CoMap evaluation

October 2, 2015

```
In [1]: %matplotlib inline
        import os
        from collections import OrderedDict
        import pandas as pd
        import matplotlib as mpl
        import matplotlib.pyplot as plt
        import seaborn as sns
        import yaml
        from pathlib import Path
        from data import CodesInDbs, Mappings, Databases
        from IPython.display import Latex
        import normalize
       pd.set_option('display.max_colwidth', 100)
        sns.set_style('whitegrid')
        #sns.set_context("poster")
        #plt.rcParams['figure.figsize'] = (4, 3)
        plt.rc("savefig", dpi=150)
        measures_palette = sns.color_palette('Set1', n_colors=2, desat=.5)
        measures_palette.reverse()
        def graded_recall_palette(n_colors):
            palette = sns.color_palette("Blues", n_colors=n_colors, desat=.6)
            palette.reverse()
            return palette
        def graded_precision_palette(n_colors):
            palette = sns.color_palette("Reds", n_colors=n_colors, desat=.6)
            palette.reverse()
            return palette
        def mystyle(palette=None, xrot=0, ha='center', ylim=(0,1), ylabel=None, savefig=None):
            class C:
                def __enter__(self):
                    if palette is not None:
                        palette.__enter__()
                def __exit__(self, exc_type, value, traceback):
                    if palette is not None:
                        palette.__exit__(exc_type, value, traceback)
                    if exc_type is None:
                        sns.despine(left=True)
```

```
plt.grid(False, axis='x')
                        if plt.gca().legend_:
                            plt.legend(loc=2, bbox_to_anchor=(1, 1))
                        plt.gca().get_lines()[0].set_visible(False)
                        plt.gca().set_ylim(*ylim)
                        plt.xticks(rotation=xrot, ha=ha)
                        if ylabel is not None:
                            plt.ylabel(ylabel)
                        if savefig:
                            plt.savefig('{}-{}'.format(PROJECT, savefig))
            return C()
        pd.set_option('display.notebook_repr_html', True)
        def _repr_latex_(self):
            #return r"\begin{center}%s\end{center}" %
            return self.to_latex()
        pd.DataFrame._repr_latex_ = _repr_latex_ # monkey patch pandas DataFrame
        PROJECT = os.getenv('COMAP_PROJECT')
        print("PROJECT:", PROJECT)
PROJECT: safeguard
In [2]: with open('../projects/{}/variations.yaml'.format(PROJECT)) as f:
            variations = yaml.load(f)
        with open('../projects/{}/config.yaml'.format(PROJECT)) as f:
            config = yaml.load(f)
            databases = Databases.of_config(config)
            coding_systems = config['coding-systems']
        with open('../projects/{}/events.yaml'.format(PROJECT)) as f:
            events = yaml.load(f)
            event_names = {}
            for event in events:
                casedef = yaml.load(open('../projects/{}/case-definitions/{}.yaml'.format(PROJECT, even
                event_names[event] = casedef['name']
        with open('.../projects/{}/mappings.yaml'.format(PROJECT)) as f:
            mappings = Mappings.of_raw_data(yaml.load(f), events, databases)
            mappings = normalize.mappings(mappings, databases)
        with open('../codes-in-dbs.json') as f:
            codes_in_dbs = CodesInDbs.of_data(json.load(f))
        with open('.../{}.code-stats.csv'.format(PROJECT)) as f:
            code_stats = pd.read_csv(f)
        def database_label(database):
            return database
            #return "{} ({})".format(database, databases.coding_system(database))
        def measure_label(measure):
           return {
                "recall": "Sensitivity",
```

```
"precision": "PPV", # "Positive predictive value",
            }[measure]
        def event_label(event):
            return event_names[event]
In [3]: df0 = pd.read_csv('.../{}.evaluations.csv'.format(PROJECT))
        df = df0['variation event database recall precision'.split()]
        df.head()
Out[3]:
```

	variation	event	database	recall	precision
0	max-recall	is	IPCI	1.000000	1.000000
1	max-recall	is	GePaRD	0.909091	0.666667
2	max-recall	is	CPRD	0.850000	0.850000
3	max-recall	is	Medicare	1.000000	0.833333
4	\max -recall	ap	IPCI	1.000000	1.000000

Notes 1

Should exclusion codes from the reference be generated? No. Exclusion codes are often added database specifically, where the codes are *not* represented in the case definition.

