	B sec-		MiB	
				onds			
2012 England/berkshire	107	$342,\!167$	890,543 356	6.04 14	7 seconds	1.06	
			Mil	B sec-		GiB	
				onds			
2020 England/berkshire	107	$365,\!905$	918,258 373	3.35 14	7 seconds	1.10	
			Mil	B sec-		GiB	
				onds			
2022 England/berkshire	107	$394,\!446$	941,655 368	3.37 14	7 seconds	1.08	
			Mil	B sec-		GiB	
				onds			
2032 England/berkshire	107	$394,\!446$	941,655 368	3.37 14	7 seconds	1.08	
			Mil	B sec-		GiB	
				onds			
2039 England/berkshire	107	408,604	949,986 367	7.21 14	7 seconds	1.08	
			Mil	B sec-		GiB	
				onds			
2012 England/bristol	55	$182,\!299$	448,233 173	8.74 6 sec-	2 seconds	527.23	
			Mil	B onds		MiB	
2020 England/bristol	55	196,940	470,039 183	8.99 6 sec-	2 seconds	547.49	
			Mil	B onds		MiB	

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2022 England/bristol	55	216,197	503,014 192.51	7 sec-	2 seconds	559.78
,			${ m MiB}$	onds		MiB
2032 England/bristol	55	216,197	503,014 192.51	7 sec-	2 seconds	559.78
,			$_{ m MiB}$	onds		MiB
039 England/bristol	55	227,770	521,371 199.72	7 sec-	2 seconds	573.40
9 ,		,	MiB	onds		MiB
012 England/buckinghai	msh 99	99,235	261,340 108.30	5 sec-	1 second	310.38
0 / 0		,	MiB	onds		MiB
020 England/buckinghai	msh99	108,999	271,050 114.31	5 sec-	1 second	400.87
0 / 0		,	MiB	onds		MiB
2022 England/buckinghai	msh99	123,578	278,548 112.39	5 sec-	1 second	393.64
3 3 3		- ,	MiB	onds		MiB
2032 England/buckinghai	msh99	123,578	278,548 112.39	5 sec-	1 second	393.64
000		,	MiB	onds		MiB
039 England/buckinghai	msh 99	130,393	281,773 112.14	5 sec-	1 second	391.52
			MiB	onds	_ 55555	MiB
2012 England/cambridges	shir@8	327,257	832,980 323.35	11	5 seconds	1013.16
101 2 2118101114/ 00111511480	01111 @ 0	321,231	MiB	sec-	0 20001142	MiB
			1,112	onds		1,112
020 England/cambridges	shir@8	348,522	863,250 341.16	12	5 seconds	1.03
02021161011147 00111511460	01111 @ 0	313,322	MiB	sec-	0 20001142	GiB
			11112	onds		GIB
022 England/cambridges	shir@8	377,634	907,166 348.75	12	5 seconds	1.03
022 121151ana / cambi rage.		011,001	MiB	sec-	o seconas	GiB
			WIID	onds		GID
032 England/cambridges	shir@8	377,634	907,166 348.75	12	5 seconds	1.03
002 England/ campinger		011,001	MiB	sec-	o seconds	GiB
			WIID	onds		GID
2039 England/cambridges	shir@8	392,478	924,170 351.39	12	5 seconds	1.04
1000 England, campinger		002,410	MiB	sec-	o seconds	GiB
			WIID	onds		GID
2012 England/cheshire	139	441,084	1,042,06 3 02.14	16	7 seconds	1.13
1012 Eligiana/ chesime	100	441,004	1,042,00 4 02.14 MiB	sec-	1 seconds	GiB
			WIID	onds		GID
020 England/cheshire	139	464,134	1,070,597116.35	16	7 seconds	1.46
020 England/Cheshire	103	404,104	1,070,59#10.55 MiB	sec-	i seconds	GiB
			MIID	onds		OID
022 England/cheshire	139	489,476	1,125,19\\ 25.27	onus 16	7 seconds	1.47
044 Eligianu/chesime	199	403,410	1,125,19 4 25.27 MiB		, seconds	GiB
			MIID	sec-		GID
				onds		

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032 England/cheshire	139	489,476	1,125,19825.27	16	7 seconds	1.47
- ,			MiB	sec-		GiB
				onds		
2039 England/cheshire	139	$501,\!501$	1,149,51 31.10	16	7 seconds	1.48
			${ m MiB}$	sec-		GiB
				onds		
012 England/cornwall	74	$232,\!659$	$549,\!616\ 208.07$	8 sec-	2 seconds	742.92
			${ m MiB}$	onds		MiB
020 England/cornwall	74	$247,\!105$	577,414 219.74	8 sec-	3 seconds	764.95
			${ m MiB}$	onds		MiB
$022\mathrm{England/cornwall}$	74	$270,\!134$	634,940 233.36	8 sec-	3 seconds	828.45
			MiB	onds		MiB
$032 \mathrm{England/cornwall}$	74	$270,\!134$	$634,940\ 233.36$	8 sec-	3 seconds	828.45
			${ m MiB}$	onds		MiB
039 England/cornwall	74	$280,\!546$	$658,\!610\ 239.72$	8 sec-	3 seconds	838.14
			MiB	onds		MiB
012 England/cumbria	64	$222,\!586$	498,624 188.03	7 sec-	2 seconds	547.33
			${ m MiB}$	onds		MiB
020 England/cumbria	64	$226,\!893$	499,873 188.73	7 sec-	2 seconds	548.52
			MiB	onds		MiB
022 England/cumbria	64	$230,\!206$	499,840 183.18	7 sec-	2 seconds	533.99
			MiB	onds		MiB
032 England/cumbria	64	$230,\!206$	499,840 183.18	7 sec-	2 seconds	533.99
			${ m MiB}$	onds		MiB
039 England/cumbria	64	$231,\!202$	498,475 181.58	7 sec-	2 seconds	530.96
			MiB	onds		MiB
012 England/derbyshire	131	$436,\!276$	1,035,35 6 97.75	15	7 seconds	1.12
			MiB	sec-		GiB
				onds		
020 England/derbyshire	131	459,743	1,064,40@109.76	16	8 seconds	1.44
			MiB	sec-		GiB
				onds		
2022 England/derbyshire	131	489,764	1,122,07819.52	16	7 seconds	1.45
			MiB	sec-		GiB
				onds		
032 England/derbyshire	131	489,764	1,122,07819.52	16	7 seconds	1.45
			MiB	sec-		GiB
				onds		
039 England/derbyshire	131	$505,\!314$	$1,\!152,\!518429.01$	16	8 seconds	1.47
			MiB	sec-		GiB
				onds		

