### Odus C

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```
[1]: # %load_ext autoreload # %autoreload 2
```

### 1 Installation

You need to have python 3.7 to run this notebook.

You'll also find that you need some packages. You'll find that out when you run into an ImportError. Usually, you can figure google the missing package and find out how to install it. Usually it's just running pip install THE\_PACKAGE\_YOU\_WANT in the terminal.

But that only works for "pypi" published packages. Some of the following (my) packages are not published yet.

Here's how to install them (assuming you have python 3.7, pip, and git):

In your python (3.7) environment...

For py2store you can just do: pip install git+https://github.com/i2mint/py2store

For ut and hyp you'll have to do a bit more:

Make a projects folder somewhere. Let's say you have it here: ~/py/proj.

Now go to that folder:

cd ~/py/proj

Run pwd and copy the full path of the proj folder somewhere warm.

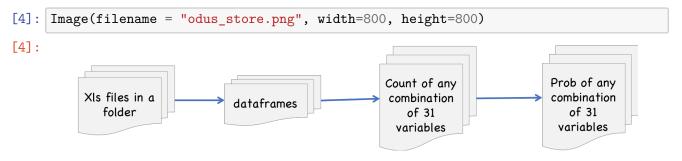
Now do this:

```
git clone https://github.com/thorwhalen/ut
Then
git clone https://github.com/thorwhalen/hyp
Then
```

git clone https://github.com/thorwhalen/odus

Now add that project folder you saved somewhere to your PYTHONPATH. How? LMGTFY:

https://stackoverflow.com/questions/3402168/permanently-add-a-directory-to-pythonpath



## 2 Getting some resources

```
[1]: from matplotlib.pylab import *
  from numpy import *
  import seaborn as sns

import os
  from py2store.stores.local_store import RelativePathFormatStore
  from py2store.mixins import ReadOnlyMixin
  from py2store.base import Store
  # from ut.util.imports.ipython_utils import *
  # from ut.util.imports.data_analysis import *
  from odus.analysis_utils import *

from io import BytesIO
  from hyp.ppi.pot import Pot, ProbPot
  from collections import UserDict, Counter
  import numpy as np
  import pandas as pd
```

```
from ut.ml.feature_extraction.sequential_var_sets import PVar, VarSet, DfData, □
→VarSetFactory
from IPython.display import Image

# get odus at https://github.com/thorwhalen/odus
# Depends on https://github.com/thorwhalen/hyp too.
from odus.dacc import DfStore, counts_of_kps, Dacc, plot_life_course, □
→VarSetCountsStore, \
    mk_pvar_struct, PotStore, _commun_columns_of_dfs, Struct, □
→mk_pvar_str_struct, \
    VarStr
```

```
[2]: from odus import data_dir, data_path_of survey_dir = data_dir data_dir
```

[2]: '/D/Dropbox/dev/p3/proj/odus/odus/data'

```
[3]: df_store = DfStore(data_dir + '/{}.xlsx')
len(df_store)
cstore = VarSetCountsStore(df_store)
v = mk_pvar_struct(df_store, only_for_cols_in_all_dfs=True)
s = mk_pvar_struct(v)
f, df = cstore.df_store.head()
pstore = PotStore(df_store)
```

# 3 Poking around

### 3.1 df store

A df\_store is a key-value store where the key is the xls file and the value is the prepared dataframe

```
[4]: len(df_store)
```

[4]: 119

```
[5]: it = iter(df_store.values())
for i in range(5): # skip five first
    _ = next(it)
df = next(it) # get the one I want
df.head(3)
```

```
[5]: category RURAL SUBURBAN URBAN/CITY HOMELESS INCARCERATION WORK \
age
11 0 1 0 0 0 0 0
12 0 1 0 0 0 0
```

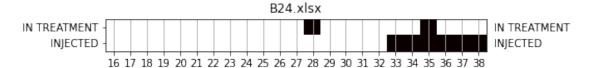
	13	0	1	0 0		0 0	
	category	SON/DAUGHTER	SIBLING	FATHER/MOTHER	SPOUSE	METHAMPHE	TAMINE \
	age				•••		
	11	1	1	0	0		0
	12	1	1	0	0		0
	13	1	1	0	0		0
	category age	AS PRESCRIBED	OPIOID :	NOT AS PRESCRIB	ED OPIOID	HEROIN \	
	11		0		0	0	
	12		1		0	0	
	13		0		0	0	
	category age	OTHER OPIOID	INJECTED	IN TREATMENT	Selects St	ates below	Georgia \
	11	0	0	0		1	1
	12	0	0			1	1
	13	0	0	0		1	1
		Pennsylvania					
	age	•					
	11	0					
	12 13	0					
	13	U					
	[3 rows x	31 columns]					
[6]:	[]: [print(df.columns.values)						
	['RURAL' 'SUBURBAN' 'URBAN/CITY' 'HOMELESS' 'INCARCERATION' 'WORK' 'SON/DAUGHTER' 'SIBLING' 'FATHER/MOTHER' 'SPOUSE' 'OTHER (WHO?, FILL IN BRACKETS HERE)' 'FRIEND USER' 'FRIEND NON USER' 'MENTAL ILLNESS' 'PHYSICAL ILLNESS' 'LOSS OF LOVED ONE' 'TOBACCO' 'MARIJUANA' 'ALCOHOL' 'HAL/LSD/XTC/CLUBDRUG' 'COCAINE/CRACK' 'METHAMPHETAMINE' 'AS PRESCRIBED OPIOID' 'NOT AS PRESCRIBED OPIOID' 'HEROIN' 'OTHER OPIOID' 'INJECTED' 'IN TREATMENT' 'Selects States below' 'Georgia' 'Pennsylvania']						
[7]:	t = df[['.t.head(3)	ALCOHOL', 'TOB	ACCO']]				
[7]:	category age	ALCOHOL TOBA	CCO				
	11	0	0				
	12	0	0				
	13	0	0				

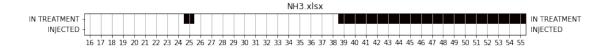
```
[8]: c = Counter()
      for i, r in t.iterrows():
          c.update([tuple(r.to_list())])
      С
 [8]: Counter({(0, 0): 6, (1, 0): 4, (1, 1): 9, (0, 1): 2})
 [9]: def count_tuples(dataframe):
          c = Counter()
          for i, r in dataframe.iterrows():
              c.update([tuple(r.to_list())])
          return c
[10]: fields = ['ALCOHOL', 'TOBACCO']
      # do it for every one
      c = Counter()
      for df in df_store.values():
          c.update(count_tuples(df[fields]))
      С
[10]: Counter({(0, 1): 903, (1, 1): 1343, (0, 0): 240, (1, 0): 179})
[11]: pd.Series(c)
[11]: 0 1
               903
      1 1
              1343
      0 0
               240
               179
      1 0
      dtype: int64
[12]: # Powerful! You can use that with several pairs and get some nice probabilities.
      → Look up Naive Bayes.
     3.2 Making a pdf of trajectories
[13]: import itertools
      from functools import partial
      from ut.pimg.utils import write_images
      ihead = lambda it: itertools.islice(it, 0, 5)
[14]: fields = [s.in_treatment, s.injected]
      def plot_life(df, fields=None, title=None, ax=None):
          if fields is None:
              fields = slice(None, None)
```

plot\_life\_course(df[fields], ax=ax);

```
plt.grid(which='both', axis='x')
if title is not None:
    plt.title(title)
    plt.gca().xaxis.set_tick_params(labeltop=False, labelbottom=True)

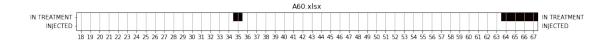
axs = list()
for k, df in itertools.islice(df_store.items(), 0, 5):
    plot_life(df, fields, title=k.split('/')[1])
    axs.append(plt.gca())
```











```
[15]: def write_trajectories_to_file(df_store, fields=None, fp='test.pdf', □ 

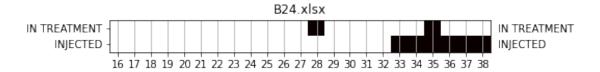
→pil_write_format=None, 

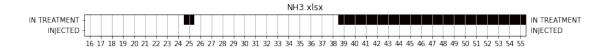
to_pil_image_kwargs=None, **pil_save_kwargs):
```

```
[16]: fig_gen = map(lambda k, df: plot_life(df, fields, title=k.split('/')[1]),__

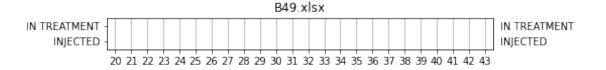
-*(zip(*ihead(df_store.items()))))
len(list(fig_gen))
```

### [16]: 5