2 Coding systems

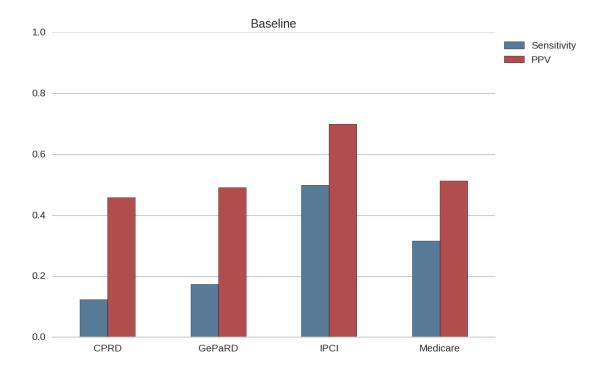
```
In [4]: pd.DataFrame([
            (database, databases.coding_system(database))
            for database in databases.databases()
        ], columns=("Database", "Coding system")).set_index("Database")
```

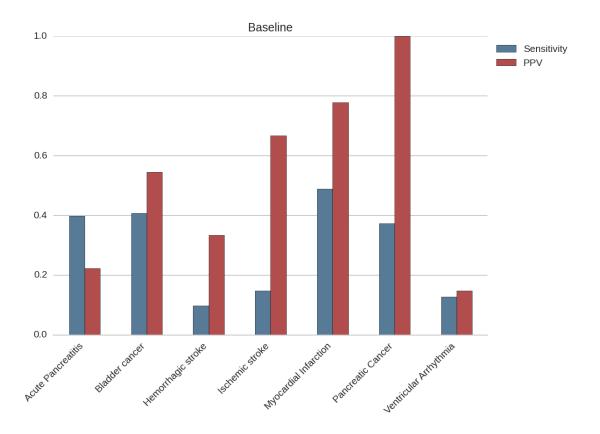
Out[4]:

	Coding system
Database	
Medicare	ICD9CM
IPCI	ICPC2EENG
CPRD	RCD2
GePaRD	ICD10CM

Baseline 3

```
In [5]: averages_compare = pd.DataFrame([
            df[df.variation == 'baseline'].groupby('database').recall.mean(),
            df[df.variation == 'baseline'].groupby('database').precision.mean(),
       ])
        averages_compare.index = averages_compare.index.map(measure_label)
       averages_compare.columns = averages_compare.columns.map(database_label)
        with mystyle(measures_palette, savefig='baseline-performance-by-db.pdf'):
            averages_compare.T.plot(kind='bar', title='Baseline')
```

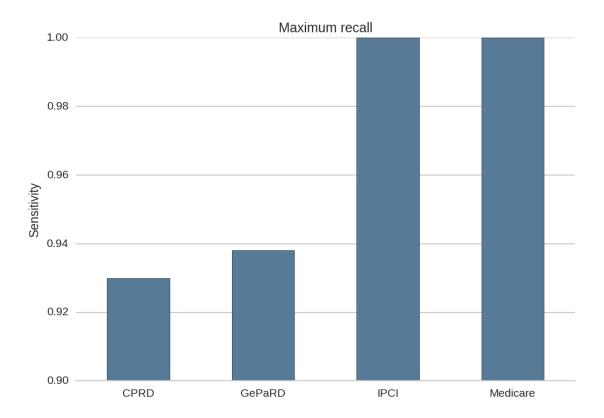




4 Max-recall

Out[7]:

	CPRD	GePaRD	IPCI	Medicare
Sensitivity	0.929871	0.938033	1	1



4.1 Reasons for imperfect sensitivity

Out[8]:

	In mapping	Not in maximum recall	%
Database			
CPRD	229	14	6.11%
GePaRD	72	3	4.17%
IPCI	7	0	0.00%
Medicare	50	0	0.00%

```
max_recall_fn = max_recall_fn.groupby('database').fn.sum().to_frame()
max_recall_fn.fn = max_recall_fn.fn.map(set).map(', '.join)
max_recall_fn.index.name = 'Database'
max_recall_fn.columns = ['False negatives of maximum recall']
max_recall_fn
```

Out[9]:

False negatives of maximum recall

Database

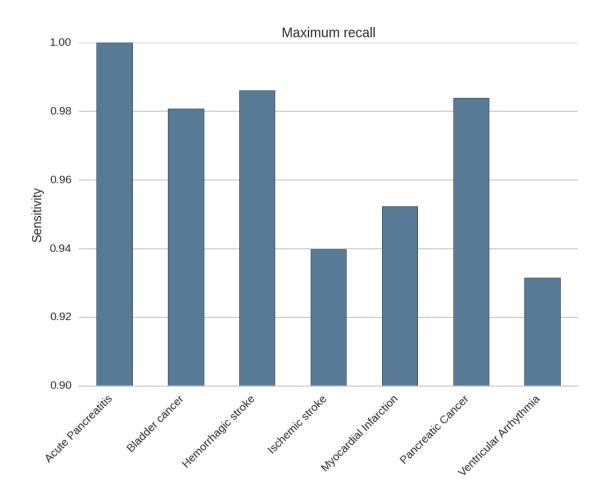
CPRD G60X., BBd9., G61X., G6W.., 7L1H7, G6X.., ByuE0, G64z4, BBaz., ByuE., BBa.., 100.., 7L1H6, BA0z. GePaRD I46.0, I64, I21.9

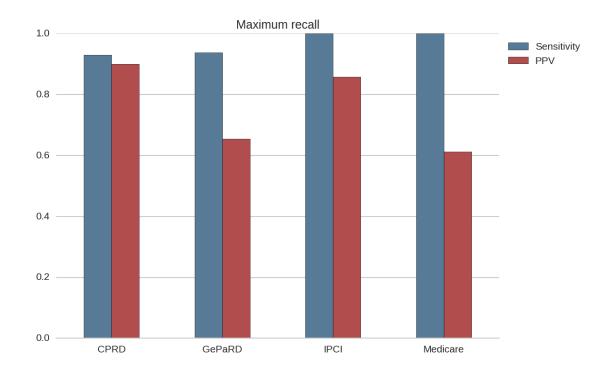
CPRD: READ2 codes from the reference are mapped to READ CTV3 codes that are not in UMLS, for example 7L1H6 (READ2) -> XaM3E, XaPuP, 7L1H6, 7L1h6.

GePaRD: Only 3 codes are missing, but those FN have large influence on sensitivity in mappings with few codes.

Out[10]:

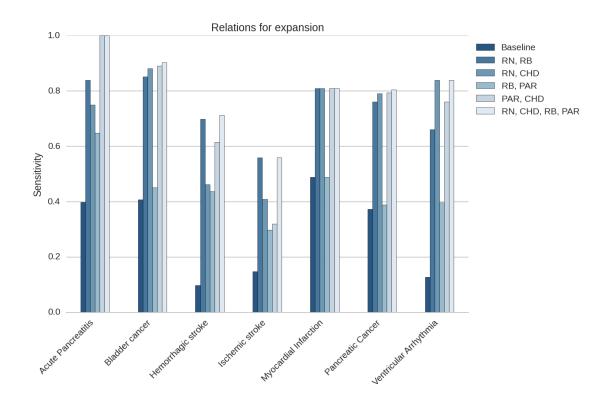
	Sensitivity
Acute Pancreatitis	1.000000
Bladder cancer	0.980769
Hemorrhagic stroke	0.986111
Ischemic stroke	0.939773
Myocardial Infarction	0.952381
Pancreatic Cancer	0.983945
Ventricular Arrhythmia	0.931481



