

A Small Example

David Wu

2024-10-23

Here, we will clean and process a small database into a temporal network of patient transfers.

All the functionality is provided in the `hospinet` library.

```
import hospinet

import polars as pl
import networkx as nx

from matplotlib import pyplot as plt # for later
```

We load the database (in the form of a csv), and clean it using the `hospinet.cleaner` submodule. We specify the column names, since they are different from the (standardised) default.

```
source_db = hospinet.cleaner.ingest_csv("./data/admissions.csv", convert_dates=True)
source_db.head(5)
```

patient i64	hospital str	admission datetime[s]	discharge datetime[s]
3	"D"	2020-04-02 02:56:19.917759	2021-06-04 14:58:15.715780
2	"E"	2020-04-10 13:56:21.043894	2021-06-07 14:35:45.036138
3	"B"	2020-04-15 03:22:06.938845	2020-05-27 10:14:20.476332
2	"A"	2020-04-20 20:33:17.858555	2020-05-23 10:31:03.186415
1	"E"	2020-04-26 03:36:11.245264	2021-01-08 13:36:08.182081

```
cleaned_db = hospinet.cleaner.clean_database(
    source_db,
    delete_missing="record",
    delete_errors="record",
```

```

    subject_id = 'patient',
    facility_id = 'hospital',
    admission_date = 'admission',
    discharge_date = 'discharge',
    retain_auxiliary_data = True,
)
# encode dates to numerics
first_date = cleaned_db.select(pl.col('Adate').min()).item()
numeric_db = hospinet.cleaner.normalise_dates(
    cleaned_db,
    cols = ['Adate', 'Ddate'],
    ref_date = first_date
)

```

```

INFO::hospinet::Checking existence of columns...
INFO::hospinet::Column existence OK.
INFO::hospinet::Standardising column names...
INFO::hospinet::Coercing types...
INFO::hospinet::Type coercion done.
INFO::hospinet::Checking for missing values...
INFO::hospinet::Checking for erroneous records...
INFO::hospinet::Removing duplicate records...
INFO::hospinet::Finding and fixing overlapping records...
INFO::hospinet::Attempting up to 100 iterations
INFO::hospinet::Iteration 0: 144 entries; 92 overlaps; 0.0026714359992183745 s
INFO::hospinet::Iteration 1: 171 entries; 62 overlaps; 0.0022785820183344185 s
INFO::hospinet::Iteration 2: 194 entries; 50 overlaps; 0.0033057639957405627 s
INFO::hospinet::Iteration 3: 215 entries; 44 overlaps; 0.002744395984336734 s
INFO::hospinet::Iteration 4: 234 entries; 38 overlaps; 0.0023581869900226593 s
INFO::hospinet::Iteration 5: 250 entries; 34 overlaps; 0.002557872037868947 s
INFO::hospinet::Iteration 6: 265 entries; 32 overlaps; 0.002452725952025503 s
INFO::hospinet::Iteration 7: 276 entries; 28 overlaps; 0.0028382480377331376 s
INFO::hospinet::Iteration 8: 286 entries; 21 overlaps; 0.002110183995682746 s
INFO::hospinet::Iteration 9: 294 entries; 18 overlaps; 0.002611069008708 s
INFO::hospinet::Iteration 10: 302 entries; 16 overlaps; 0.002957379969302565 s
INFO::hospinet::Iteration 11: 310 entries; 16 overlaps; 0.002604701032396406 s
INFO::hospinet::Iteration 12: 317 entries; 15 overlaps; 0.0023700519814155996 s
INFO::hospinet::Iteration 13: 324 entries; 14 overlaps; 0.0022482519852928817 s
INFO::hospinet::Iteration 14: 330 entries; 13 overlaps; 0.0024886580067686737 s
INFO::hospinet::Iteration 15: 335 entries; 12 overlaps; 0.0023688190267421305 s
INFO::hospinet::Iteration 16: 339 entries; 9 overlaps; 0.0027188339736312628 s
INFO::hospinet::Iteration 17: 243 entries; 7 overlaps; 0.0024568139924667776 s

```

```

INFO::hospinet::Iteration 18: 188 entries; 5 overlaps; 0.0030460780253633857 s
INFO::hospinet::Iteration 19: 120 entries; 4 overlaps; 0.002818793000187725 s
INFO::hospinet::Iteration 20: 122 entries; 4 overlaps; 0.00281722896033898 s
INFO::hospinet::Iteration 21: 123 entries; 3 overlaps; 0.0025224960409104824 s
INFO::hospinet::Iteration 22: 124 entries; 2 overlaps; 0.0029321539914235473 s
INFO::hospinet::Iteration 23: 125 entries; 2 overlaps; 0.002912027994170785 s
INFO::hospinet::Iteration 24: 126 entries; 2 overlaps; 0.00280821998603642 s
INFO::hospinet::Iteration 25: 127 entries; 2 overlaps; 0.0027175010181963444 s
INFO::hospinet::Iteration 26: 128 entries; 2 overlaps; 0.00297291501192376 s
INFO::hospinet::Iteration 27: 129 entries; 2 overlaps; 0.0025346739566884935 s
INFO::hospinet::Iteration 28: 130 entries; 2 overlaps; 0.0030280210194177926 s
INFO::hospinet::Iteration 29: 131 entries; 2 overlaps; 0.002856035018339753 s
INFO::hospinet::Iteration 30: 132 entries; 2 overlaps; 0.002567083982285112 s
INFO::hospinet::Iteration 31: 133 entries; 2 overlaps; 0.002165095997042954 s
INFO::hospinet::Iteration 32: 134 entries; 2 overlaps; 0.0023509410093538463 s
INFO::hospinet::Iteration 33: 135 entries; 2 overlaps; 0.0025193950277753174 s
INFO::hospinet::Iteration 34: 135 entries; 1 overlaps; 0.0026860389625653625 s
INFO::hospinet::Iteration 35: 0 entries; 0 overlaps; 0.002382597012910992 s
INFO::hospinet::History of non-overlapping patient records:
INFO::hospinet::[0, 0...{16x}, 99, 57, 70, 0...{15x}, 135]
INFO::hospinet::0 overlaps remaining after iterations

