Possible partial demolitions

This analysis is based on files downloaded and filtered in the <code>01_Download_process</code> notebook in this repo. It is construction permits for "Addition" or "Remodel or Addition" that have the term "partial demo" in the description, definitions we've learned from interviews with City of Austin officials. They are only residential permits for single-family or duplex homes.

It's *IMPORTANT* to note that not all of these permits are demolitions that lead to an increase in the size or shape of the home. In some cases, this might be simply a bathroom or kitchen remodel.

We did a lot of investigation into trying to find "single-wall teardowns", defined as homes where all but one wall was torn down, typically to preserve a historical setback zoning rule. There was not way we could find to find permits for that type of demolition vs any other, with any accuracy.

```
In [1]: import pandas as pd
        import altair as alt
        pd.options.display.max colwidth = 60
In [2]: # Column type fixes
        column types = {
             'ApplicantPhone': pd.np.str,
             'ContractorPhone': pd.np.str,
             'CalendarYearIssued': pd.np.str,
            'OriginalZip': pd.np.str,
              'IssuedDate': pd.np.str.
        }
        # import raw data
        data_raw = pd.read_csv(
             '../data-processed/demolitions_partial.csv',
            index col=None,
            dtype=column types,
            parse_dates=['IssuedDate']
        )
        demolitions = data raw
        demolitions.shape
Out[2]: (5060, 68)
```

Total cases

```
In [3]: print("Partial demolition permits that _could be_ additions: {}".format(len(de molitions)))
```

Partial demolition permits that _could be_ additions: 5060

```
In [4]: # create data from of permits by year
demos_by_year = demolitions.CalendarYearIssued.value_counts().reset_index()
demos_by_year.columns = ['Year', 'Count']
demos_by_year.sort_values('Year')
```

Out[4]:

	Year	Count
10	2008	154
9	2009	255
8	2010	283
6	2011	404
5	2012	430
4	2013	494
3	2014	532
1	2015	709
0	2016	757
2	2017	681
7	2018	361

```
In [5]: # make chart based on dataframe above
alt.Chart(
          demos_by_year.reset_index(),
          title='Possible "Additions" demolitions each year'
).mark_bar().encode(
          x=alt.X("Year:0", axis=alt.Axis(title="Year", labelAngle=0)),
          y=alt.Y("Count:Q", axis=alt.Axis(title="Permits")),
).properties(width=600)
```

Out[5]:

Possible "Additions" demolitions each year 100-

Year

Partial demos by zip code

```
In [6]: # make a datafram with counts of permits by zip code
    demos_by_zip = demolitions.OriginalZip.value_counts().reset_index()
    demos_by_zip.columns = ['Zip', 'Count']
    demos_by_zip.head()
```

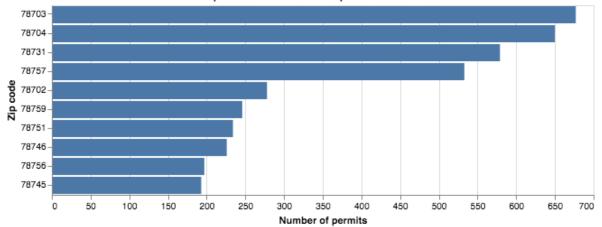
Out[6]:

	Zip	Count
0	78703	677
1	78704	650
2	78731	579
3	78757	533
4	78702	278

```
In [7]: # chart based on dataframe above
        alt.Chart(
            demos_by_zip.head(10),
            title="Zip codes with the most possible additions"
        ).mark_bar().encode(
            x=alt.X("Count:Q",
                     axis=alt.Axis(title="Number of permits")),
            y=alt.Y("Zip:0",
                     # this is the sorting magic
                     sort=alt.SortField(
                         field="Count",
                         order="descending",
                         op="sum"
                     ),
                     axis=alt.Axis(title="Zip code")),
        ).properties(width=600)
```

Out[7]:

Zip codes with the most possible additions



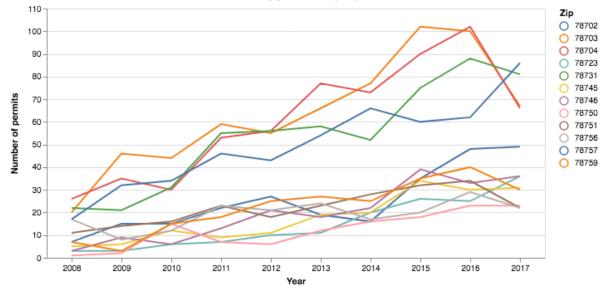
Possible additions by zip by year

This chart is interesting in that it shows the hottest zip codes (03 and 04) had drops in 2017, falling out of the lede for the first time since 2012. Not sure it is worth finding out why at this point, but... it's interesting.

```
In [8]: # Filter out 2018 because it is a partial year
        before 2018 = demolitions[demolitions['CalendarYearIssued'] != '2018']
        # build a dataframe of yr, zip, count
        demos_yr_zip = before_2018.groupby(['CalendarYearIssued','OriginalZip']).agg(d
        ict(
            PermitNum='count'
        )).reset index()
        demos_yr_zip.columns = ['Year', 'Zip', 'Count']
        demos_yr_zip.shape
        # demos yr zip.head()
        # filter to 2017
        zips 2017 = demos yr zip[demos yr zip['Year'] == '2017']
        # filter the 2017 list to those with 20+ demos
        # sort the list by count descending
        zips_2017_top = zips_2017[zips_2017.Count > 19].sort_values('Count', ascending
        =False)
        # Use the Zip code column above to filter the original demos_yr_zip datatable
        # and create the new dataframe to chart
        top_demos_yr_zip_new = demos_yr_zip[demos_yr_zip['Zip'].isin(zips_2017_top.Zip
        )]
        alt.Chart(
            top demos yr zip new,
            title="Demolitions by year and top Zip codes"
        ).mark line().encode(
            x=alt.X("Year:0",
                     axis=alt.Axis(title="Year", labelAngle=0)),
            y=alt.Y("Count:Q",
                     axis=alt.Axis(title="Number of permits")),
            color="Zip"
        ).properties(width=600)
```

Out[8]:

Demolitions by year and top Zip codes



```
In [ ]:
```