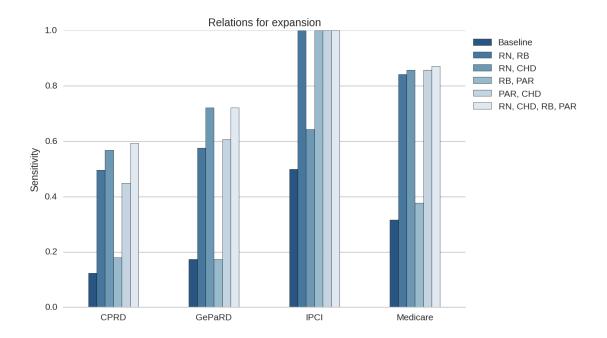


5 Compare relations for expansion

```
In [12]: compare_variations = OrderedDict([
             ('baseline', 'Baseline'),
             ('1-RN-RB.expand', 'RN, RB'),
             ('1-RN-CHD.expand', 'RN, CHD'),
             ('1-RB-PAR.expand', 'RB, PAR'),
             ('1-PAR-CHD.expand', 'PAR, CHD'),
             ('1-RN-CHD-RB-PAR.expand', 'RN, CHD, RB, PAR'),
         averages_compare = pd.DataFrame([
             df[df.variation == variation].groupby('event').recall.mean()
             for variation in compare_variations
         ], index=compare_variations)
         averages_compare.columns = averages_compare.columns.map(event_names.get)
         averages_compare.index = compare_variations.values()
         with mystyle(graded_recall_palette(len(compare_variations)), xrot=45, ha='right', savefig='rel
             averages_compare.T.plot(kind='bar', title="Relations for expansion")
             plt.ylabel(measure_label('recall'))
```

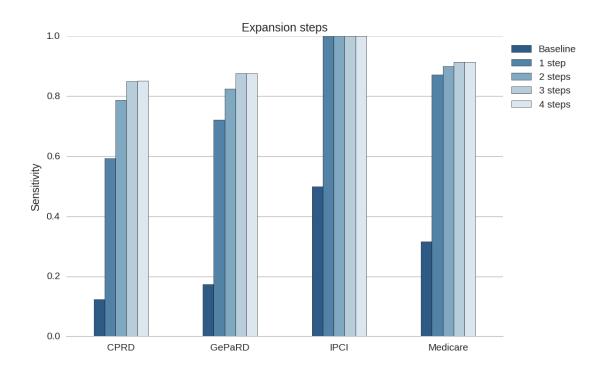


```
In [13]: compare_variations = OrderedDict([
             ('baseline', 'Baseline'),
             ('1-RN-RB.expand', 'RN, RB'),
             ('1-RN-CHD.expand', 'RN, CHD'),
             ('1-RB-PAR.expand', 'RB, PAR'),
             ('1-PAR-CHD.expand', 'PAR, CHD'),
             ('1-RN-CHD-RB-PAR.expand', 'RN, CHD, RB, PAR'),
         ])
         averages_compare = pd.DataFrame([
             df[df.variation == variation].groupby('database').recall.mean()
             for variation in compare_variations
         ], index=compare_variations)
         averages_compare.columns = averages_compare.columns.map(database_label)
         averages_compare.index = compare_variations.values()
         with mystyle(graded_recall_palette(len(compare_variations)), savefig='relations-recall-by-db.p
             averages_compare.T.plot(kind='bar', title="Relations for expansion")
             plt.ylabel(measure_label('recall'))
```



