
CLK hash Documentation

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N1 Analytics

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`clkhsh` is a python implementation of cryptographic linkage key hashing as described by Rainer Schnell, Tobias Bachteler, and Jörg Reiher in *A Novel Error-Tolerant Anonymous Linking Code* [[Schnell2011](#)].

Clkhash is Apache 2.0 licensed, supports Python versions 2.7+, 3.4+, and runs on Windows, OSX and Linux.

Install with pip:

```
pip install clkhsh
```

Hint: If you are interested in comparing CLKs (i.e carrying out record linkage) you might want to check out [anonlink](#) - our Python library for computing similarity scores, and best guess matches between two sets of cryptographic linkage keys.

1.1 Tutorial

For this tutorial we are going to process a data set for private linkage with clkhsh using the Python API. Note you can also use the command line tool.

The Python package `recordlinkage` has a [tutorial](#) linking data sets in the clear, we will try duplicate that in a privacy preserving setting.

First install clkhsh, recordlinkage and a few data science tools (pandas and numpy).

```
In [ ]: !pip install -U clkhsh recordlinkage numpy pandas
```

```
In [24]: import io
import numpy as np
import pandas as pd
```

```
In [17]: import clkhsh
import recordlinkage
from recordlinkage.datasets import load_febrl4
```

1.1.1 Data Exploration

First we have a look at the dataset.

```
In [4]: dfA, dfB = load_febrl4()
```

```
dfA.head()
```

```
Out[4]: given_name  surname street_number  address_1 \
rec_id
rec-1070-org  michaela  neumann      8      stanley street
rec-1016-org  courtney  painter      12     pinkerton circuit
rec-4405-org   charles   green      38    salkauskas crescent
rec-1288-org   vanessa   parr      905    macquoid place
rec-3585-org   mikayla  malloney     37    randwick road
```

	address_2	suburb	postcode	state	\
rec_id					
rec-1070-org	miami	winston hills	4223	nsw	
rec-1016-org	bega flats	richlands	4560	vic	
rec-4405-org	kela	dapto	4566	nsw	
rec-1288-org	broadbridge manor	south grifton	2135	sa	
rec-3585-org	avalind	hoppers crossing	4552	vic	

	date_of_birth	soc_sec_id
rec_id		
rec-1070-org	19151111	5304218
rec-1016-org	19161214	4066625
rec-4405-org	19480930	4365168
rec-1288-org	19951119	9239102
rec-3585-org	19860208	7207688

For this linkage we will **not** use the social security id column.

```
In [18]: dfA.columns
Out[18]: Index(['given_name', 'surname', 'street_number', 'address_1', 'address_2',
               'suburb', 'postcode', 'state', 'date_of_birth', 'soc_sec_id'],
              dtype='object')

In [41]: a_csv = io.StringIO()
          dfA.to_csv(a_csv)
          a_csv.seek(0)

Out[41]: 0
```

1.1.2 Linkage Schema Definition

A basic schema definition instructs clckhash how to treat each column. The identifiers are found in [clckhash/identifier_types.py](#). The INDEX columns will not become part of the CLK.

Note: The schema specification is under heavy renovation for the next version.

```
In [42]: column_metadata = [
          {"identifier": 'INDEX'},
          {"identifier": 'NAME Surname'},
          {"identifier": 'NAME First Name'},
          {"identifier": 'ADDRESS House Number'},
          {"identifier": 'ADDRESS Place Name'},
          {"identifier": 'ADDRESS Place Name'},
          {"identifier": 'ADDRESS Place Name'},
          {"identifier": 'ADDRESS POSTCODE'},
          {"identifier": 'ADDRESS Place Name'},
          {"identifier": 'DOB YYYY/MM/DD'},
          {"identifier": 'INDEX'}
        ]

In [43]: schema = clckhash.schema.get_schema_types(column_metadata)
```

1.1.3 Hash the data

We can now hash our PII data from the CSV file using our defined schema. We must provide two *secret keys* to this command - these keys have to be used by both parties hashing data.