```
A60.xlsx

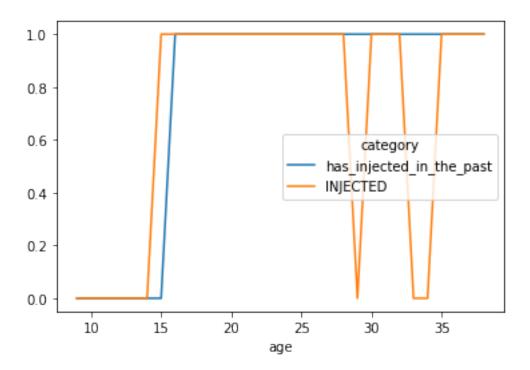
IN TREATMENT
INJECTED

18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64 66 66 67
```

```
[]:
[17]: # write_trajectories_to_file(df_store, fp='test.pdf')
[18]: np.hstack(([False], array(cumsum(df[s.injected]) > 0)[:-1]))
[18]: array([False, False, False, False, False, False, False, False, False,
            False, False, False, False, False, False, False, False, False,
            False, False, False, False, False, False, False, False,
            False, False, False, False, False, False, False, False, False,
            False, False, False, False, False, False, False, False,
            False, False, False, False])
[19]: k, df = next(itertools.islice(df_store.items(), 7, 100))
      print(k)
      df['has_injected_in_the_past'] = np.hstack(([False], array(cumsum(df[s.
      \rightarrowinjected]) > 0)[:-1])).astype(int)
      # plot(df[s.injected], 'r-')
      # plot('bo')
      df[['has_injected_in_the_past', s.injected]].plot()
```

surveys/A21.xlsx

[19]: <matplotlib.axes.\_subplots.AxesSubplot at 0x12b71f4f0>



```
3.3 Demo s and v

[21]: print(list(filter(lambda x: not x.startswith('__'), dir(s))))

['alcohol', 'as_prescribed_opioid', 'cocaine_crack', 'father_mother',
    'hal_lsd_xtc_clubdrug', 'heroin', 'homeless', 'in_treatment', 'incarceration',
    'injected', 'loss_of_loved_one', 'marijuana', 'mental_illness',
    'methamphetamine', 'not_as_prescribed_opioid', 'other_opioid',
    'physical_illness', 'rural', 'sibling', 'son_daughter', 'suburban', 'tobacco',
    'urban_city', 'work']

[22]: s.heroin

[22]: 'HEROIN'

[23]: v.heroin

[24]: PVar('HEROIN', 0)
```

#### 3.4 cstore

```
[25]: cstore[v.alcohol, v.tobacco]
[25]: Counter({(0, 1): 903, (1, 1): 1343, (0, 0): 240, (1, 0): 179})
[26]: cstore[v.alcohol, v.tobacco, v.heroin]
[26]: Counter({(0, 0, 1): 427,
               (1, 0, 1): 656,
               (1, 1, 1): 687,
               (0, 0, 0): 189,
               (0, 1, 1): 476,
               (0, 1, 0): 51,
               (1, 0, 0): 133,
               (1, 1, 0): 46
[27]: cstore[v.alcohol-1, v.alcohol]
[27]: Counter({(0, 0): 994, (1, 1): 1375, (1, 0): 90, (0, 1): 87})
[28]: cstore[v.alcohol-1, v.alcohol, v.tobacco]
[28]: Counter({(0, 0, 1): 807,
               (1, 1, 1): 1220,
               (1, 0, 0): 26,
               (0, 1, 1): 76,
               (0, 0, 0): 187,
               (1, 1, 0): 155,
               (0, 1, 0): 11,
               (1, 0, 1): 64)
     3.5 pstore
[29]: t = pstore[s.alcohol-1, s.alcohol]
      t
[29]:
                         pval
      ALCOHOL-1 ALCOHOL
      0
                0
                          994
                1
                           87
                0
      1
                           90
                1
                         1375
[30]: t.tb
```

```
[30]: ALCOHOL-1 ALCOHOL pval
               0
                       0 994
               0
                       1
                            87
               1
                       0 90
               1
                       1 1375
[31]: t / []
[31]:
                          pval
     ALCOHOL-1 ALCOHOL
               0
                       0.390416
               1
                       0.034171
     1
               0
                       0.035350
               1
                       0.540063
[32]: t / t[s.alcohol-1]
[32]:
                          pval
     ALCOHOL-1 ALCOHOL
              0
                       0.919519
               1
                       0.080481
               0
     1
                       0.061433
               1
                      0.938567
[33]: tt = pstore[s.alcohol, s.tobacco]
[33]:
                     pval
     ALCOHOL TOBACCO
                      240
             0
             1
                      903
            0
     1
                      179
             1
                     1343
[34]: tt / tt[s.alcohol]
[34]:
                        pval
     ALCOHOL TOBACCO
            0
                     0.209974
             1
                     0.790026
           0
     1
                     0.117608
            1
                     0.882392
[35]: tt / tt[s.tobacco]
[35]:
                         pval
     ALCOHOL TOBACCO
```

