

Chicago Tribute Policy Lab

Appeals Analysis

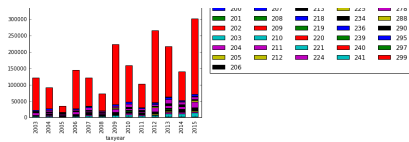
At the beginning of the quarter, I added a `clean_name` column for others to use. Here is the code for that:

Other useful functions

UNAMCO O pin		taxyear	appealnum	attorneysarep	prior_ar	proposse_ar	assr_final	board_ar	housenum	area	subarea	ass win	bor win	bor rev	ass rev	total rev	classification	win	clean_name	
0	0	10110102000000	203	164	KEARNEY & PHELAN, LTD.	59618	59618	59618	48260	1177	1	101	0	1	11358	0	11358	517	1	KEARNEY&PHELAN, LTD
1	1	10110200010000	203	142	REFF & SCHRAMM	114862	114862	114862	11182	105	1	101	0	0	0	0	0	0	0	REFF&SCHRAMM
2	2	10110200000000	203	142	REFF & SCHRAMM	124517	124517	124517	124517	207	1	101	0	0	0	0	0	0	0	REFF&SCHRAMM
3	3	10110000000000	203	145	EUGENE L. GRIFFIN & ASSOCIATES	492099	871480	863336	490927	257	1	101	1	1	71239	308144	379393	591	1	EUGENE.GRIFFIN&ASSOCIATES
4	4	00000000000000	206	28	MCCRACKEN,WALSH,DELAVERN&HETLER	321777	600000	494139	386419	301	1	101	1	0	0	222621	222621	92	1	MCCRACKEN,WALSH,DELAVERN&HETLER

Appeals per year





Below we can compute according to TRI, given tax data has been merged in:

```
In [8]:

df.groupby(['taxyear', 'TRI']).size().unstack().plot(kind = "bar", stacked=True)
plt.tight_layout()

Out[8]:

Index([('Unnamed: 0', 'u'pin', 'u'taxyear', 'u'appelnum', 'u'attorneytaxrep',
       'u'pinir_av', 'u'propose_av', 'u'ass_final', 'u'board_av', 'u'housewma',
       'u'dir', 'u'street_name', 'u'street_suffix', 'u'city_name', 'u'zip',
       'u'area',
       'u'ubarea', 'u'ass_win', 'u'bor_win', 'u'bor_rev', 'u'ass_rev',
       'u'total_rev', 'u'classification', 'u'win', 'u'clean_name', 'u'h_rev',
       'u'case_type'),
      dtype=object)

In [ ]:

df.groupby(['taxyear_x', 'TRI']).mean()[['h_rev']].unstack().plot(kind = "bar", stacked=True)
fig = plt.legend(bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0.)
#plt.tight_layout()

In [ ]:

df_2.groupby(['taxyear', 'condo']).size().groupby(level=0).apply(lambda x: 100*float(x.sum())/x.sum()).unstack()

In [ ]:

def remove_dup_merge(df1_duplicates, df2, duplicate_cols, match_cols):
    """ use with sales_df, appeals_df, ['pin'], ['pin'] """
    df1_duplicates = df1_duplicates.drop_duplicates(duplicate_cols)
    merged = df1.merge(df2_duplicates, on = match_cols, how = "outer")
    return merged

def add_class_type_col(df, classif_column):
    """
    in appeals data, classification column = classification
    in sales data, classification column = bor_class
    """
    df['class_type'] = df[classif_column].apply(lambda x: str(x)[0])

def keep_2_5_only(df):
    final = df[(df['class_type'] == "2") | (df['class_type'] == "5")]
    return final

def analysis_year(year):
    # year ==
    path = "/Users/Dan/Dropbox/Muni_Finance_Lab/Raw_Data/Appals"
    appeals = pd.read_csv(path + "/Appals.csv")
    year_index_col = None, header=0)
    path = "/Users/Dan/Dropbox/Muni_Finance_Lab/Raw_Data/Sales_Ratios/lat_pasa"
    sales = pd.read_csv(path + "/results.csv")
    year

    df = remove_dup_merge(sales, appeals, ['pin'], ['pin'])

    ## houses that SOLD but didn't appeal - no appealsnum
    a= df[pd.isnull(df['appealnum'])].shape

    ## pinid is only in sales data, that APPEALED ONLY - no pinid
    b = df[pd.isnull(df['pinid'])].shape

    df['matched'] = np.where((pd.notnull(df['pi_saiclass'])) & (pd.notnull(df['appealnum'])), 1, 0)

    ## sold only
    df['sold'] = np.where(pd.isnull(df['appealnum']), 1, 0)

    print 'median netconsideration appealed ', df[df['matched'] == 1]['NetConsideration'].median()
    print 'median netconsideration sold ', df[df['sold'] == 1]['NetConsideration'].median()
    print 'median pi_valratio appealed ', df[df['matched'] == 1]['pi_valratio'].median()
    print 'median pi_valratio sold ', df[df['sold'] == 1]['pi_valratio'].median()

    #houses who won have a bit higher sales ratio
    print 'median netcons for lost / won ', df[df['matched'] == 1].groupby(['win'])['NetConsideration'].median()

    # those who won have a higher netconsideration
    print 'median pi_valratio for lost / won ', df[df['matched'] == 1].groupby(['win'])['pi_valratio'].median()

    # those who won have a higher propose_av
    print 'count for lost / won ', df[df['matched'] == 1].groupby(['win'])['NetConsideration'].count()

    return df
```

Given the functions above, we can run a general analysis for all years

just call the analysis_year function with a two digit year

```
In [ ]:

df15 = analysis_year(15)
```

Code for the appeals-tax-sale-community data merge is in the data_merge.py file