156 156	494,106 523,033	1,165,95 2 38.60 MiB 1,212,38 7 59.44	18 sec- onds	8 seconds	1.49 GiB
	523,033				GiB
	523,033	1 212 38 7 59 44	onds		0.12
	523,033	1 212 387159 44			
		1,212,00 100.11	18	8 seconds	1.53
		MiB	sec-		GiB
			onds		
156	$567,\!011$	1,304,87478.71	19	9 seconds	1.64
		MiB	sec-		GiB
156	567,011	, ,		9 seconds	1.64
		MiB			GiB
4 F A	X 00.1 T 0	1 0 10 ==== 00 00		0 1	1.00
156	589,178	, ,		9 seconds	1.66
		MiB			GiB
4 4 =	200 452	011 001 040 50		. 1	1.00
117	390,472	<i>'</i>		5 seconds	1.03
		MıB			GiB
115	407.000	000 104 050 50		F 1	1.05
117	407,828			5 seconds	1.05
		MiB			GiB
117	405 611	050 001 950 09		۳ 1	1.04
117	425,611	,		5 seconds	1.04 C:D
		MIB			GiB
117	405 611	050 001 256 62		۲ ا	1.04
117	425,011	,		o seconds	1.04 C:P
		MIID			GiB
117	434 502	050 555 257 66		5 seconds	1.04
111	404,030			o seconds	GiB
		MID			מוט
102	355 257	897 703 313 71		5 seconds	987.31
102	555,251	,		o seconds	967.31 MiB
		MIID			MIID
102	380 804	853 970 394 09		6 seconds	1006.13
102	300,034			o acconda	MiB
		MIID			141117
102	423 181	895 907 329 56		5 seconds	1008.58
102	720,101	,		o acconda	MiB
		MIID			141117
	156 156 117 117 117 117 117 102 102	156 589,178 117 390,472 117 407,828 117 425,611 117 434,593 102 355,257 102 380,894	MiB 156 589,178 1,342,77\struct 88.23 MiB 117 390,472 911,601 349.78 MiB 117 407,828 930,184 359.59 MiB 117 425,611 952,801 356.63 MiB 117 434,593 959,555 357.66 MiB 102 355,257 827,703 313.71 MiB 102 380,894 853,970 324.02 MiB	MiB seconds 156 589,178 1,342,77 ♣88.23 19 MiB seconds 117 390,472 911,601 349.78 12 MiB seconds 117 407,828 930,184 359.59 12 MiB seconds 117 425,611 952,801 356.63 12 MiB seconds 117 425,611 952,801 356.63 12 MiB seconds 117 434,593 959,555 357.66 12 MiB seconds 102 355,257 827,703 313.71 12 MiB seconds 102 380,894 853,970 324.02 12 MiB seconds 102 423,181 895,907 329.56 12	156 567,011 1,304,87478.71 19 9 seconds MiB seconds 156 589,178 1,342,77488.23 19 9 seconds MiB seconds 117 390,472 911,601 349.78 12 5 seconds MiB seconds 117 407,828 930,184 359.59 12 5 seconds MiB seconds 117 425,611 952,801 356.63 12 5 seconds MiB seconds 117 434,593 959,555 357.66 12 5 seconds 117 434,593 959,555 357.66 12 5 seconds 102 355,257 827,703 313.71 12 5 seconds 102 380,894 853,970 324.02 12 6 seconds 102 423,181 895,907 329.56 12 5 seconds 12 5 seconds 102 423,181 895,907 329.56 12 5 seconds 12 5 seconds 12 5 seconds 138 148

year study_area	num_	_msows_ho	us eluoh dspe pþ lefile	_sizentim	e commuting	_mentiony_
2032 England/east-sussex	102	423,181	895,907 329.56	12	5 seconds	1008.58
			MiB	sec-		MiB
				onds		
2039 England/east-sussex	102	446,000	$915,\!014\ 335.45$	12	5 seconds	1020.75
			${ m MiB}$	sec-		MiB
				onds		
$2012 \mathrm{England/east}$	75	$255,\!848$	$593,\!271\ 227.49$	8 sec-	3 seconds	778.75
yorkshire-with-hull			${ m MiB}$	onds		MiB
$020 \mathrm{England/east}$	75	262,609	$602,\!286\ 233.14$	8 sec-	3 seconds	835.04
yorkshire-with-hull			${ m MiB}$	onds		MiB
$2022 \mathrm{England/east}$	75	$272,\!805$	$613,721\ 230.34$	8 sec-	3 seconds	824.50
yorkshire-with-hull			${ m MiB}$	onds		MiB
032 England/east-	75	272,805	$613,721\ 230.34$	8 sec-	3 seconds	824.50
yorkshire-with-hull			${ m MiB}$	onds		MiB
2039 England/east-	75	277,770	617,357 230.45	8 sec-	3 seconds	825.00
yorkshire-with-hull			${ m MiB}$	onds		MiB
2012 England/essex	211	722,974	1,786,316690.77	30	19	2.06
J ,		ŕ	MiB	sec-	seconds	GiB
				onds		
2020 England/essex	211	773,454	1,857,20526.02	32	20	2.13
J ,		ŕ	MiB	sec-	seconds	GiB
				onds		
022 England/essex	211	858,552	1,981,994761.40	33	20	2.19
<u> </u>		ŕ	MiB	sec-	seconds	GiB
				onds		
032 England/essex	211	858,552	1,981,994761.40	33	20	2.19
9 ,		,	MiB	sec-	seconds	GiB
				onds		
039 England/essex	211	906,640	2,042,404777.71	33	21	2.21
3 /		,	MiB	sec-	seconds	GiB
				onds		
2012 England/gloucestershi	re107	365,240	889,836 344.16	13	5 seconds	1.02
3, 8		,	MiB	sec-		GiB
				onds		S-1
2020 England/gloucestershi	re107	392,643	933,909 362.90	13	6 seconds	1.06
		352,013	MiB	sec-	Jeconas	GiB
			1,1117	onds		3.12
022 England/gloucestershi	re107	432,216	1,025,07 3 89.56	14	6 seconds	1.10
o== = 115 mind/ Stodecoversini	10101	102,210	1,020,07 b03.00 MiB	sec-	5 becomes	GiB
			MIID			מוט
				onds		

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032 England/gloucestersh	nire107	432,216	1,025,07389.56	14 6 seconds	1.10
•			${ m MiB}$	sec-	GiB
				onds	
039 England/gloucestersh	nire107	$453,\!383$	1,068,48403.87	14 6 seconds	1.43
			${ m MiB}$	sec-	GiB
				onds	
012 England/greater-	983	3,283,305	8,581,24 3 .27	12 11	11.80
london			GiB	min- minutes utes	GiB
020 England/greater-	983	3,574,266	8,983,773.48	12 11	12.22
london		, ,	$_{ m GiB}$	min- minutes	GiB
				utes	
$022 \mathrm{England/greater}$	983	3,997,548	$9,\!452,\!04$ 3.55	12 11	12.25
london			GiB	min- minutes	GiB
000E 1 1/	0.00	2 00 10	0.450.040.55	utes	10.05
2032 England/greater-	983	3,997,548	9,452,04 3 .55	12 11	12.25
london			GiB	min- minutes utes	GiB
039 England/greater-	983	4,229,017	9,688,50 6 .58	12 11	12.95
london			GiB	min- minutes utes	GiB
012 England/greater-	346	1,128,371	2,745,451.05	70 53	3.56
manchester		, ,	GiB	sec- seconds	GiB
000E 1 1/	0.40	1 100 5 15	2 2 4 2 4 2 4 4 2	onds	2.00
020 England/greater-	346	1,192,547	2,840,43 1 .10	71 54	3.66
manchester			GiB	sec- seconds	GiB
022 England/greater-	346	1 272 690	2,974,954.13	$\begin{array}{cc} \text{onds} \\ 73 & 55 \end{array}$	3.69
manchester	940	1,414,009	2,974,954.15 GiB	sec- seconds	5.09 GiB
manchester			GID	onds	GID
032 England/greater-	346	1 272 689	2,974,954.13	74 56	3.69
manchester	010	1,212,000	GiB	sec- seconds	$_{ m GiB}$
11101101100001			QID.	onds	OID.
039 England/greater-	346	1,319,090	3,049,727.15	76 58	3.73
manchester		, ,	GiB	sec- seconds	GiB
				onds	
012 England/hampshire	225	733,611	1,810,51 % 98.03	32 20	2.07
- , *		,	MiB	sec- seconds	GiB
				onds	

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020 England/hampshire	225	777,116	1,861,25 7 21.62	32	20	2.12
- , -			MiB	sec- onds	seconds	GiB
022 England/hampshire	225	836,451	1,931,669728.97	32	20	2.12
, <u>,</u>		,	MiB	sec- onds	seconds	GiB
032 England/hampshire	225	836,451	1,931,669728.97	33	20	2.12
5 , 1		,	MiB	sec- onds	seconds	GiB
039 England/hampshire	225	867,417	1,960,190735.50	33	20	2.13
, <u>, , , , , , , , , , , , , , , , , , </u>			MiB	sec- onds	seconds	GiB
012 England/herefordshire	23	79,083	188,362 72.21	4 sec-	1 second	234.89
,		,	MiB	onds		MiB
020 England/herefordshire	23	83,238	195,194 74.71	4 sec-	1 second	239.36
·			MiB	onds		MiB
022 England/herefordshire	23	89,574	209,784 77.63	4 sec-	1 second	242.83
·			${ m MiB}$	onds		MiB
32 England/herefordshire	23	89,574	209,784 77.63	4 sec-	1 second	242.83
			${ m MiB}$	onds		MiB
39 England/herefordshire	23	$92,\!605$	216,508 79.43	4 sec-	1 second	245.69
			${ m MiB}$	onds		MiB
12 England/hertfordshire	153	$457,\!276$	$1{,}160{,}15{,}458.65$	19	11	1.56
			MiB	sec- onds	seconds	GiB
020 England/hertfordshire	153	$494,\!661$	1,190,043477.18	19	11	1.59
			MiB	sec- onds	seconds	GiB
022 England/hertfordshire	153	$546,\!573$	$1,\!219,\!12476.55$	19	11	1.67
			MiB	$\frac{\text{sec-}}{\text{onds}}$	seconds	GiB
032 England/hertfordshire	153	$546,\!573$	$1,\!219,\!124\!\!\!\!476.55$	19	11	1.67
			MiB	sec- onds	seconds	GiB
039 England/hertfordshire	153	$575,\!179$	$1,\!233,\!572476.97$	19	11	1.67
			MiB	sec- onds	seconds	GiB
012 England/isle-of-	18	61,636	139,732 53.88	3 sec-	1 second	188.79
wight		•	MiB	onds		MiB
020 England/isle-of-	18	65,140	143,268 54.99	3 sec-	1 second	190.45
wight		,	MiB	onds		MiB

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2022 England/isle-of-	18	70,496	151,582 55.55	3 sec-	1 second	200.99
wight			MiB	onds		MiB
032 England/isle-of-	18	$70,\!496$	$151,582\ 55.55$	3 sec-	1 second	200.99
wight			MiB	onds		MiB
039 England/isle-of-	18	72,968	$154,\!841\ 56.14$	3 sec-	1 second	202.13
wight			MiB	onds		MiB
$012\mathrm{England/kent}$	220	$718,\!544$	1,793,70 2 700.10	29	17	2.08
			MiB	$\frac{\text{sec-}}{\text{onds}}$	seconds	GiB
$020\mathrm{England/kent}$	220	781,933	$1,\!873,\!45$ 7 37.20	30	18	2.15
			MiB	sec- onds	seconds	GiB
2022 England/kent	220	875,515	2,008,857773.24	31	19	2.21
- · ·			MiB	sec- onds	seconds	GiB
$032\mathrm{England/kent}$	220	$875,\!515$	2,008,85773.24	32	19	2.21
			MiB	sec- onds	seconds	GiB
039 England/kent	220	$926,\!571$	2,069,08788.47	35	19	2.23
,			MiB	sec- onds	seconds	GiB
012 England/lancashire	191	619,861	1,476,46\$71.94	24	14	1.83
			MiB	sec- onds	seconds	GiB
020 England/lancashire	191	640,196	1,511,89 6 89.78	24	14	1.87
- ,			MiB	sec- onds	seconds	GiB
022 England/lancashire	191	663,637	1,567,39594.49	24	14	1.87
- ,			MiB	sec- onds	seconds	GiB
032 England/lancashire	191	663,637	1,567,39594.49	24	14	1.87
,		,	MiB	sec- onds	seconds	GiB
039 England/lancashire	191	674,387	1,591,90 % 00.02	25	14	1.88
J ,		,	MiB	sec- onds	seconds	GiB
012 England/leicestershire	120	$370,\!305$	$958,\!470\ 373.02$	14	7 seconds	1.08
			MiB	sec- onds		GiB

year study_area	num	_maoaa_ho	us ehool dspe pb<u>le</u>file_	_sizentim	e commuting_	_mentiony_
2020 England/leicestershire	120	397,467	1,016,63 2 97.28	14	7 seconds	1.13
·			MiB	sec-		GiB
				onds		
2022 England/leicestershire	120	$438,\!413$	$1{,}118{,}737426.12$	15	8 seconds	1.48
			${ m MiB}$	sec-		GiB
				onds		
2032 England/leicestershire	120	$438,\!413$	$1,\!118,\!737426.12$	15	8 seconds	1.48
			${ m MiB}$	sec-		GiB
				onds		
039 England/leicestershire	120	$459,\!655$	$1,\!164,\!678440.87$	16	8 seconds	1.50
			MiB	sec-		GiB
				onds		
012 England/lincolnshire	134	449,394	1,064,40303.05	15	7 seconds	1.43
			MiB	sec-		GiB
				onds		
2020 England/lincolnshire	134	$475,\!646$	1,098,40 4 19.31	15	7 seconds	1.46
			MiB	sec-		GiB
				onds		
2022 England/lincolnshire	134	$507,\!295$	1,152,29 9 2 7 .55	15	7 seconds	1.47
			${ m MiB}$	sec-		GiB
000 D 1 1/1 1 1 1	104		1 150 00005 55	onds		4.5
032 England/lincolnshire	134	$507,\!295$	1,152,299427.55	16	7 seconds	1.47
			${ m MiB}$	sec-		GiB
000 5 1 1/1: 1 1:	104	700 740	1 150 00000 00	onds	— 1	1 47
039 England/lincolnshire	134	$523,\!548$	1,172,92 3 430.83	16	7 seconds	1.47
			${ m MiB}$	sec-		GiB
010	104	609 409	1 200 2000 22 00	onds	1.1	1 77
012 England/merseyside	184	603,483	1,399,20 \$ 33.96	20	11	1.75
			MiB	sec-	seconds	GiB
0000 England /magazzai 1-	104	699 617	1 495 75559 99	$\frac{\text{onds}}{20}$	11	1.70
2020 England/merseyside	184	632,617	1,435,75 5 53.33	20	11	1.79 C:D
			MiB	sec-	seconds	GiB
000 England / 1	104	661 76C	1 400 11070 01	onds	10	1 00
022 England/merseyside	184	665,766	1,498,51 \$ 70.21	20	10	1.82 C:P
			MiB	sec-	seconds	GiB
022 England /manageraid-	104	66F 766	1 400 51070 01	$\frac{\text{onds}}{21}$	11	1 00
032 England/merseyside	184	665,766	1,498,51 \$ 70.21 MiB	21	11	1.82 GiB
			MID	sec-	seconds	GID
				onds		

year study_area	num_	_maoaa_ho	us ehoo h <u>ds</u> pe pþ lefile	_sinzentim	e commuting	_mentiony
2039 England/merseyside	184	685,165	1,528,03777.48	21	11	1.83
			${ m MiB}$	sec-	seconds	GiB
				onds		
$2012\mathrm{England/norfolk}$	110	$374,\!491$	882,793 333.07	12	5 seconds	1017.16
			MiB	sec-		MiB
				onds		
$020 \mathrm{England/norfolk}$	110	397,770	916,799 348.41	12	5 seconds	1.02
			MiB	sec-		GiB
				onds		
$022\mathrm{England/norfolk}$	110	$432,\!187$	$982,755\ 362.27$	13	5 seconds	1.04
			MiB	sec-		GiB
				onds		
2032 England/norfolk	110	$432,\!187$	982,755 362.27	13	5 seconds	1.04
			MiB	sec-		GiB
				onds		
2039 England/norfolk	110	450,068	1,013,21 4 71.39	13	5 seconds	1.06
			${ m MiB}$	sec-		GiB
				onds		
$2012\mathrm{England/northampton}$	ns l9il re	$289,\!575$	$720,263\ 284.39$	10	4 seconds	941.33
			${ m MiB}$	sec-		MiB
				onds		
$2020 \mathrm{England/northamptor}$	ns l9il re	$316,\!553$	$762,382\ 304.36$	10	4 seconds	981.14
			${ m MiB}$	sec-		MiB
				onds		
$2022 \mathrm{England/northamptor}$	ns l9il re	$352,\!529$	828,003 320.81	11	5 seconds	1005.64
			${ m MiB}$	sec-		MiB
				onds		
$2032 \mathrm{England/northamptor}$	ns l9il re	$352,\!529$	828,003 320.81	11	5 seconds	1005.64
			${ m MiB}$	sec-		MiB
				onds		
$2039 \mathrm{England/northamptor}$	ns !9il re	$370,\!555$	855,812 328.03	11	5 seconds	1016.86
			${ m MiB}$	sec-		MiB
				onds		
2012 England/northumberl	lan 40	138,928	$315,894\ 120.66$	5 sec-	1 second	423.11
			${ m MiB}$	onds		MiB
$020 \mathrm{England/northumberl}$	lan 40	$143,\!516$	$322,616\ 121.94$	5 sec-	1 second	423.87
			${ m MiB}$	onds		MiB
2022 England/northumberl	lan#0	148,792	$333,\!456\ 122.06$	5 sec-	1 second	421.48
			${ m MiB}$	onds		MiB
$2032 \mathrm{England/northumberl}$	lan#0	148,792	$333,\!456\ 122.06$	5 sec-	1 second	421.48
			${ m MiB}$	onds		MiB

year study_area	num	_maoaa_ho	us ehoo h <u>ds</u> pe pb<u>le</u>file_	_sizentim	$e commuting_{\underline{}}$	_mentiony
2039 England/northumber	rlan#0	150,259	337,186 122.24	5 sec-	1 second	421.48
,			MiB	onds		MiB
2012 England/north-	138	$460,\!050$	1,085,067413.05	16	7 seconds	1.45
yorkshire			MiB	sec-		GiB
				onds		
2020 England/north-	138	$478,\!639$	$1,\!107,\!92823.18$	16	7 seconds	1.47
yorkshire			${ m MiB}$	sec-		GiB
				onds		
2022 England/north-	138	$499,\!392$	$1,\!134,\!723420.60$	16	7 seconds	1.45
yorkshire			${ m MiB}$	sec-		GiB
				onds		
$2032 \mathrm{England/north}$	138	$499,\!392$	$1{,}134{,}723420.60$	16	7 seconds	1.45
yorkshire			${ m MiB}$	sec-		GiB
				onds		
2039 England/north-	138	509,099	$1,\!143,\!8921.52$	16	7 seconds	1.46
yorkshire			${ m MiB}$	sec-		GiB
				onds		
2012 England/nottingham	nshi 16 8	460,022	$1{,}123{,}00{2\!\!4}32.35$	16	8 seconds	1.49
			${ m MiB}$	sec-		GiB
				onds		
020 England/nottingham	nshi 16 8	486,163	$1,\!169,\!489453.68$	16	8 seconds	1.53
			MiB	sec-		GiB
				onds		
$1022 \mathrm{England/nottingham}$	nshi 16 8	522,944	$1,\!248,\!80473.35$	17	8 seconds	1.56
			MiB	sec-		GiB
				onds		
2032 England/nottingham	nshi 16 8	522,944	1,248,80473.35	17	8 seconds	1.56
			MiB	sec-		GiB
	_			onds		
2039 England/nottingham	nshi 16 8	$543,\!291$	1,281,812482.21	18	9 seconds	1.66
			MiB	sec-		GiB
				onds	_	
2012 England/oxfordshire	86	$261,\!235$	671,997 260.43	9 sec-	3 seconds	852.84
			MiB	onds		MiB
2020 England/oxfordshire	86	274,908	695,490 271.62	10	4 seconds	918.90
			MiB	sec-		MiB
1000 F 1 1/ 4 13:	0.0	200 200	H 00 000 0 H 7 10	onds	, ,	010 01
2022 England/oxfordshire	86	293,368	729,866 275.40	10	4 seconds	919.34
			MiB	sec-		MiB
				onds		

year study_area	num_	_maoaaa_hou	us ehoo hdsp	e pþ lefile	_sizentim	${ m e\ commuting}$	_mentiong_u	asa
2032 England/oxfordshire	86	293,368	729,866	275.40	10	4 seconds	919.34	
				MiB	sec-		MiB	
					onds			
2039 England/oxfordshire	86	$303,\!035$	743,227	277.51	10	4 seconds	922.19	
				MiB	sec-		MiB	
					onds			
2012 England/rutland	5	14,912	38,314	16.37	3 sec-	1 second	54.07	
				MiB	onds		MiB	
$2020\mathrm{England/rutland}$	5	16,698	40,381	17.09	3 sec-	1 second	57.95	
				MiB	onds		MiB	
2022 England/rutland	5	18,198	44,193	18.26	3 sec-	1 second	60.08	
				MiB	onds		MiB	
2032 England/rutland	5	18,198	44,193	18.26	3 sec-	1 second	60.08	
				MiB	onds		MiB	
2039 England/rutland	5	18,914	45,659	18.71	3 sec-	1 second	61.20	
				MiB	onds		MiB	
2012 England/shropshire	62	197,768	483,414	186.29	7 sec-	2 seconds	550.97	
·				MiB	onds		MiB	
2020 England/shropshire	62	211,035	508,233	195.76	7 sec-	2 seconds	568.62	
				MiB	onds		MiB	
2022 England/shropshire	62	$228,\!285$	558,755	207.29	7 sec-	2 seconds	740.58	
				MiB	onds		MiB	
2032 England/shropshire	62	$228,\!285$	558,755	207.29	7 sec-	2 seconds	740.58	
- , -				MiB	onds		MiB	
2039 England/shropshire	62	236,015	581,476	213.23	7 sec-	2 seconds	749.82	
- , -				MiB	onds		MiB	
2012 England/somerset	124	329,040	790,346	303.62	11	4 seconds	970.03	
- ,				MiB	sec-		$_{ m MiB}$	
					onds			
2020 England/somerset	124	353,976	822,271	317.43	11	4 seconds	996.71	
- ,				MiB	sec-		MiB	
					onds			
2022 England/somerset	124	388,675	880,441	331.61	12	4 seconds	1018.53	
,		,	,	MiB	sec-		MiB	
					onds			
2032 England/somerset	124	388,675	880,441	331.61	12	4 seconds	1018.53	
Ų į		,	,	MiB	sec-		MiB	
					onds			
2039 England/somerset	124	406,157	906,545	339.87	12	4 seconds	1.01	
J , a		,,	.,-	MiB	sec-		GiB	
					onds			

year study_area	num_	_m noan _ho	us edwoh dspe pþ lefile_	_sinzaentim	$\frac{1}{2}$ e commuting	_mentiony
2012 England/south-	172	566,664	1,372,43 5 28.11	20	11	1.75
yorkshire		,	MiB	sec-	seconds	GiB
				onds		
$2020 \mathrm{England/south}$	172	597,694	$1,\!418,\!84$ $\!648.59$	21	11	1.79
yorkshire			${ m MiB}$	sec-	seconds	GiB
				onds		
$2032 \mathrm{England/south}$	172	$637,\!411$	1,493,54463.91	21	11	1.81
yorkshire			${ m MiB}$	sec-	seconds	GiB
				onds		
2039 England/south-	172	$659,\!843$	1,531,31 3 75.31	22	12	1.83
yorkshire			${ m MiB}$	sec-	seconds	GiB
				onds		
2012 England/staffordshire	143	$464,\!441$	$1,\!111,\!144\!\!\!\!/25.27$	16	8 seconds	1.47
			MiB	sec-		GiB
				onds		
2020 England/staffordshire	143	486,645	1,139,752437.51	16	8 seconds	1.49
			MiB	sec-		GiB
				onds		
2022 England/staffordshire	143	510,634	1,188,857444.87	17	8 seconds	1.50
			MiB	sec-		GiB
	1.40	F 10.604	1 100 058 11 05	onds	0 1	1 50
2032 England/staffordshire	143	510,634	1,188,857444.87	17	8 seconds	1.50
			MiB	sec-		GiB
2020 7 1 1/4 6 11:	1.40	*00 000	1 015 00050 04	onds	0 1	1 50
2039 England/staffordshire	143	$522,\!882$	1,215,00@452.94	17	8 seconds	1.52
			MiB	sec-		GiB
2012 Fl 1/#-11-	00	196 149	207 240 100 12	onds	11	440.27
2012 England/suffolk	90	136,142	327,349 128.13 MiB	5 sec-	1 second	440.37
2020 England/suffolk	90	146,277		onds	1 second	MiB
2020 England/Sunoik	90	140,277	333,781 130.90 MiB	5 sec-	1 second	445.14 MiB
2022 England/suffolk	90	150 000		onds	1 goeond	442.14
2022 England/Sunoik	90	159,882	344,534 130.76 MiB	5 sec-	1 second	
2032 England/suffolk	90	159,882	344,534 130.76	$\frac{\text{onds}}{5 \text{ sec}}$	1 second	MiB 442.14
2002 Eligianu/Sulloik	90	109,002	344,554 150.76 MiB	onds	1 Second	442.14 MiB
2039 England/suffolk	90	166,718	350,358 132.54	5 sec-	1 second	446.24
2009 Eligiana, sunoik	90	100,710	MiB	onds	1 Second	440.24 MiB
		450 100		21	13	1.55
2012 England /surrey	151	458 1118	I INX II ZIAN AU			
2012 England/surrey	151	458,108	1,168,11 2 56.50 MiB	sec-	seconds	GiB

year study_area	num_	_manaa_ho	us ehoo h <u>ds</u> pe pþ <u>le</u> file_	_sinzentim	e commuting_	_mentione_
2020 England/surrey	151	480,930	1,195,509472.89	21	13	1.58
			MiB	$\frac{\text{sec-}}{\text{onds}}$	seconds	GiB
2022 England/surrey	151	518,720	1,214,557467.03	21	13	1.56
5 , ,		,	MiB	$\frac{\text{sec-}}{\text{onds}}$	seconds	GiB
2032 England/surrey	151	518,720	$1,\!214,\!557467.03$	21	13	1.56
			MiB	sec- onds	seconds	GiB
2039 England/surrey	151	538,941	$1,\!221,\!227464.71$	21	13	1.64
			MiB	sec- onds	seconds	GiB
2012 England/tyne-and-	145	483,909	$1{,}119{,}03{@427.35}$	15	7 seconds	1.47
wear			MiB	$\frac{\text{sec-}}{\text{onds}}$		GiB
$2020 \mathrm{England/tyne}$ -and-	145	$501,\!383$	1,143,19439.09	15	7 seconds	1.50
wear			MiB	$\frac{\text{sec-}}{\text{onds}}$		GiB
2022 England/tyne-and-	145	521,777	$1,\!168,\!078440.03$	15	7 seconds	1.49
wear			MiB	$\frac{\text{sec-}}{\text{onds}}$		GiB
2032 England/tyne-and-	145	521,777	$1,\!168,\!078440.03$	14	6 seconds	1.49
wear			MiB	$\frac{\text{sec-}}{\text{onds}}$		GiB
2039 England/tyne-and-	145	$532,\!652$	$1,\!177,\!34041.36$	15	7 seconds	1.58
wear			MiB	$\frac{\text{sec-}}{\text{onds}}$		GiB
2012 England/warwickshire	108	$361,\!467$	896,673 347.44	13	6 seconds	1.03
			MiB	$\frac{\text{sec-}}{\text{onds}}$		GiB
2020 England/warwickshire	108	$392,\!639$	958,833 373.63	14	6 seconds	1.08
			MiB	$\frac{\text{sec-}}{\text{onds}}$		GiB
2022 England/warwickshire	108	$432,\!682$	1,061,95205.95	15	7 seconds	1.44
			MiB	$\frac{\text{sec-}}{\text{onds}}$		GiB
2032 England/warwickshire	108	$432,\!682$	1,061,95205.95	14	7 seconds	1.44
			MiB	$\frac{\text{sec-}}{\text{onds}}$		GiB

year study_area	num_	_m noan _hou	s eduod dspe ppl efile_	_sizzentime com	muting_ mention y
2039 England/warwickshire	108	454,732	1,112,23@24.10	15 7 se	conds 1.47
			${ m MiB}$	sec-	${ m GiB}$
				onds	
$2012 \mathrm{England/west}$	314	$958,\!034$	$2,\!477,\!39990.27$	56 38	3.24
$\operatorname{midlands}$			MiB	sec- seco	onds GiB
				onds	
$2020 \mathrm{England/west}$	314	1,002,273	2,572,395.01	58 40	3.33
midlands			GiB	sec- seco	onds GiB
				onds	
2022 England/west-	314	1,046,146	2,664,228.04	60 41	3.37
midlands			GiB	sec- seco	onds GiB
2000 F 1 1/	014	1 0 0 0 0 1 0	2 702 217 21	onds	2 7 7
2032 England/west-	314	1,079,612	2,706,242.04	61 41	3.55
midlands			GiB	sec- seco	onds GiB
1000 F 1 1/	014	1 100 000	2 707 000 07	onds	9.50
2039 England/west-	314	1,128,890	2,787,990.07	$62 \qquad 42$	3.59
midlands			GiB	sec- seco	onds GiB
0019 El l /+	100	249 766	026 646 201 17	onds	1- 1004 45
2012 England/west-sussex	100	348,766	836,646 321.17 MiB		conds 1004.45 MiB
			MID	sec- onds	MID
020 England/west-sussex	100	375,837	871,029 337.76		conds 1.01
020 Eligiand/ west-sussex	100	313,031	MiB	sec-	GiB
			MID	onds	GID
022 England/west-sussex	100	419,347	931,573 350.11		conds 1.03
022 Eligiand, west-sussex	100	413,541	MiB	sec-	GiB
			WIID	onds	GID
2032 England/west-sussex	100	419,347	931,573 350.11		conds 1.03
002 Ziigiaiia/ West susseii	100	110,011	MiB	sec-	GiB
				onds	5
2039 England/west-sussex	100	442,292	958,567 356.77		conds 1.04
3 7, 11- 11- 11- 11- 11- 11- 11- 11- 11-		, -	MiB	sec-	GiB
				onds	
2012 England/west-	299	921,242	2,271,83 3 93.87	46 31	3.05
yorkshire		,	MiB	sec- seco	
v				onds	
2020 England/west-	299	963,460	2,339,93930.47	48 33	3.12
yorkshire		,	MiB	sec- seco	onds GiB
				onds	

year study_area	num_	_m noan _hou	s ehooh dspe pp lefile	_sizentime	e commuting_	_mentiony
2022 England/west-	299	1,021,830	2,434,90 2 45.77	48	33	3.13
yorkshire			MiB	sec-	seconds	GiB
				onds		
$2032 \mathrm{England/west}$	299	1,021,830	2,434,90 2 945.77	49	33	3.13
yorkshire			MiB	sec-	seconds	GiB
2020 F 1 1/	200	1 050 050	0 401 05055 40	onds	22	2.22
2039 England/west-	299	1,053,859	, ,	49	33	3.32
yorkshire			MiB	sec-	seconds	GiB
2012 England /wiltahina	90	205 600	704 401 974 59	$\frac{1}{2}$ onds	2 gogonda	021.09
2012 England/wiltshire	89	285,600	704,491 274.58 MiB	9 sec- onds	3 seconds	921.08 MiB
2020 England/wiltshire	89	309,159	735,088 288.20	9 sec-	3 seconds	947.43
2020 Dugiana/ wittenine	OÐ	909,109	MiB	onds	o becomes	947.45 MiB
2022 England/wiltshire	89	335,400	774,105 292.69	10	3 seconds	949.16
		555,100	MiB	sec-	5 55551145	MiB
				onds		
2032 England/wiltshire	89	335,400	774,105 292.69	10	3 seconds	949.16
0 ,		,	MiB	sec-		MiB
				onds		
2039 England/wiltshire	89	$348,\!866$	$792,075\ 296.40$	10	3 seconds	955.08
			${ m MiB}$	sec-		MiB
				onds		
2012 England/worcestershi	re 85	240,958	578,628 221.47	8 sec-	3 seconds	770.62
			MiB	onds		MiB
2020 England/worcestershi	re 85	$255,\!594$	601,116 231.59	8 sec-	3 seconds	790.42
0000 T 1 1/ / 1:	0.5	974 900	MiB	onds	0 1	MiB
2022 England/worcestership	re 85	274,309	644,922 241.99	8 sec-	3 seconds	849.84
2032 England/worcestershi	ro 85	274,309	MiB 644,922 241.99	onds	3 seconds	MiB 849.84
2092 England/ worcestersiii.	1609	414,309	MiB	8 sec- onds	o seconds	649.64 MiB
2039 England/worcestershi	re 85	283,275	666,303 248.38	8 sec-	3 seconds	861.38
2000 England, wordestersin	1000	200,210	MiB	onds	o seconds	MiB
2012 special/northwest tra	an s 129 nr	in&.653.096		5	4 minutes	7.74
	1	. , ,	GiB	min-		GiB
				utes		
$2020\mathrm{special/northwest_tra}$	ns (29 nr	in@,788,624	6,616,112.56	5	4 minutes	7.95
- , — —	-	. ,	$_{ m GiB}$	min-		GiB
				utes		
$2022 { m special/northwest_tra}$	ns (29 nr	in@,960,285	6,908,374.62	5	4 minutes	8.02
			GiB	min-		GiB
				utes		

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$032\mathrm{special/northwest_tra}$	ınsq29nı	nin@,960,285	6,908,37	4 2.62	5	5 minutes	8.02
,				GiB	min-		GiB
					utes		
$039\mathrm{special/northwest_tra}$	ın sp29 nı	$\sin 3.058,114$	7,059,12		5	5 minutes	8.08
				GiB	min-		GiB
					utes		
012 Wales/bridgend-and-	38	119,725	283,159		5 sec-	1 second	382.24
neath-port-talbot				MiB	onds		MiB
020 Wales/bridgend-and-	38	123,909	289,896		5 sec-	1 second	387.45
neath-port-talbot				MiB	onds		MiB
022 Wales/bridgend-and-	38	124,921	292,227		4 sec-	1 second	387.72
neath-port-talbot	0.0	100 001	201 707	MiB	onds		MiB
32 Wales/bridgend-and-	38	128,601	301,529		5 sec-	1 second	390.82
neath-port-talbot	0.0	100 - 10	20 - 202	MiB	onds		MiB
039 Wales/bridgend-and-	38	129,740	307,260		5 sec-	1 second	391.29
neath-port-talbot				MiB	onds		MiB
012 Wales/cardiff-and-	63	199,208	484,182		6 sec-	2 seconds	558.19
vale-of-glamorgan	0.0	244.0=0	400.0-0	MiB	onds		MiB
20 Wales/cardiff-and-	63	$214,\!676$	499,272		7 sec-	2 seconds	572.89
vale-of-glamorgan	0.0	210.001	* 00 * 00	MiB	onds	0 1	MiB
22 Wales/cardiff-and-	63	218,981	502,763		6 sec-	2 seconds	576.04
vale-of-glamorgan	0.0	240 112	* • • • • • • • • • • • • • • • • • • •	MiB	onds	2 1	MiB
32 Wales/cardiff-and-	63	240,112	522,526		7 sec	2 seconds	577.84
vale-of-glamorgan	0.0	054.100	F01 F40	MiB	onds	0 1	MiB
39 Wales/cardiff-and-	63	254,162	531,549		7 sec-	2 seconds	737.29
vale-of-glamorgan	20	104 601	000 501	MiB	onds	1 1	MiB
012 Wales/central-valleys	38	124,691	296,581		5 sec-	1 second	396.20
000 Walaa / 00r-+ 1 11	20	120.079	201 007	MiB	onds	1	MiB
20 Wales/central-valleys	38	130,072	301,907	117.77 MiB	18	1 second	400.97 MiB
				MIID	$\frac{\text{sec-}}{\text{onds}}$		MID
022 Wales/central-valleys	38	121 209	303,557	119.40	8 sec-	1 second	424.47
022 Waies/Centrai-vaneys	30	131,383	505,557	MiB	onds	1 second	424.47 MiB
032 Wales/central-valleys	20	126 404	310,032			1 second	421.13
552 wates/central-valleys	38	136,404	310,032	118.04 MiB	5 sec- onds	1 second	421.13 MiB
39 Wales/central-valleys	38	138,735	314 709		5 sec-	1 second	423.02
oos wates/central-valleys	50	190,799	314,703	119.17 MiB	onds	1 Second	423.02 MiB
012 Wales/conwy-and-	30	92,732	211,205		4 sec-	1 second	251.47
denbighshire	90	94,134	211,200	MiB		1 Second	251.47 MiB
020 Wales/conwy-and-	30	95,314	213,302		onds	1 second	253.63
-	9 0	99,914	213,302		4 sec-	1 Second	
denbighshire				MiB	onds		MiB

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2022 Wales/conwy-and-	30	95,881	214,182	81.85	4 sec-	1 second	254.21
denbighshire				MiB	onds		MiB
032 Wales/conwy-and-	30	97,683	218,122	81.11	4 sec-	1 second	251.17
denbighshire				MiB	onds		MiB
039 Wales/conwy-and-	30	97,687	220,933	80.92	4 sec-	1 second	249.76
denbighshire		,	,	MiB	onds		MiB
012 Wales/flintshire-and-	38	$122,\!180$	288,696	113.32	5 sec-	1 second	393.63
wrexham	20	105 000	202.054	MiB	onds		MiB
020 Wales/flintshire-and-	38	$127,\!660$	292,056		5 sec-	1 second	395.27
wrexham				MiB	onds		MiB
022 Wales/flintshire-and-	38	$129,\!007$	292,644		5 sec-	1 second	396.56
wrexham				MiB	onds		MiB
032 Wales/flintshire-and-	38	$134,\!527$	$292,\!817$		5 sec-	1 second	410.92
wrexham				MiB	onds		MiB
039 Wales/flintshire-and-	38	$136,\!425$	$293,\!540$		5 sec-	1 second	410.77
wrexham				MiB	onds		MiB
012 Wales/gwent-valleys	46	$144,\!178$	$341,\!543$	132.17	5 sec-	1 second	451.03
				MiB	onds		MiB
020 Wales/gwent-valleys	46	148,386	344,566	132.83	5 sec-	1 second	450.89
				MiB	onds		MiB
22 Wales/gwent-valleys	46	149,374	345,498	132.72	5 sec-	1 second	450.23
, ,				MiB	onds		MiB
32 Wales/gwent-valleys	46	151,842	347,976	130.50	5 sec-	1 second	442.86
, 0		,	,	MiB	onds		MiB
039 Wales/gwent-valleys	46	151,729	350,397		5 sec-	1 second	443.03
			000,000	MiB	onds		MiB
012 Wales/gwynedd	17	52,926	122,595		3 sec-	1 second	141.47
		0=,0=0	,000	MiB	onds	_ 5550Ha	MiB
020 Wales/gwynedd	17	55,064	124,569		3 sec-	1 second	143.70
0_0 // mios/ 8 // jiiodd		55,001	121,000	MiB	onds	2 5000Ha	MiB
022 Wales/gwynedd	17	55,683	125,030		3 sec-	1 second	143.45
o maios/8wymoda	±1	55,555	120,000	MiB	onds	1 5000Hq	MiB
032 Wales/gwynedd	17	58,372	128,844		3 sec-	1 second	143.80
302 wates/gwynedd	11	50,512	120,044	49.81 MiB	onds	1 second	143.60 MiB
)20 Wolos /mm odd	17	50 746	120 049			1 seems	
039 Wales/gwynedd	17	59,746	130,948		3 sec-	1 second	145.62 M;D
019337-1/:-1	0	20.707	CO 010	MiB	onds	1 1	MiB
012 Wales/isle-of-	9	30,797	69,919	27.65	3 sec-	1 second	96.81
anglesey	0	01 000	00 0 15	MiB	onds		MiB
020 Wales/isle-of-	9	31,366	69,845	27.85	3 sec-	1 second	97.39
anglesey				MiB	onds		MiB

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2022 Wales/isle-of-	9	31,488	69,864	27.91	3 sec-	1 second	97.71
anglesey				MiB	onds		MiB
2032 Wales/isle-of-	9	31,601	69,502	27.09	3 sec-	1 second	95.51
anglesey				MiB	onds		MiB
039 Wales/isle-of-	9	31,337	69,423	26.91	3 sec-	1 second	95.37
anglesey				MiB	onds		MiB
012 Wales/monmouthshire	- 31	100,402	240,491	94.44	4 sec-	1 second	280.40
and-newport				MiB	onds		MiB
020 Wales/monmouthshire	- 31	104,394	250,185	98.11	4 sec-	1 second	286.97
and-newport				MiB	onds		MiB
022 Wales/monmouthshire	- 31	105,481	253,282	99.27	4 sec-	1 second	289.03
and-newport				MiB	onds		MiB
2032 Wales/monmouthshire	- 31	109,752	265,785	102.21	4 sec-	1 second	371.39
and-newport				MiB	onds		MiB
039 Wales/monmouthshire	- 31	111,246	273,319	103.90	4 sec-	1 second	373.82
and-newport				MiB	onds		MiB
012 Wales/powys	19	59,028	132,725	51.21	4 sec-	1 second	185.06
,				MiB	onds		MiB
020 Wales/powys	19	59,972	132,328	50.60	4 sec-	1 second	183.38
·				MiB	onds		MiB
022 Wales/powys	19	60,190	132,467	50.46	4 sec-	1 second	182.88
				MiB	onds		MiB
032 Wales/powys	19	$59,\!586$	133,010	49.63	4 sec-	1 second	180.64
				MiB	onds		MiB
2039 Wales/powys	19	57,969	133,514	49.35	4 sec-	1 second	179.80
				MiB	onds		MiB
2012 Wales/south-west-	50	165,004	383,260	145.79	5 sec-	1 second	474.32
wales				MiB	onds		MiB
2020 Wales/south-west-	50	170,327	385,937	146.53	5 sec-	1 second	474.47
wales				MiB	onds		MiB
2022 Wales/south-west-	50	171,623	386,901	147.00	5 sec-	1 second	476.10
wales				MiB	onds		MiB
2032 Wales/south-west-	50	175,897	392,107	145.19	6 sec-	1 second	469.31
wales				MiB	onds		MiB
039 Wales/south-west-	50	$176,\!482$	394,303	144.53	6 sec-	1 second	467.47
wales				MiB	onds		MiB
012 Wales/swansea	31	104,423	242,128	93.10	4 sec-	1 second	276.14
·			•	MiB	onds		MiB
2020 Wales/swansea	31	110,304	247,820	95.72	4 sec-	1 second	281.38
		,	,	MiB			MiB

year study_area	num_	_maonana_hou	us chwoh dspe pþ lefile_	_sirzantime	e commuting_	mentiony_usage
2022 Wales/swansea	31	111,940	249,098 96.10	4 sec-	1 second	282.16
			MiB	onds		MiB
2032 Wales/swansea	31	119,141	$257,\!653\ 98.28$	4 sec-	1 second	285.53
			MiB	onds		MiB
2039 Wales/swansea	31	123,450	262,306 99.93	4 sec-	1 second	366.61
			MiB	onds		MiB

Notes:

- pb_file_size refers to the size of the uncompressed protobuf file in data/output/
- The total runtime is usually dominated by matching workers to businesses, so commuting_runtime gives a breakdown
- Measuring memory usage of Linux processes isn't straightforward, so memory_usage should just be a guide
- These measurements were all taken on one developer's laptop, and they don't represent multiple runs. This table just aims to give a general sense of how long running takes.
 - That machine has 24 cores, which matters for the parallelized commuting calculation.
- The time *usually* doesn't include downloading or decompressing raw data. For some areas, it might!
- scripts/collect_stats.py produces the table above