6 Increasing sensitivity with more expansion steps

```
In [14]: compare_variations = OrderedDict([
             ('baseline', 'Baseline'),
             ('1-RN-CHD-RB-PAR.expand', '1 step'),
             ('2-RN-CHD-RB-PAR.expand', '2 steps'),
             ('3-RN-CHD-RB-PAR.expand', '3 steps'),
             ('4-RN-CHD-RB-PAR.expand', '4 steps'),
         ])
         averages_compare = pd.DataFrame([
             df[df.variation == variation].groupby('database').recall.mean()
             for variation in compare_variations
         ], index=compare_variations)
         averages_compare.columns = averages_compare.columns.map(database_label)
         averages_compare.index = compare_variations.values()
         with mystyle(graded_recall_palette(len(compare_variations)), savefig='steps-recall-by-db.pdf')
             averages_compare.T.plot(kind='bar', title="Expansion steps")
             plt.ylabel(measure_label('recall'))
```

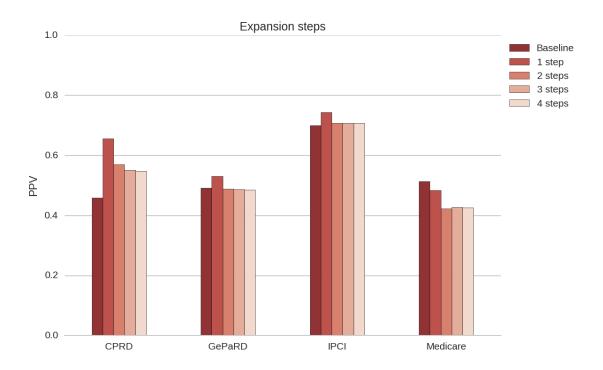


6.1 Reasons for low performance in IPCI when including exclusion codes

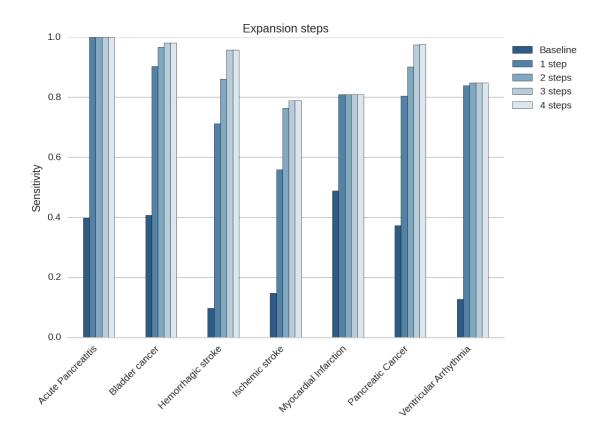
Exclusion codes are not in the evaluation any more. See note above.

The IPCI mapping contains very broad codes that are refined with additional terms. For example

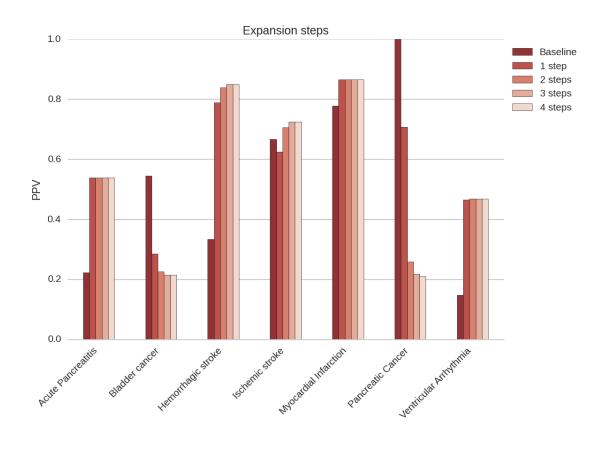
- K24 (Fear of heart attack)
- K90 (stroke)
- K93 (Pulmonary embolism)
- D70 (Dementia) OR "dementia" AND "infarct"
- U14 (Kidney symptom/complaint) OR "nier" AND "infarct"