```
0
              0
                       0.572792
      1
              0
                       0.427208
              1
      0
                       0.402048
      1
              1
                       0.597952
 []:
 []:
 []:
     4 Potential Calculus Experimentations
[36]: # survey_dir = '/D/Dropbox/others/Miriam/python/ProcessedSurveys'
      df_store = DfStore(survey_dir + '/{}.xlsx')
      len(df_store)
[36]: 119
[37]: cstore = VarSetCountsStore(df store)
      v = mk_pvar_struct(df_store, only_for_cols_in_all_dfs=True)
      s = mk_pvar_str_struct(v)
      f, df = cstore.df_store.head()
      df.head(3)
[37]: category RURAL SUBURBAN URBAN/CITY HOMELESS INCARCERATION
                                                                      WORK
      age
      16
                    0
                              1
                                          0
                                                    0
                                                                   1
                                                                         0
      17
                                          0
                                                    0
                                                                   0
                    0
                              1
                                                                         1
      18
                    0
                              1
                                                                   0
                                                                         1
      category SON/DAUGHTER SIBLING FATHER/MOTHER SPOUSE
      age
      16
                           1
                                    1
                                                   1
                                                           0
      17
                           1
                                    1
                                                   1
                                                           0
      18
                           1
      category HAL/LSD/XTC/CLUBDRUG COCAINE/CRACK METHAMPHETAMINE \
      age
      16
                                   0
                                                  0
                                                                   0
      17
                                   0
                                                  0
                                                                   0
                                                  0
                                                                   0
      18
      category AS PRESCRIBED OPIOID NOT AS PRESCRIBED OPIOID HEROIN \
      age
      16
                                   0
                                                             0
                                                                     0
```

```
18
                                    0
                                                               1
      category OTHER OPIOID INJECTED IN TREATMENT Massachusetts
      age
      16
                           0
                                      0
                                                    0
                                                                    1
                                      0
      17
                           0
                                                    0
                                                                    1
      18
                           0
                                      0
                                                    0
                                                                    1
      [3 rows x 29 columns]
[38]: cstore = VarSetCountsStore(df_store)
      cstore.mk_pvar_attrs()
[39]: from odus.dacc import DfStore, counts_of_kps, Dacc, plot_life_course,_
      →VarSetCountsStore, mk_pvar_struct, PotStore
      pstore = PotStore(df_store)
      pstore.mk_pvar_attrs()
      p = pstore[v.homeless - 1, v.incarceration]
      p
[39]:
                                pval
     HOMELESS-1 INCARCERATION
                 0
                                 1690
                                  577
                 1
      1
                 0
                                  192
                 1
                                   87
[40]: p / []
[40]:
                                     pval
     HOMELESS-1 INCARCERATION
                 0
                                 0.663786
                                 0.226630
                 1
      1
                 0
                                 0.075412
                 1
                                 0.034171
[41]: pstore[v.incarceration]
[41]:
                     pval
      INCARCERATION
      0
                     1989
                      676
      1
[42]: pstore[v.alcohol-1, v.loss_of_loved_one]
```

```
[42]:
                                    pval
      ALCOHOL-1 LOSS OF LOVED ONE
                0
                                     990
                1
                                      91
      1
                0
                                    1321
                1
                                     144
[43]: tw = pstore[v.tobacco, v.work]
      mw = pstore[v.marijuana, v.work]
      aw = pstore[v.alcohol, v.work]
      w = pstore[v.work]
[44]: evid_t = Pot.from_hard_evidence(**{s.tobacco: 1})
      evid_m = Pot.from_hard_evidence(**{s.marijuana: 1})
      evid_a = Pot.from_hard_evidence(**{s.alcohol: 1})
      evid_a
[44]:
               pval
      ALCOHOL
      1
                  1
[45]: aw
[45]:
                    pval
      ALCOHOL WORK
              0
                     431
                     712
              1
      1
              0
                     448
              1
                    1074
[46]:
     w / []
[46]:
                pval
      WORK
      0
            0.329831
      1
            0.670169
[47]: (evid_m * mw) / []
[47]:
                           pval
      MARIJUANA WORK
                0
                       0.350603
                      0.649397
                1
[48]: (evid_t * tw) / []
```

```
[48]:
                pval
    TOBACCO WORK
    1 0
                0.313001
                0.686999
           1
[49]: (evid_a * aw) / []
                 pval
[49]:
    ALCOHOL WORK
           0
                0.29435
                0.70565
           1
[]:
[]:
```