Knowledge of these keys is sufficient to reconstruct the PII information from a CLK! **Do not share these keys with anyone, except the other participating party.**

```
In [44]: hashed_data_a = clkh.hash.generate_clk_from_csv(a_csv, ('key1', 'key2'), schema)
generating CLKs: 100%|| 5.00K/5.00K [00:02<00:00, 1.46Kclk/s, mean=916, std=27.1]
```

1.1.4 Inspect the output

clkh.hash has hashed the PII, creating a Cryptographic Longterm Key for each entity. The output of `generate_clk_from_csv` shows that the mean popcount is quite high (916 out of 1024) which can effect accuracy.

```
In [45]: len(hashed_data_a)
```

```
Out[45]: 5000
```

Each CLK is serialized in a JSON friendly base64 format:

```
In [46]: hashed_data_a[0]
```

```
Out[46]: '+79/+3//33/+O////fv//N67/7/+u/vP////+3/+7//d////////ft////////vM/7v/v//1+///9/fv9f7/zu//5+/1'
```

1.1.5 Hash data set B

Now we hash the second dataset using the same keys and same schema.

```
In [47]: b_csv = io.StringIO()
         dfB.to_csv(b_csv)
         b_csv.seek(0)
         hashed_data_b = clkh.hash.generate_clk_from_csv(b_csv, ('key1', 'key2'), schema)
```

```
generating CLKs: 100%|| 5.00K/5.00K [00:02<00:00, 2.44Kclk/s, mean=909, std=32]
```

```
In [48]: len(hashed_data_b)
```

```
Out[48]: 5000
```

1.1.6 Wrapping Up

That is all for this tutorial, next you might want to look at comparing the CLKs with [anonlink](#), or uploading them to an Entity Service.

Note the `clkh.hash` command line tool includes commands to upload to an entity service run by Data61.

1.2 Command Line Tool

This command line tool can be used to process PII data into Cryptographic Longterm Keys.

The command line tool can be accessed in two ways:

- Using the `clktutil` script which should have been added to your path during installation.
- directly running the python module `clkh.hash.cli` with `python -m clkh.hash.cli`.

1.2.1 Hashing

The command line tool `clkutil` can be used to hash a csv file of personally identifiable information. The tool needs to be provided with keys and a [Hashing Schema](#); it will output a file containing json serialized hashes.

Example

Assume a csv (`fake-pii.csv`) contains rows like the following:

```
0,Libby Slemmer,1933/09/13,F
1,Garold Staten,1928/11/23,M
2,Yaritza Edman,1972/11/30,F
```

It can be hashed using `clkutil` with:

```
$ clkutil hash --schema simple-schema.json fake-pii.csv horse staple clk.json
```

Where:

- `horse staple` is the two part secret key that both participants will use to hash their data.
- `simple-schema.json` is a [Hashing Schema](#) describing how to hash the csv. E.g, ignore the first column, use bigram tokens of the name, use positional unigrams of the date of birth etc.
- `clk.json` is the output file.

1.2.2 Data Generation

The cli tool has an option for generating fake pii data.

```
$ clkutil generate 1000 fake-pii-out.csv
$ head -n 4 fake-pii-out.csv
INDEX,NAME freetext,DOB YYYY/MM/DD,GENDER M or F
0,Libby Slemmer,1933/09/13,F
1,Garold Staten,1928/11/23,M
2,Yaritza Edman,1972/11/30,F
```

The yaml schema used for the generated data is the following:

```
- identifier: "INDEX"
  notes: "Ignored"
- identifier: "NAME freetext"
- identifier: "DOB YYYY/MM/DD"
- identifier: "GENDER M or F"
```

1.2.3 Benchmark

A quick hashing benchmark can be carried out to determine the rate at which the current machine can generate 10000 clks from a simple schema (data as generated [above](#)):

```
python -m clkhash.cli benchmark
generating CLKs: 100%          10.0K/10.0K [00:01<00:00, 7.72Kclk/s, mean=521,
↪std=34.7]
10000 hashes in 1.350489 seconds. 7.40 KH/s
```

As a rule of thumb a single modern core will hash around 1M entities in about 20 minutes.

Note: Hashing speed is effected by the number of features and the corresponding schema. Thus these numbers will, in general, not be a good predictor for the performance of a specific use-case.

The output shows a running mean and std deviation of the generated clks' popcounts. This can be used as a basic sanity check - ensure the CLK's popcount is not around 0 or 1024.