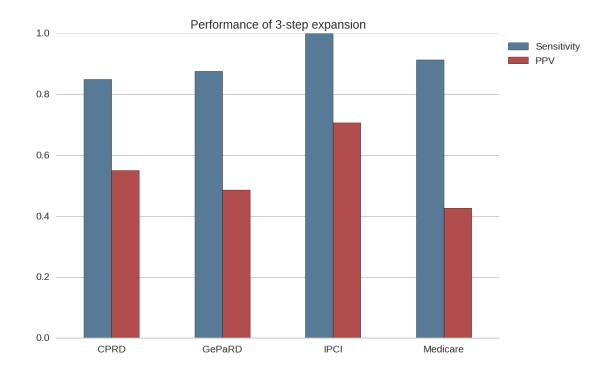


```
In [16]: compare_variations = OrderedDict([
             ('baseline', 'Baseline'),
             ('1-RN-CHD-RB-PAR.expand', '1 step'),
             ('2-RN-CHD-RB-PAR.expand', '2 steps'),
             ('3-RN-CHD-RB-PAR.expand', '3 steps'),
             ('4-RN-CHD-RB-PAR.expand', '4 steps'),
        ])
         averages_compare = pd.DataFrame([
             df[df.variation == variation].groupby('event').recall.mean()
             for variation in compare_variations
         ], index=compare_variations)
         averages_compare.columns = averages_compare.columns.map(event_names.get)
         averages_compare.index = compare_variations.values()
         with mystyle(graded_recall_palette(len(compare_variations)), xrot=45, ha='right', savefig='steps
             averages_compare.T.plot(kind='bar', title="Expansion steps")
             plt.ylabel(measure_label('recall'))
```

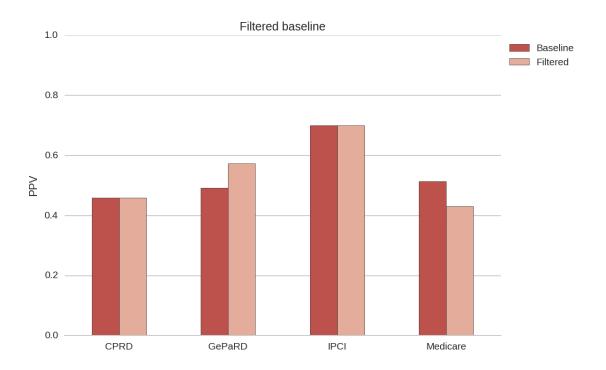


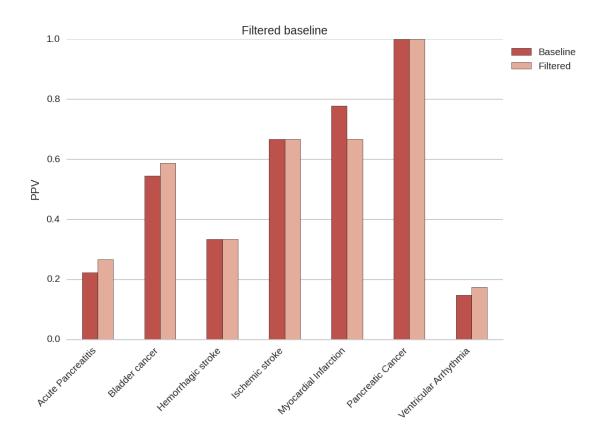
```
In [17]: compare_variations = OrderedDict([
             ('baseline', 'Baseline'),
             ('1-RN-CHD-RB-PAR.expand', '1 step'),
             ('2-RN-CHD-RB-PAR.expand', '2 steps'),
             ('3-RN-CHD-RB-PAR.expand', '3 steps'),
             ('4-RN-CHD-RB-PAR.expand', '4 steps'),
         ])
         averages_compare = pd.DataFrame([
             df[df.variation == variation].groupby('event').precision.mean()
             for variation in compare_variations
         ], index=compare_variations)
         averages_compare.columns = averages_compare.columns.map(event_names.get)
         averages_compare.index = compare_variations.values()
         with mystyle(graded_precision_palette(len(compare_variations)), xrot=45, ha='right', savefig='
             averages_compare.T.plot(kind='bar', title="Expansion steps")
             plt.ylabel(measure_label('precision'))
```