```

We can then process this into a `TemporalNetwork` object directly using the `from_presence` class method

```

network = hospinet.TemporalNetwork.from_presence(
    numeric_db,
    discretisation=1,
)

```

This object is also a `networkx.DiGraph` object, so we can use the native plotting functionality. We project this down to static nodes first, so that we don't get the full temporal graph.

```

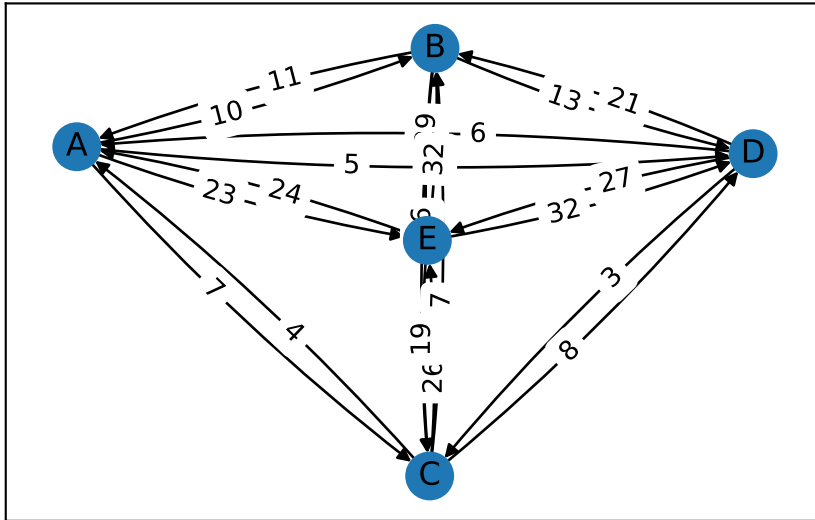
static_network = network.to_static()
pos = nx.spring_layout(static_network, seed=1451)
edge_labels = {
    tuple(edge): f"{attr}"
    for *edge, attr in static_network.edges(data='weight')
}
nx.draw_networkx(static_network, pos=pos, connectionstyle='arc3,rad=0.05')
nx.draw_networkx_edge_labels(
    static_network,

```

```

pos=pos,
edge_labels=edge_labels,
label_pos=0.4,
connectionstyle='arc3,rad=0.05'
);

```



We provide some basic indexing support via methods:

```

print("Nodes with presence at time 15: ", network.nodes_at_time(15))
print("When hospital D is occupied: ", network.when_present('D'))

```

Nodes with presence at time 15: ['E' 'B']

When hospital D is occupied: [0 1 2 3 4 5 6 7 8 9 10 11 12 13 38]

42	43	44	45	46	47	48	49	50	51	52	53	54	55	56	57	58	59
71	72	73	74	75	76	77	78	79	80	81	82	83	84	85	86	87	88
89	90	91	92	93	94	95	96	97	98	99	100	101	102	103	113	114	115
116	117	118	119	120	121	122	123	124	125	126	127	128	129	130	131	132	133
134	135	136	137	138	139	140	141	142	143	144	145	146	147	148	149	150	151
152	153	154	155	156	157	158	159	160	175	181	182	183	184	185	186	187	188
189	190	191	192	193	194	195	196	197	198	199	200	201	202	203	204	205	206
207	208	209	210	211	212	213	214	215	216	217	218	219	220	221	222	223	224
225	226	227	228	229	230	231	232	233	234	235	236	237	238	239	240	241	242
243	244	245	246	247	248	249	250	251	252	253	254	255	256	257	258	259	260
261	262	263	264	265	266	267	268	269	270	271	272	278	292	293	294	295	296
297	298	299	300	301	302	303	304	305	306	307	308	309	310	311	312	313	314

315 316 317 318 319 320 321 322 323 324 325 326 327 328 329 330 345 346
347 348 349 350 351 352 353 355 356 357 358 359 376 377 378 379 380 381
382 383 384 385 386 387 388 389 390 391 392 393 394 395 396 397 398 399
400 401 402 403 404 405 406 407 408 409 410 411 412 413 414 415 416 417
418 419 420 421 422 423 424 425 426 427 428 429 430 431 432 433 434 435
436 437 438 439 440 441 442 443 444 445]