1.2.4 Interaction with Entity Service

There are several commands that interact with a REST api for carrying out privacy preserving linking. These commands are:

- status
- create
- upload
- results

1.3 Hashing Schema

Caution: This document and the referenced `schema.json` file are still in a draft status.

As CLKs are usually used for privacy preserving linkage, it is important, that participating organisations agree on how raw personally identifiable information is hashed to create the CLKs.

We call the configuration of how to create CLKs a *hashing schema*. The organisations agree on one hashing schema as configuration to ensure that the have been created in the same way.

This aims to be an open standard such that different client implementations could take the schema and create identical CLKs given the same data.

The hashing-schema is a detailed description of exactly what is fed to the hashing operation, along with any configuration for the hashing itself.

The format of the hashing schema is defined in a separate JSON Schema document [hashing-schema.json](#).

1.3.1 Basic Structure

A hashing schema consists of three parts:

- *version*, contains the version number of the hashing schema
- *clkConfig*, CLK wide configuration, independent of features
- *features*, configuration that is specific to the individual features

1.3.2 Example Schema

```
{
  "version": 1,
  "clkConfig": {
    "l": 1024,
    "k": 20,
    "hash": {
      "type": "doubleHash"
    },
    "kdf": {
      "type": "HKDF"
    }
  },
  "features": [
    {
      "identifier": "full name",
      "format": {
        "type": "string",
        "maxLength": 30,
        "encoding": "utf-8"
      },
      "hashing": { "ngram": 2 }
    },
    {
      "identifier": "gender",
      "format": {
        "type": "enum",
        "values": ["M", "F", "O"]
      },
      "hashing": { "ngram": 1 }
    },
    {
      "identifier": "postcode",
      "format": {
        "type": "integer",
        "minimum": 1000,
        "maximum": 9999
      },
      "hashing": { "ngram": 1, "positional": true }
    }
  ]
}
```

A more advanced example can be found [here](#).

1.3.3 Schema Components

Version

Integer value which describes the version of the hashing schema.

clkConfig

Describes the general construction of the CLK.

name	type	optional	description
l	integer	no	the length of the CLK in bits
k	integer	no	max number of indices per n-gram
xor-Folds	integer	yes	number of XOR folds (as proposed in [Schnell2016]).
kdf	KDF	no	defines the key derivation function used to generate individual secrets for each feature derived from the master secret
hash	Hash	no	defines the hashing scheme to encode the n-grams

KDF

We currently only support HKDF (for a basic description, see <https://en.wikipedia.org/wiki/HKDF>).

name	type	optional	description
type	string	no	must be set to “HKDF”
hash	enum	yes	hash function used by HKDF, either “SHA256” or “SHA512”
salt	string	yes	base64 encoded bytes
info	string	yes	base64 encoded bytes
keySize	integer	yes	size of the generated keys in bytes

Hash

Describes and configures the hash that is used to encode the n-grams.

Choose one of:

- *double hash*, as described in [\[Schnell2011\]](#).

name	type	optional	description
type	string	no	must be set to “doubleHash”
prevent_singularity	boolean	yes	see discussion in https://github.com/n1analytics/clkehash/issues/33

- *blake hash*

name	type	optional	description
type	string	no	must be set to “blakeHash”

featureConfig

A feature is configured in three parts:

- identifier, the name of the feature
- format, describes the expected format of the values of this feature
- hashing, configures the hashing

name	type	optional	description
identifier	string	no	the name of the feature
description	string	yes	free text, ignored by clkhash
hashing	<i>hashingConfig</i>	no	configures feature specific hashing parameters
format	one of: <i>textFormat</i> , <i>textPatternFormat</i> , <i>numberFormat</i> , <i>dateFormat</i> , <i>enumFormat</i>	no	describes the expected format of the feature values

hashingConfig

name	type	optional	description
ngram	integer	no	specifies the n in n-gram (the tokenization of the input values).
positional	boolean	yes	adds the position to the n-grams. String “222” would be tokenized (as uni-grams) to “1 2”, “2 2”, “3 2”
weight	float	yes	positive number, which adjusts the number of hash functions (k) used for encoding. Thus giving this feature more or less importance compared to others.

textFormat

name	type	optional	description
type	string	no	has to be “string”
encoding	enum	yes	one of “ascii”, “utf-8”, “utf-16”, “utf-32”. Default is “utf-8”.
case	enum	yes	one of “upper”, “lower”, “mixed”.
minLength	integer	yes	positive integer describing the minimum length of the input string.
maxLength	integer	yes	positive integer describing the maximum length of the input string.
description	string	yes	free text, ignored by clkhash.

textPatternFormat

name	type	optional	description
type	string	no	has to be “string”
encoding	enum	yes	one of “ascii”, “utf-8”, “utf-16”, “utf-32”. Default is “utf-8”.
pattern	string	no	a regular expression describing the input format.
description	string	yes	free text, ignored by clkhash.

numberFormat

name	type	optional	description
type	string	no	has to be “integer”
minimum	integer	yes	positive integer describing the lower bound of the input values.
maximum	integer	yes	positive integer describing the upper bound of the input values.
description	string	yes	free text, ignored by clkhash.

dateFormat

name	type	optional	description
type	string	no	has to be “date”
format	enum	no	one of [“rfc3339”]. That’s the standard internet format: yyyy-mm-dd.
description	string	yes	free text, ignored by clkhash.

enumFormat

name	type	optional	description
type	string	no	has to be “enum”
values	array	no	an array of items of type “string”
description	string	yes	free text, ignored by clkhash.

1.4 Development

1.4.1 API Documentation

Bloom filter

Generate a Bloom filter

class `clkhash.bloomfilter.NgramEncodings`

Bases: `enum.Enum`

Lists the available schemes for encoding n-grams.

Note: the slightly awkward looking construction with the calls to `partial` and the overwrite of `__call__` are due to compatibility issues with Python 2.7.

BLAKE_HASH = `functools.partial(<function blake_encode_ngrams>)`

uses the BLAKE2 hash function, which is one of the fastest modern hash functions, and does less hash function calls compared to the DOUBLE_HASH based schemes. It avoids one of the exploitable weaknesses of the DOUBLE_HASH scheme. Also see `blake_encode_ngrams()`

DOUBLE_HASH = `functools.partial(<function double_hash_encode_ngrams>)`

the initial encoding scheme as described in Schnell, R., Bachteler, T., & Reiher, J. (2011). A Novel Error-Tolerant Anonymous Linking Code. Also see `double_hash_encode_ngrams()`

`DOUBLE_HASH_NON_SINGULAR = functools.partial(<function double_hash_encode_ngrams_non_singular>, very similar to DOUBLE_HASH, but avoids singularities in the encoding. Also see double_hash_encode_ngrams_non_singular()`

`clkhash.bloomfilter.blake_encode_ngrams(ngrams, key, k, l)`

Computes the encoding of the provided ngrams using the BLAKE2 hash function.

We deliberately do not use the double hashing scheme as proposed in [Schnell2011], because this would introduce an exploitable structure into the Bloom filter. For more details on the weakness, see [Kroll2015].

In short, the double hashing scheme only allows for l^2 different encodings for any possible n-gram, whereas the use of k different independent hash functions gives you $\sum_{j=1}^k \binom{l}{j}$ combinations.

Our construction

It is advantageous to construct Bloom filters using a family of hash functions with the property of *k-independence* to compute the indices for an entry. This approach minimises the change of collisions.

An informal definition of *k-independence* of a family of hash functions is, that if selecting a function at random from the family, it guarantees that the hash codes of any designated k keys are independent random variables.

Our construction utilises the fact that the output bits of a cryptographic hash function are uniformly distributed, independent, binary random variables (well, at least as close to as possible. See [Kaminsky2011] for an analysis). Thus, slicing the output of a cryptographic hash function into k different slices gives you k independent random variables.

We chose Blake2 as the cryptographic hash function mainly for two reasons:

- it is fast.
- in keyed hashing mode, Blake2 provides MACs with just one hash function call instead of the two calls in the HMAC construction used in the double hashing scheme.

Warning: Please be aware that, although this construction makes the attack of [Kroll2015] infeasible, it is most likely not enough to ensure security. Or in their own words:

However, we think that using independent hash functions alone will not be sufficient to ensure security, since in this case other approaches (maybe related to or at least inspired through work from the area of Frequent Itemset Mining) are promising to detect at least the most frequent atoms automatically.

Parameters

- **ngrams** – list of n-grams to be encoded
- **key** – secret key for blake2 as bytes
- **k** – number of hash functions to use per element of the ngrams
- **l** – length of the output bitarray (has to be a power of 2)

Returns bitarray of length l with the bits set which correspond to the encoding of the ngrams

`clkhash.bloomfilter.calculate_bloom_filters(dataset, schema, keys, xor_folds=0)`

Parameters

- **dataset** – A list of indexable records.
- **schema** – An iterable of identifier types.
- **keys** – A tuple of two lists of secret keys used in the HMAC.

- **xor_folds** – number of XOR folds to perform

Returns List of bloom filters as 3-tuples, each containing bloom filter (bitarray), record first element - usually index, bitcount (int)

```
clckhash.bloomfilter.crypto_bloom_filter(record,           tokenizers,           keys,
                                         xor_folds=0,        l=1024,             k=30,
                                         ngram_encoding=<NgramEncodings.DOUBLE_HASH:
                                         functools.partial(<function
                                                         dou-
                                                         ble_hash_encode_ngrams>)>)
```

Makes a Bloom filter from a record with given tokenizers and lists of keys.

Using the method from <http://www.record-linkage.de/-download=wp-grlc-2011-02.pdf>

Parameters

- **record** – plaintext record tuple. E.g. (index, name, dob, gender)
- **tokenizers** – A list of IdentifierType tokenizers (one for each record element)
- **keys** – tuple of tuple of keys for the hash functions as bytes
- **xor_folds** – number of XOR folds to perform
- **l** – length of the Bloom filter in number of bits
- **k** – number of hash functions to use per element
- **ngram_encoding** –

Returns 3-tuple: - bloom filter for record as a bitarray - first element of record (usually an index) - number of bits set in the bloomfilter

```
clckhash.bloomfilter.double_hash_encode_ngrams(ngrams, keys, k, l)
```

Computes the double hash encoding of the provided ngrams with the given keys.

Using the method from <http://www.record-linkage.de/-download=wp-grlc-2011-02.pdf>

Parameters

- **ngrams** – list of n-grams to be encoded
- **keys** – hmac secret keys for md5 and sha1 as bytes
- **k** – number of hash functions to use per element of the ngrams
- **l** – length of the output bitarray

Returns bitarray of length l with the bits set which correspond to the encoding of the ngrams

```
clckhash.bloomfilter.double_hash_encode_ngrams_non_singular(ngrams, keys, k, l)
```

computes the double hash encoding of the provided n-grams with the given keys.

The original construction of [Schnell2011] displays an abnormality for certain inputs: An n-gram can be encoded into just one bit irrespective of the number of k.

Their construction goes as follows: the k different indices g_i of the Bloom filter for an n-gram x are defined as:

$$g_i(x) = (h_1(x) + ih_2(x)) \mod l$$

with $0 \leq i < k$ and l is the length of the Bloom filter. If the value of the hash of x of the second hash function is a multiple of l , then

$$h_2(x) = 0 \mod l$$

and thus

$$g_i(x) = h_1(x) \mod l,$$

irrespective of the value i . A discussion of this potential flaw can be found [here](#).

Parameters

- **ngrams** – list of n-grams to be encoded
- **key_sha1** – hmac secret keys for sha1 as bytes
- **key_md5** – hmac secret keys for md5 as bytes
- **k** – number of hash functions to use per element of the ngrams
- **l** – length of the output bitarray

Returns bitarray of length l with the bits set which correspond to the encoding of the ngrams

`clckhash.bloomfilter.fold_xor(bloomfilter, folds)`

Performs XOR folding on a Bloom filter.

If the length of the original Bloom filter is n and we perform r folds, then the length of the resulting filter is $n / 2^{**r}$.

Parameters

- **bloomfilter** – Bloom filter to fold
- **folds** – number of folds

Returns folded bloom filter

`clckhash.bloomfilter.serialize_bitarray(ba)`

Serialize a bitarray (bloomfilter)

`clckhash.bloomfilter.stream_bloom_filters(dataset, schema_types, keys, xor_folds=0)`

Yield bloom filters

Parameters

- **dataset** – An iterable of indexable records.
- **schema_types** – An iterable of identifier type names.
- **keys** – A tuple of two lists of secret keys used in the HMAC.
- **xor_folds** – number of XOR folds to perform

Returns Yields bloom filters as 3-tuples

CLK

Generate CLK from CSV file

`clckhash.clk.chunks(l, n)`

Yield successive n-sized chunks from l .

`clckhash.clk.generate_clk_from_csv(input, keys, schema_types, no_header=False, progress_bar=True, xor_folds=0)`

`clckhash.clk.generate_clks(pii_data, schema_types, key_lists, xor_folds, callback=None)`

`clkhask.clk.hash_and_serialize_chunk(chunk_pii_data, schema_types, keys, xor_folds)`

Generate Bloom filters (ie hash) from chunks of PII then serialize the generated Bloom filters. It also computes and outputs the Hamming weight (or popcount) – the number of bits set to one – of the generated Bloom filters.

Parameters

- **chunk_pii_data** – An iterable of indexable records.
- **schema_types** – An iterable of identifier type names.
- **keys** – A tuple of two lists of secret keys used in the HMAC.
- **xor_folds** – Number of XOR folds to perform. Each fold halves the hash length.

Returns A list of serialized Bloom filters and a list of corresponding popcounts

identifier types

Convert PII to tokens

`clkhask.identifier_types.identifier_type_from_description(schema_object)`

Convert a dictionary describing a feature into an IdentifierType

Parameters **schema_object** –

Returns An IdentifierType

IdentifierType

class `clkhask.identifier_types.IdentifierType(unigram=False, weight=1, **kwargs)`
 Bases: `object`

Base class used for all identifier types.

Required to provide a mapping of schema to hash type uni-gram or bi-gram.

__call__ (*entry*)

Call self as a function.

__init__ (*unigram=False, weight=1, **kwargs*)

Parameters

- **unigram** (*bool*) – Use uni-gram instead of using bi-grams
- **weight** (*float*) – adjusts the “importance” of this identifier in the Bloom filter. Can be set to zero to skip
- **kwargs** – Extra keyword arguments passed to the tokenizer

Note: For each n-gram of an identifier, we compute k different indices in the Bloom filter which will be set to true. There is a global $k_{default}$ value, and the k value for each identifier is computed as

$$k = weight * k_{default},$$

rounded to the nearest integer.

Reasons why you might want to set weights:

- Long identifiers like street name will produce a lot more n-grams than small identifiers like zip code. Thus street name will flip more bits in the Bloom filter and will have a bigger influence in the overall matching score.

- The matching might produce better results if identifiers that are stable and / or have low error rates are given higher prominence in the Bloom filter.
-

`__weakref__`

list of weak references to the object (if defined)

key derivation

```
class clkhash.key_derivation.HKDFconfig(master_secret, salt=None, info=None,
                                         hash_algo='SHA256')
```

Bases: `object`

```
static check_is_bytes (value)
```

```
static check_is_bytes_or_none (value)
```

```
supported_hash_algos = ('SHA256', 'SHA512')
```

```
clkhash.key_derivation.generate_key_lists(master_secrets, num_identifier, key_size=64,
                                           salt=None, info=None, kdf='HKDF')
```

Generates a derived key for each identifier for each master secret using a key derivation function (KDF).

The only supported key derivation function for now is 'HKDF'.

The previous key usage can be reproduced by setting `kdf` to 'legacy'. This is highly discouraged, as this strategy will map the same n-grams in different identifier to the same bits in the Bloom filter and thus does not lead to good results.

Parameters

- **master_secrets** – a list of master secrets (either as bytes or strings)
- **num_identifier** – the number of identifiers
- **key_size** – the size of the derived keys
- **salt** – salt for the KDF as bytes
- **info** – optional context and application specific information as bytes
- **kdf** – the key derivation function algorithm to use

Returns The derived keys. First dimension is of size `num_identifier`, second dimension is the same as `master_secrets`. A key is represented as bytes.

```
clkhash.key_derivation.hkdf(hkdf_config, num_keys, key_size=64)
```

Executes the HKDF key derivation function as described in rfc5869 to derive `num_keys` keys of size `key_size` from the `master_secret`.

Parameters

- **hkdf_config** – an `HKDFconfig` object containing the configuration for the HKDF.
- **num_keys** – the number of keys the kdf should produce
- **key_size** – the size of the produced keys

Returns Derived keys

random names

Module to produce a dataset of names, genders and dates of birth and manipulate that list

Currently very simple and not realistic. Additional functions for manipulating the list of names - producing reordered and subset lists with a specific overlap

ClassList class - generate a list of length n of [id, name, dob, gender] lists

TODO: Get age distribution right by using a mortality table TODO: Get first name distributions right by using distributions TODO: Generate realistic errors TODO: Add RESTfull api to generate reasonable name data as requested

```
class clkhash.randomnames.NameList(n)
```

Bases: `object`

List of randomly generated names

```
generate_random_person(n)
```

Generator that yields details on a person with plausible name, sex and age.

Yields Generated data for one person tuple - (id: int, name: str('First Last'), birthdate: str('DD/MM/YYYY'), sex: str('M' | 'F'))

```
generate_subsets(sz, overlap=0.8)
```

Return a pair of random subsets of the name list with a specified proportion of elements in common.

Parameters

- **sz** – length of subsets to generate
- **overlap** – fraction of the subsets that should have the same names in them

Raises `ValueError` – if there aren't sufficiently many names in the list to satisfy the request; more precisely, raises if $sz + (1 - \text{overlap}) * sz > n = \text{len}(\text{self.names})$

Returns pair of subsets

```
load_names()
```

This function loads a name database into globals firstNames and lastNames

initial version uses data files from <http://www.quietaffiliate.com/free-first-name-and-last-name-databases-csv-and-sql/>

```
schema = [{'identifier': 'INDEX'}, {'identifier': 'NAME freetext'}, {'identifier':
```

```
schema_types
```

```
clkhash.randomnames.load_csv_data(resource_name)
```

Loads a specified CSV data file and returns the first column as a Python list

```
clkhash.randomnames.random_date(start, end)
```

This function will return a random datetime between two datetime objects.

Parameters

- **start** – datetime of start
- **end** – datetime of end

Returns random datetime between start and end

```
clkhash.randomnames.save_csv(data, schema, file)
```

Output generated data to file as CSV with header.

Parameters

- **data** – An iterable of tuples containing raw data.
- **schema** – Iterable of schema definition dicts
- **file** – A writeable stream in which to write the CSV

schema

```
clkh.hash.schema.get_schema_types(schema)
```

```
clkh.hash.schema.load_schema(schema_file)
```

tokenizer

Functions to tokenize words (PII)

```
clkh.hash.tokenizer.bigramlist(word, toremove=None)
```

Make bigrams from word with pre- and ap-pended spaces

s -> [' ' + s0, s0 + s1, s1 + s2, .. sN + ' ']

Parameters

- **word** – string to make bigrams from
- **toremove** – List of strings to remove before construction

Returns list of bigrams as strings

```
clkh.hash.tokenizer.positional_unigrams(instr)
```

Make positional unigrams from a word.

E.g. 1987 -> ["1 1", "2 9", "3 8", "4 7"]

Parameters **instr** – input string

Returns list of strings with unigrams

```
clkh.hash.tokenizer.unigramlist(instr, toremove=None, positional=False)
```

Make 1-grams (unigrams) from a word, possibly excluding particular substrings

Parameters

- **instr** – input string
- **toremove** – Iterable of strings to remove

Returns list of strings with unigrams

1.4.2 Testing

Make sure you have all the required modules before running the tests (modules that are only needed for tests are not included during installation):

```
$ pip install -r requirements.txt
```

Now run the unit tests and print out code coverage with *py.test*:

```
$ python -m pytest --cov=clkh.hash
```

Note several tests will be skipped by default. To enable the command line tests set the *INCLUDE_CLI* environment variable. To enable the tests which interact with an entity service set the *TEST_ENTITY_SERVICE* environment variable to the target service's address:

```
$ TEST_ENTITY_SERVICE= INCLUDE_CLI= python -m pytest --cov=clkhsh
```

1.4.3 Type Checking

clkhsh uses static typechecking with mypy. To run the type checker (in Python 3.5 or later):

```
$ pip install mypy
$ mypy clkhsh --ignore-missing-imports --strict-optional --no-implicit-optional --
  ↳disallow-untyped-calls
```

1.5 References

CHAPTER 2

External Links

- [clckhash on Github](#)
- [clckhash on PyPi](#)

CHAPTER 3

Indices and tables

- `genindex`
- `modindex`

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