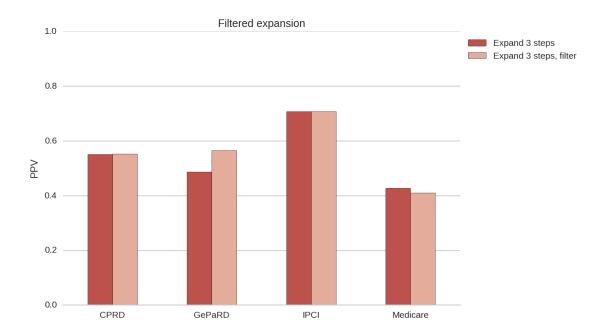


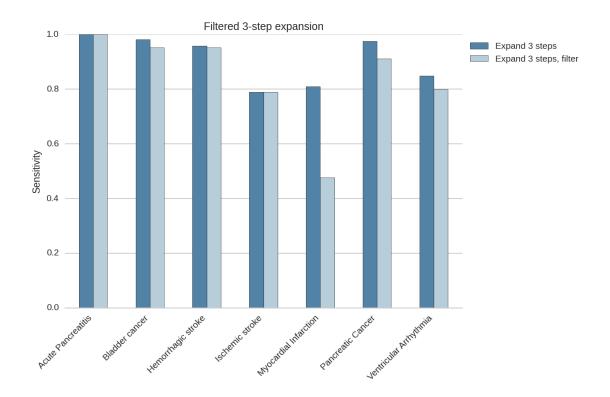


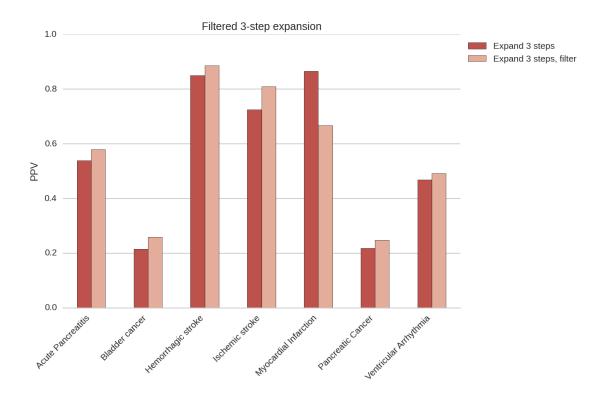
7 Removing unused codes

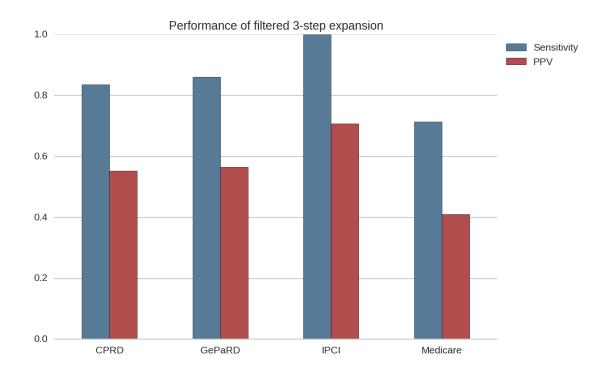












The drop in PPV for Myocardial infarction is caused by the mapping to codes 410.* (Acute myocardial infarction) in Medicare which is *not* used in the ARS database.

8 Codes in reference mappings, not in databases

Codes that might be removed from the TP when filtering.

Out[25]:

	In ref	Not in DB	%	Codes
Database				
CPRD	229	5	2.18%	Gyu60, ByuF., 8531., B91zz, BB46.
GePaRD	72	1	1.39%	C25
IPCI	7	0	0.00%	NaN
Medicare	50	14	28.00%	$410.6,\ 410.9,\ 410.4,\ 410.5,\ 410.1,\ 188,\ 410.3,\ 410.7,\ 410.0,\ 157,\ 410,\ 410.2,\ 410.1,\ 41$

In [26]: