# Sample

#### October 12, 2021

```
[1]: import pandas as pd
  from copy import deepcopy
  import matplotlib.pyplot as plt
  import re
  import numpy as np
  from matplotlib import gridspec
  import matplotlib
```

## 1 Helper functions

These are borrowed from the Convert.ipynb file.

```
[3]: df = pd.read_excel('../Dataset/dataset.xlsx',header=3,index_col=1)
df = df.drop('Unnamed: 0',axis=1).T#.reset_index().rename({'Building_□}

→ Identifier': 'index', 'index': 'Building Identifier'},axis=1)
df = df[df.index.str.contains('0')]
```

```
[4]: df[[c for c in df.columns if 'kg' in c]] = df[[c for c in df.columns if 'kg' in_\[ \dots c]].astype('float')
```

```
[5]: df = pd.concat([df[headings].groupby(lambda x: x.split('.')[0],axis=0).

→max(),df[[c for c in df.columns if 'kg' in c]].groupby(lambda x: x.split('.

→')[0],axis=0).mean(numeric_only=True)],axis=1)
```

```
[6]: name_conversion = pd.read_csv('name_conversion.csv')
building_name_conversion = pd.read_csv('building_type_name_conversion.csv')
```

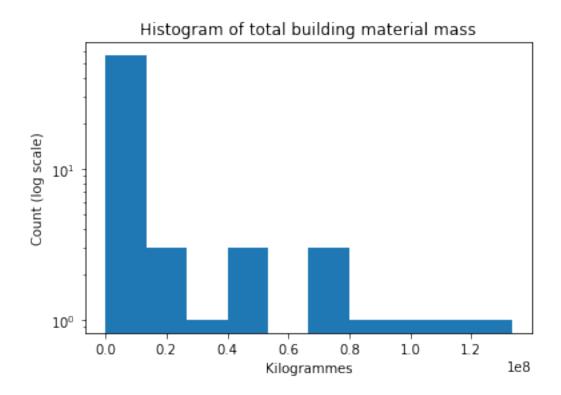
```
[7]: building_name_map = {k['Building Code']:k['Building Type'] for _,k in_
→building_name_conversion.iterrows()}
```

```
[8]: name_map = {k.Code:k.Category for _,k in name_conversion.iterrows()}
[9]: additional categories map = {v:k for k,v in {
         'Continuous Footings':'OCF',
         'Foundation Walls':'OFW',
         'Spread Footings':'OSF',
         'Column Piers':'OCP',
         'Columns Supporting Floors':'CSF',
         'Floor Girders and Beams': 'FGB',
         'Floor Trusses':'OFT',
         'Floor Joists':'OFJ',
         'Columns Supporting Roofs':'CSR',
         'Roof Girders and Beams': 'RGB',
         'Roof Trusses':'ORT',
         'Roof Joists':'ORJ',
         'Parking Bumpers':'OPB',
         'Precast Concrete Stair Treads': 'PCS',
         'Roof Curbs':'ORC',
         'Exterior Wall Construction': 'EWC',
         'Composite Decking':'CPD',
         'Cast-in-Place concrete':'CIC',
         'Floor Structural Frame': 'FSF',
         'Associated Metal Fabrications':'AMF',
         'Floor Construction Supplementary Components': 'FCS',
         'Roof Construction Supplementary Components':'RCS',
         'Residential Elevators':'ORE',
         'Vegetated Low-Slope Roofing':'VLR',
         'Swimming Pools':'SWP',
         'Excavation Soil Anchors': 'ESA',
         'Floor Trusses':'FTS',
         'Roof Window and Skylight Performance': 'RWS',
         'Rainwater Storage Tanks': 'RST',
         'Gray Water Tanks': 'GWT'}.items()
     }
     additional_categories_map['OFT'] = 'Floor Trusses'
```

# 2 1. Plot sample figures

Here we plot building material mass.

```
[10]: plt.hist(df[[c for c in df.columns if 'kg' in c]].sum(axis=1));
    plt.title('Histogram of total building material mass')
    plt.yscale('log')
    plt.xlabel('Kilogrammes')
    plt.ylabel('Count (log scale)');
```



## 3 2. Investigate a specific material

In this example, we select only columns that match the MasterFormat code for Concrete. Then, we aggregate based on Level 2 UniFormat code.

```
[11]: cols = [d for d in df.columns if ('_03' in d or '_04 22' in d) and not '_03 20'__
       \hookrightarrowin d]
[12]: f = lambda x: re.split('[_\.\]',x)[1][0:3]
      concrete_df = pd.concat([df[headings],df[cols].groupby(f,axis=1).sum()],axis=1).
       →rename(columns=name_map)
[13]:
      concrete_df
[13]: Building Identifier Country City Quality / Stage of Data
                                                                   Construction Date
      001
                                CA TOR
                                                                                 2021
                                                            OOIFC
      002
                                                                                 2021
                                CA
                                    TOR
                                                            OOIFC
      003
                                CA
                                    TOR
                                                            OOIFC
                                                                                 2021
      004
                                                            00IFC
                                                                                 2021
                                CA
                                    TOR
      005
                                CA
                                    TOR
                                                            OOIFC
                                                                                 2011
      066
                                CA
                                    TOR
                                                            OOIFT
                                                                                 2020
      067
                                CA
                                    TOR
                                                            OOIFC
                                                                                 2019
```

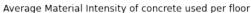
068 069 070	CA CA CA	TOR TOR TOR	0	IFBP OIFC OIFC	2021 2020 2021
Building Identifier 001 002 003 004 005	Building	Type Gross SND SND SND SND OFF MIX	521.18 389.24 411.64 269.56 11248.00	Foundations \ 1.710150e+05 1.082862e+05 1.911912e+05 6.739916e+04 1.278753e+06 1.566786e+07	
067 068 069 070		LNW LNW LNW LNW		2.657254e+04 2.721844e+04 2.846246e+04 3.930037e+03	
Building Identifier 001 002 003 004 005	Subgrade	0.000 0.000 0.000 0.000 0.000 1027239.110	6.751475 3.578757 3.254672 1.618022	e+04 e+04 e+04 e+04	
066 067 068 069 070		3484448.795 0.000 0.000 0.000 0.000	2.924485 2.206696 1.264324	e+04 e+04 e+04	
Building Identifier 001 002 003 004 005 066 067 068 069	Substruc	0.000000e+ 0.000000e+ 0.000000e+ 0.000000e+ 7.359709e+ 1.368861e+ 0.000000e+ 0.000000e+	00 00 00 00 05 07 00 00	0.0 0.0 0.0 0.0 0.0  0.0 0.0	
070 Building Identifier 001 002 003	Substruc	0.000000e+		1.949675e+03 1.409585e+03	\

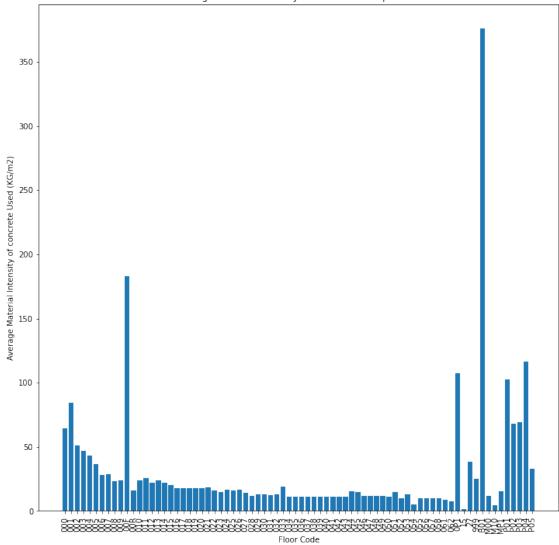
```
004
                                                    0.0
                                                            2.269760e+01
005
                                                    0.0
                                                            7.126901e+06
. .
066
                                                    0.0
                                                            5.737703e+07
067
                                                    0.0
                                                            0.000000e+00
068
                                                    0.0
                                                            0.000000e+00
069
                                                    0.0
                                                            0.000000e+00
070
                                                    0.0
                                                            0.000000e+00
Building Identifier Exterior Vertical Enclosures \
001
                                                0.00
002
                                                0.00
003
                                                0.00
004
                                                0.00
005
                                           311760.72
                                           71331.23
066
067
                                                0.00
                                                0.00
068
069
                                                0.00
070
                                                0.00
Building Identifier Exterior Horizontal Enclosures
                                                       Interior Construction \
001
                                                   0.0
                                                                  0.000000e+00
002
                                                   0.0
                                                                  0.000000e+00
003
                                                   0.0
                                                                  0.000000e+00
004
                                                   0.0
                                                                  0.000000e+00
005
                                                 552.0
                                                                  1.175564e+06
. .
                                                   0.0
066
                                                                  1.463901e+07
067
                                                   0.0
                                                                  0.000000e+00
068
                                                   0.0
                                                                  0.000000e+00
069
                                                   0.0
                                                                  0.000000e+00
070
                                                   0.0
                                                                  0.000000e+00
Building Identifier
                        Conveying Plumbing Special Construction \
001
                            0.000
                                         0.0
                                                               0.000
002
                            0.000
                                                               0.000
                                         0.0
003
                                         0.0
                                                               0.000
                            0.000
004
                            0.000
                                         0.0
                                                               0.000
005
                            0.000
                                         0.0
                                                               0.000
. .
                      8273703.915
                                         0.0
066
                                                         711760.625
067
                            0.000
                                         0.0
                                                               0.000
068
                                         0.0
                                                               0.000
                            0.000
069
                            0.000
                                         0.0
                                                               0.000
070
                            0.000
                                         0.0
                                                               0.000
```

```
Building Identifier Site Improvements
001
                                  0.0000
002
                                  0.0000
003
                                  0.0000
004
                                  0.0000
005
                             169830.9495
. .
                                  0.0000
066
067
                                  0.0000
068
                                  0.0000
069
                                  0.0000
070
                                  0.0000
```

[70 rows x 20 columns]

```
[14]: grouping_function = lambda x: x.split('_')[0] #This function takes in a full_\( \to column name, like "000_G2010.20.000_03 00 00.00_m3_1", and returns only the_\( \to floor. \)
to_draw = df[cols].groupby(grouping_function,axis=1).sum().replace(0,np.NaN).
\( \to div(df['Gross Floor Area'],axis='rows').mean() \)
plt.figure(figsize=(12,12))
plt.bar(to_draw.keys(), to_draw.values)
plt.xticks(rotation=90)
plt.title('Average Material Intensity of concrete used per floor')
plt.ylabel('Average Material Intensity of concrete Used (KG/m2)')
plt.xlabel('Floor Code');
```





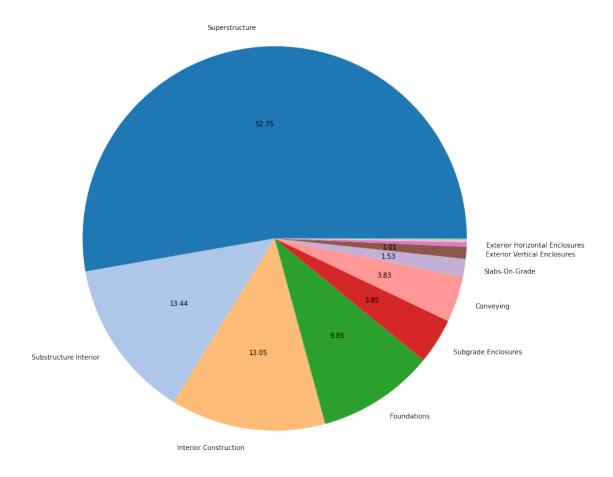
Now, we will aggregate to Level 3 MasterFormat codes, and display these values for the first three entries.

- [15]: f = lambda x: name\_map[re.split('[\_\.\]',x)[1][0:3]] #This function takes in a\_\( \) \( \rightarrow full column name and returns only the Level 3 MasterFormat code. \) \( \)
- [16]: concrete\_df.mean().sort\_values(ascending=False)

```
Foundations
                                   1.277363e+06
Subgrade Enclosures
                                   4.997662e+05
Conveying
                                   4.970189e+05
Slabs-On-Grade
                                   1.989609e+05
Exterior Vertical Enclosures
                                   1.306903e+05
Exterior Horizontal Enclosures
                                   5.030072e+04
Special Construction
                                   1.543692e+04
Substructure Related Activities
                                   1.208292e+04
Site Improvements
                                   5.666442e+03
Plumbing
                                   5.186825e+03
Water And Gas Mitigation
                                   1.219826e+03
dtype: float64
```

## 3.1 Pie chart version A: on-pie chart labels for all > 1%

```
[17]: def my_autopct(pct):
        return ('%.2f' % pct) if pct > 1 else ''
        to_plot = concrete_df.mean().sort_values(ascending=False)
        to_plot.plot.pie(figsize=(12,12),colormap='tab20',autopct=my_autopct,labels=[k_\sum_if v > 35000 else '' for k,v in to_plot.items()])
        plt.ylabel('')
        plt.title('Percentage of total concrete used in each building element_\sum_category');
        # plt.legend(loc='center left',bbox_to_anchor=(-0.20, 0.75));
        plt.tight_layout();
```



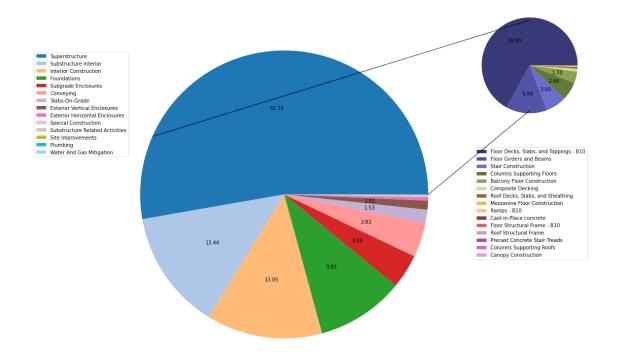
## 3.2 Pie version B: external legend with slice labels

```
fig = plt.figure(figsize=(16,12))
gs = gridspec.GridSpec(2, 2, width_ratios=[3, 1])
ax0 = plt.subplot(gs[:,0])

def my_autopct(pct):
    return ('%.2f' % pct) if pct > 1 else ''
to_plot = concrete_df.mean().sort_values(ascending=False)
to_plot.plot.pie(ax=ax0,colormap='tab20',autopct=my_autopct,labeldistance=None)
plt.ylabel('')
plt.legend(loc='center left',bbox_to_anchor=(-0.20, 0.75));
plt.tight_layout();

ax1 = plt.subplot(gs[0,1])
f = lambda x: \
```

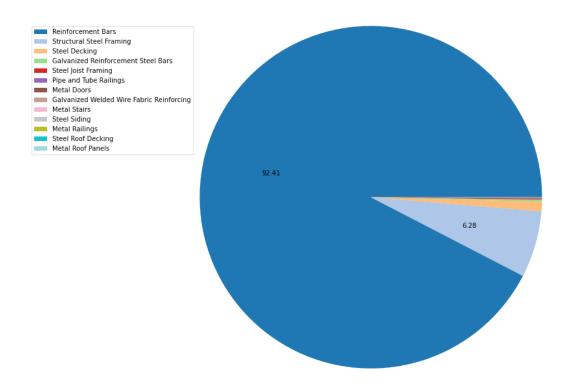
```
additional_categories_map[re.split('[_\.\]',x)[3]] \
   if \
   re.split('[_\.\]',x)[3] != '000' \
   else \
   name_map['.'.join(re.split('[_\.\]',x)[1:3])]
superstructure_df = df[[c for c in cols if 'B10' in c]].groupby(f,axis=1).sum()
to_plot = superstructure_df.mean().sort_values(ascending=False)
def my autopct(pct):
   return ('%.2f' % ((pct * 0.4335))) if pct > 1 else ''
to_plot.plot.pie(ax=ax1,colormap='tab20b',autopct=my_autopct,labeldistance=None)
plt.ylabel('')
plt.legend(loc='center right',bbox_to_anchor=(1, -0.65));
plt.tight_layout();
transFigure = fig.transFigure.inverted()
coord1a = transFigure.transform(ax0.transData.transform([1,0]))
coord2a = transFigure.transform(ax1.transData.transform([0,-0.72]))
coord1b = transFigure.transform(ax0.transData.transform([-0.91,0.35]))
coord2b = transFigure.transform(ax1.transData.transform([0,0.72]))
linea = matplotlib.lines.Line2D((coord1a[0],coord2a[0]),(coord1a[1],coord2a[1]),
                               transform=fig.transFigure,c='black',alpha=0.7)
lineb = matplotlib.lines.Line2D((coord1b[0],coord2b[0]),(coord1b[1],coord2b[1]),
                                transform=fig.transFigure,c='black',alpha=0.7)
fig.lines = linea,lineb,
plt.savefig('concrete_breakdown_pie.pdf')
```

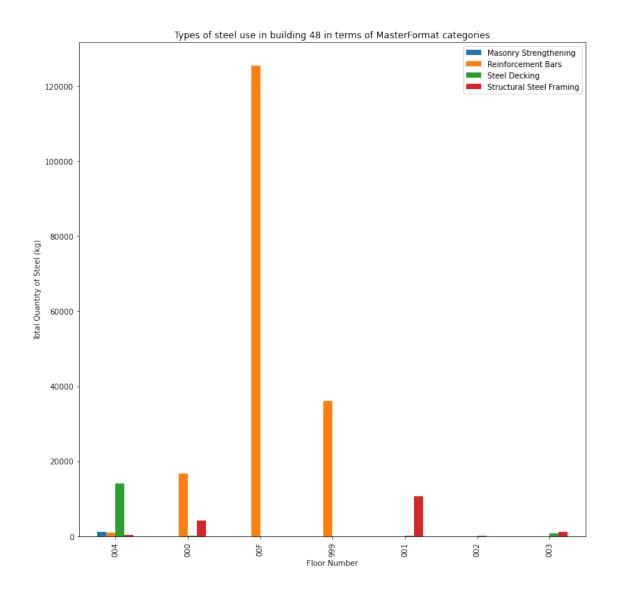


We can produce a pie chart for a single building, also.

```
[19]: mf_codes = pd.read_csv('mf_name_conversion.csv')
[20]: tofind = [
          'Plain Steel Reinforcement Bars',
          'Reinforcement Bars',
          'Structural Steel Framing',
          'Fabric and Grid Reinforcing',
          'Metal Doors',
          'Metal Roof Panel',
          'Metal Stairs',
          'Metal Railings',
          'Steel Decking',
          'Steel Joist Framing',
          'Steel'
      ] #List of terms we are looking to identify in column names.
      tokeep = [
          c for c in mf_codes.Title.values if any(t in c for t in tofind)
      ] #For each codes' corresponding in MasterFormat
      steel_codes = mf_codes[mf_codes.Title.isin(tokeep)]
```

```
[21]: columns_to_keep = []
     for column in df.columns:
         if 'kg' in column:
             code = re.split('_',column)[2]
             for k,c in steel_codes.values:
                 if c in code:
                     columns_to_keep.append(column)
[22]: f = lambda x: mf_codes[mf_codes.Code == str.replace(re.split('_',x)[2],'00','').
      steel_df = df[columns_to_keep].groupby(f,axis=1).sum()
[23]: (steel_df>0).sum(axis=1).sort_values()
[23]: 035
            1
     067
            1
     066
            1
     023
            1
     036
            1
     058
            4
     049
            4
     050
            4
     020
            4
     048
     Length: 70, dtype: int64
[24]: def my_autopct(pct):
         return ('%.2f' % (pct)) if pct > 1 else ''
     to_plot = steel_df.sum().sort_values(ascending=False)
     to_plot.plot.
      →pie(figsize=(12,12),colormap='tab20',autopct=my_autopct,labeldistance=None)
     plt.legend(loc='center left',bbox_to_anchor=(-0.30, 0.75));
     plt.ylabel('')
     plt.title(f'Types of steel use in all buildings in terms of MasterFormat⊔
      ⇔categories');
     plt.tight_layout();
     plt.savefig('steel_composition_pie.pdf')
```





We can also calculate the average for each Level 3 MasterFormat code by year of construction:

```
[27]: concrete_df = pd.concat([df[headings[1:]],df[cols].groupby(f,axis=1).

→sum()],axis=1)

concrete_df.groupby('Construction Date').mean()
```

```
[27]: Building Identifier Gross Floor Area Cast Decks and Underlayment/002 \
      Construction Date
      1913
                                  161.080000
                                                                           0.0
      1917
                                  199.930000
                                                                           0.0
      1969
                                  373.605000
                                                                           0.0
      1988
                                21934.000000
                                                                           0.0
      2007
                                73600.000000
                                                                           0.0
      2009
                                73083.000000
                                                                           0.0
```

2011	11282.500000	54943.2
2016	26841.666667	0.0
2017	35280.510000	0.0
2018	43365.090000	0.0
2019	107.050000	0.0
2020	10236.270000	0.0
2021	427.277895	0.0
2025	112537.000000	0.0
Building Identifier	Cast Decks and Underlayment/003	3 \
Construction Date	·	
1913	0.0	)
1917	0.0	
1969	0.0	)
1988	0.0	
2007	0.0	
2009	0.0	
2011	65145.6	
2016	0.0	
2017	0.0	
2018	0.0	
2019	0.0	
2020	0.0	
2021	0.0	
2025	0.0	
2020	0.0	,
Building Identifier	Cast Decks and Underlayment/999	) \
Construction Date		
1913	0.000000e+00	)
1917	0.000000e+00	)
1969	0.000000e+00	)
1988	0.000000e+00	)
2007	1.329816e+06	3
2009	0.000000e+00	)
2011	0.00000e+00	)
2016	0.00000e+00	)
2017	2.587372e+04	<u>l</u>
2018	0.00000e+00	)
2019	0.00000e+00	)
2020	0.00000e+00	
2021	0.00000e+00	
2025	0.000000e+00	
Building Identifier	Cast-in-Place Concrete/000 Cas	st-in-Place Concrete/001 \
Construction Date		
1913	0.00000e+00	0.00000e+00
1917	0.00000e+00	0.00000e+00

1969 1988 2007 2009 2011 2016 2017 2018 2019 2020 2021 2025		0.000000e+00 3.999773e+06 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00		0.000000e+00 1.435583e+06 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00	
Building Identifier Construction Date	Cast-in-Place	Concrete/002	Cast-in-Place	Concrete/003	\
		0 000000-100		0 000000-100	
1913		0.000000e+00		0.000000e+00	
1917		0.000000e+00		0.000000e+00	
1969		0.000000e+00		0.000000e+00	
1988		1.502795e+06 0.000000e+00		1.423554e+06 0.000000e+00	
2007					
2009		0.000000e+00		0.000000e+00	
2011		0.000000e+00		0.000000e+00	
2016		0.000000e+00		0.000000e+00	
2017		0.000000e+00		0.000000e+00	
2018 2019		0.000000e+00 0.000000e+00		0.000000e+00 0.000000e+00	
2020		0.000000e+00		0.000000e+00	
2021		0.000000e+00		0.000000e+00	
2025		0.000000e+00		0.000000e+00	
2025		0.00000e+00		0.00000e+00	
Building Identifier	Cast-in-Place	Concrete/004	Cast-in-Place	Concrete/005	\
Construction Date					
1913		0.000000e+00		0.00000	
1917		0.000000e+00		0.00000	
1969		0.000000e+00		0.00000	
1988		1.318964e+06	7	788129.689933	
2007		0.000000e+00		0.00000	
2009		0.000000e+00		0.00000	
2011		0.000000e+00		0.00000	
2016		0.000000e+00		0.00000	
2017		0.000000e+00		0.00000	
2018		0.000000e+00		0.000000	
2019		0.00000e+00		0.000000	
2020		0.00000e+00		0.000000	
2021		0.000000e+00		0.000000	
2025		0.00000e+00		0.000000	

Building Identifier	Structural	Concrete/99	99 Structu	ral Concrete/B	801 \
Construction Date	•••	^	0	C402E 4000	.00
1913	•••	0		64035.1900	
1917	•••		. 0	114018.4600	
1969	•••	0		132278.0150	
1988	•••		.0	0.0000	
2007	•••		.0	0.0000	
2009	•••	0		0.0000	
2011		0		0.0000	
2016		156360		0.0000	
2017	•••	205476		0.0000	
2018	•••	593112		0.0000	
2019	•••	0		47353.6840	
2020	•••	34108	.8	98902.9340	000
2021	•••	0	. 0	156066.4752	284
2025	***	847704	. 0	0.0000	000
Building Identifier Construction Date	Structural Co	oncrete/M00	Structural	Concrete/M10	\
1913		0.0		0.0	
1917		0.0		0.0	
1969		0.0		0.0	
1988		0.0		0.0	
2007		0.0		0.0	
2009		0.0		0.0	
2011		0.0		0.0	
2016		141136.0		0.0	
2017		0.0		0.0	
2018		633824.0		0.0	
2019		0.0		0.0	
2020		65894.4		0.0	
2021		0.0		0.0	
2025		0.0		391968.0	
Building Identifier Construction Date	Structural Co	oncrete/MP1	Structural	Concrete/P01	\
1913		0.0		0.0	
1917		0.0		0.0	
1969		0.0		0.0	
1988		0.0		0.0	
2007		0.0		0.0	
2009		0.0		0.0	
2011		0.0		0.0	
2016		0.0		1471112.0	
2017		0.0		2764302.0	
2018		0.0		2899816.0	
2019		0.0		0.0	

2020	0.0	610046.4
2021	0.0	0.0
2025	1405272.0	7396368.0
Building Identifier Construction Date	Structural Concrete/P02	Structural Concrete/P03
1913	0.0	0.0
1917	0.0	0.0
1969	0.0	0.0
1988	0.0	0.0
2007	0.0	0.0
2009	0.0	0.0
2011	0.0	0.0
2016	1143352.0	1064296.0
2017	2067108.0	2037768.0
2018	2405792.0	1837944.0
2019	0.0	0.0
2020	468100.8	466708.8
2021	0.0	0.0
2025	5522424.0	4559496.0
Building Identifier	Structural Concrete/P04	Structural Concrete/P05
Construction Date		
1913	0.0	0.0
1917	0.0	0.0
1969	0.0	0.0
1988	0.0	0.0
2007	0.0	0.0
2009	0.0	0.0
2011	0.0	0.0
2016	6087984.0	0.0
2017	1602108.0	609738.0
2018	2728856.0	0.0
2019	0.0	0.0
2020	1820392.8	0.0
2021	0.0	0.0
2025	6789888.0	0.0

\

[14 rows x 322 columns]

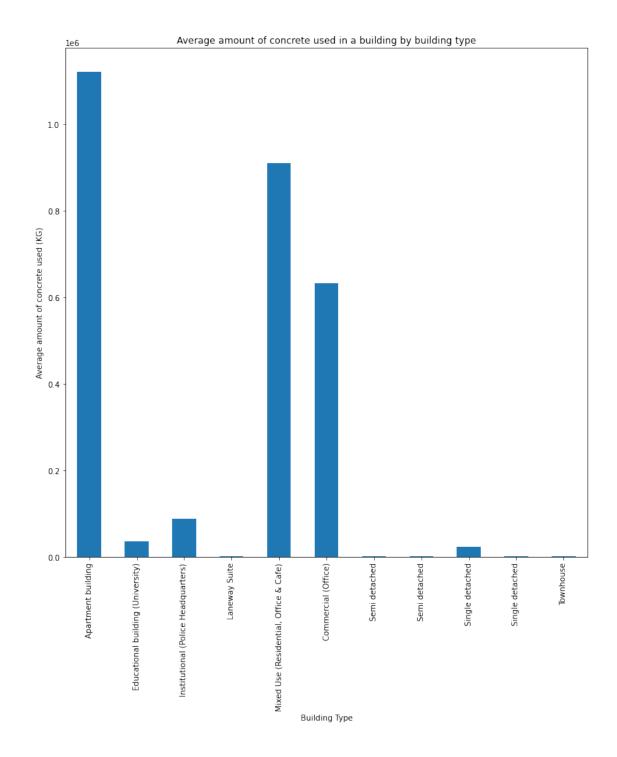
We can get the average amount of steel in KG used per building type:

```
[28]: concrete_df.groupby('Building Type').sum().mean(axis=1).

→rename(index=building_name_map).plot(kind='bar',figsize=(12,12))

plt.ylabel('Average amount of concrete used (KG)')

plt.title('Average amount of concrete used in a building by building type');
```



# 4 3. Uncertainty by Building Type

In this section, we look at the uncertainty score associated with each material takeoff. We collect these by building type and then report the number of each value per type of building.

```
[29]: uncertainty_level = {}
      for k,v in df.iterrows():
          #Initialise empty lists for each building type as they occur
          if v['Building Type'] not in uncertainty_level.keys():
              uncertainty_level[v['Building Type']] = []
          #Append the uncertainty value for each column that is non-NaN
          for key in v[~v.isna()].keys()[7:]:
              uncertainty_level[v['Building Type']].append(key.split('_')[-1])
[30]: from collections import Counter
[31]: for k,v in uncertainty level.items():
          uncertainty_level[k] = Counter(v) #Construct a Counter object per building_
       \hookrightarrow type
[32]: uncertainty_level
[32]: {'SND': Counter({'3': 626, '2': 1582, '5': 284}),
       'OFF': Counter({'2': 491, '4': 307}),
       'APB': Counter({'2': 1844, '3': 1, '4': 1601}),
       'SMR': Counter({'2': 20, '3': 26, '5': 8}),
       'SNR': Counter({'2': 55, '3': 70, '5': 52}),
       'SMD': Counter({'2': 167, '3': 34, '5': 19}),
       'EDU': Counter({'2': 91, '4': 24, '3': 6}),
       'INS': Counter({'4': 77, '2': 89, '3': 1}),
       'MIX': Counter({'2': 1262, '4': 1047}),
       'LNW': Counter({'3': 92, '2': 287, '5': 21}),
       'TWN': Counter({'2': 58, '4': 6})}
     Next, we aggregate columns by the purporse of the material and uncertainty combined, and report
     the average by building type.
[33]: f = lambda x: name_map[re.split('[_\.\]',x)[1][0]] + '/' + x.split('_')[-1].
       →split('.')[0] #From a full code, return only the use code and uncertainty ⊔
       \rightarrowscore.
      by function df = pd.concat([df[headings[1:]],df[cols].groupby(f,axis=1).
       \rightarrowsum()],axis=1)
[34]: by_function_df.groupby('Building Type').mean().rename(index=building_name_map).

→drop(['Construction Date'],axis=1).round(2)
[34]: Building Identifier
                                               Gross Floor Area Interiors/2 \
      Building Type
      Apartment building
                                                        39160.26
                                                                   5624203.35
      Educational building (University)
                                                         7901.00
                                                                    480382.15
      Institutional (Police Headquarters)
                                                        21934.00
                                                                   1295281.75
      Laneway Suite
                                                                         0.00
                                                          128.88
```

Mixed Use (Residential, Office & Cafe) Commercial (Office)		60.42 1273 43.67 989	16484.57 98215.44		
Semi detached	24	48.84	0.00		
Semi detached	19	99.93	0.00		
Single detached	47	78.40	0.00		
Single detached	30	02.76	0.00		
Townhouse	356	66.00	0.00		
Building Identifier Building Type	Interiors/3	Interiors	s/4 Servio	ces/2	\
Apartment building	0.00	171337	.00 15292	274.0	
Educational building (University)	3096.66	14080	. 27	0.0	
Institutional (Police Headquarters)	0.00	40860	.46	0.0	
Laneway Suite	0.00	0	.00	0.0	
Mixed Use (Residential, Office & Cafe)	0.00			736.0	
Commercial (Office)	0.00			0.0	
Semi detached	0.00		.00	0.0	
Semi detached	0.00		.00	0.0	
Single detached	0.00		.00	0.0	
Single detached	68.77		.00	0.0	
Townhouse	0.00	0	.00	0.0	
Building Identifier Building Type	Services/4	Shell	/2 Shell	1/3 \	
Apartment building	50074.69	20886862.8	34 0.	.00	
Educational building (University)	0.00	1520252.5		. 64	
Institutional (Police Headquarters)	0.00	17371405.9		.00	
Laneway Suite	0.00	0.0	0.	.00	
Mixed Use (Residential, Office & Cafe)	237801.46	51743951.2		.00	
Commercial (Office)	0.00	43308969.3	36 0.	.00	
Semi detached	0.00	1866.9	95 5.	. 41	
Semi detached	0.00	0.0	00 40.	. 11	
Single detached	0.00	1549.4	49 22.	. 18	
Single detached	0.00	2504.9	95 6.	. 65	
Townhouse	0.00	0.0	0.	.00	
Building Identifier Building Type	Shell/4	Shell/5	Sitework/2	\	
Apartment building	761128.06	0.00	14493.0		
Educational building (University)	7713.03	0.00	0.0		
Institutional (Police Headquarters)	656655.11	0.00	0.0		
Laneway Suite	0.00	0.00	0.0		
Mixed Use (Residential, Office & Cafe)	1909970.03	0.00	0.0		
Commercial (Office)	1621345.80	0.00	89288.0		
Semi detached	0.00	0.00	0.0		
Semi detached	0.00	0.00	0.0		
Single detached	0.00	0.93	0.0		
DIEGIO GERACHEA	0.00	0.30	0.0		

Single detached Townhouse	0.00 0.00 0.0 0.00 0.00 0.0	
Building Identifier Building Type	Sitework/4 \	
Apartment building	474.04	
Educational building (University)	0.00	
Institutional (Police Headquarters)	0.00	
Laneway Suite	0.00	
Mixed Use (Residential, Office & Cafe)	0.00	
Commercial (Office)	3016.86	
Semi detached	0.00	
Semi detached	0.00	
Single detached	0.00	
Single detached	0.00	
Townhouse	0.00	
Building Identifier Building Type	Special Construction And Demolition/2	\
Apartment building	37698.0	
Educational building (University)	0.0	
Institutional (Police Headquarters)	0.0	
Laneway Suite	0.0	
Mixed Use (Residential, Office & Cafe)	249760.0	
Commercial (Office)	0.0	
Semi detached	0.0	
Semi detached	0.0	
Single detached	0.0	
Single detached	0.0	
Townhouse	0.0	
Building Identifier Building Type	Special Construction And Demolition/4	\
Apartment building	339.88	
Educational building (University)	0.00	
Institutional (Police Headquarters)	0.00	
Laneway Suite	0.00	
Mixed Use (Residential, Office & Cafe)	9000.47	
Commercial (Office)	0.00	
Semi detached	0.00	
Semi detached	0.00	
Single detached	0.00	
Single detached	0.00	
Townhouse	0.00	
Building Identifier	Substructure/2 Substructure/3 \	
Building Type		

Apartment building	15661850.24	109212.00
Educational building (University)	2793438.68	0.00
Institutional (Police Headquarters)	8890567.75	0.00
Laneway Suite	48858.34	2104.66
Mixed Use (Residential, Office & Cafe)	22801051.57	0.00
Commercial (Office)	12411535.27	0.00
Semi detached	97751.05	7.78
Semi detached	110261.75	8921.68
Single detached	181911.50	5413.20
Single detached	93196.84	19429.34
Townhouse	534318.99	0.00
Building Identifier	Substructure //	Substructure/5
Dullding Identifier	Substitucture/4	bubber accure, o
Building Type	Substituture/4	bubber de care, o
_	365922.73	0.00
Building Type		
Building Type Apartment building	365922.73	0.00
Building Type Apartment building Educational building (University)	365922.73 91853.12	0.00
Building Type Apartment building Educational building (University) Institutional (Police Headquarters)	365922.73 91853.12 239579.15	0.00 0.00 0.00
Building Type Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite	365922.73 91853.12 239579.15 0.00	0.00 0.00 0.00 0.65
Building Type Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite Mixed Use (Residential, Office & Cafe)	365922.73 91853.12 239579.15 0.00 645320.97	0.00 0.00 0.00 0.65 0.00
Building Type Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite Mixed Use (Residential, Office & Cafe) Commercial (Office)	365922.73 91853.12 239579.15 0.00 645320.97 354767.84	0.00 0.00 0.00 0.65 0.00
Building Type Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite Mixed Use (Residential, Office & Cafe) Commercial (Office) Semi detached	365922.73 91853.12 239579.15 0.00 645320.97 354767.84 0.00	0.00 0.00 0.00 0.65 0.00 0.00
Building Type Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite Mixed Use (Residential, Office & Cafe) Commercial (Office) Semi detached Semi detached	365922.73 91853.12 239579.15 0.00 645320.97 354767.84 0.00 0.00	0.00 0.00 0.00 0.65 0.00 0.00 6.93 0.00
Building Type Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite Mixed Use (Residential, Office & Cafe) Commercial (Office) Semi detached Semi detached Single detached	365922.73 91853.12 239579.15 0.00 645320.97 354767.84 0.00 0.00	0.00 0.00 0.00 0.65 0.00 0.00 6.93 0.00 38.46

Next, we report the total amount of material falling under each uncertainty score by year of construction.

```
[35]: f = lambda x: x.split('_')[-1].split('.')[0] #Select only the uncertainty score.

print('Average amount of material used per building, by year and uncertainty

→score (%)')

result = pd.concat([df['Construction Date'],df[[c for c in df.columns if 'kg'

→in c]].groupby(f,axis=1).sum()],axis=1).groupby('Construction Date').mean()

for k,v in result.iterrows():

result.loc[k,:] = v/v.sum()

display(result.round(2))
```

Average amount of material used per building, by year and uncertainty score (%)

	2	3	4	5
Construction Date				
1913	0.85	0.08	0.00	0.07
1917	0.75	0.14	0.00	0.11
1969	0.50	0.37	0.00	0.13
1988	0.97	0.00	0.03	0.00
2007	0.97	0.00	0.03	0.00
2009	0.97	0.00	0.03	0.00

```
2011
                 0.94 0.03 0.03 0.00
2016
                 0.96 0.02 0.03 0.00
                 0.97 0.00 0.03 0.00
2017
2018
                 0.97 0.00 0.03 0.00
2019
                 0.98
                       0.02 0.00 0.00
2020
                 0.97
                       0.00
                            0.03 0.00
2021
                 0.78
                       0.09 0.00 0.13
2025
                 0.97
                       0.00 0.03 0.00
```

## 5 4. Material Intensity

We can easily calculate material intensity by dividing takeoffs which are measured in kilograms by the Gross Floor Area:

```
[36]: kilogram_columns = [d for d in df.columns if 'kg' in d]
      df_mi = df[kilogram_columns].div(df['Gross Floor Area'],axis=0)
[37]: kilogram_columns = [d for d in df.columns if 'kg' in d]
      df_mi = df[kilogram_columns].div(df['Gross Floor Area'],axis=0)
      f = lambda x: name_map[re.split('[_\.\]',x)[1][0:3]]
      pd.concat([df[headings[1:]],df mi[kilogram columns].groupby(f,axis=1).
       [37]: Building Identifier City Quality / Stage of Data Construction Date \
      001
                           TOR
                                                 00IFC
                                                                      2021
      002
                           TOR
                                                 00IFC
                                                                      2021
      003
                           TOR
                                                 OOIFC
                                                                      2021
      004
                           TOR
                                                 00IFC
                                                                      2021
      007
                           TOR
                                                 00IFC
                                                                      2021
      800
                           TOR
                                                 00IFC
                                                                      2021
      009
                           TOR
                                                 00IFC
                                                                      2021
      010
                           TOR
                                                 00IFC
                                                                      2021
      013
                           TOR
                                                 00IFC
                                                                      2021
      014
                                                                      2021
                           TOR
                                                 OOIFC
      015
                                                                      2021
                           TOR
                                                 OOIFC
      016
                           TOR
                                                 OOIFC
                                                                      2021
      019
                                                                      2021
                           TOR
                                                 00IFC
      020
                                                 OOIFC
                                                                      2021
                           TOR
      021
                                                                      2020
                           TOR
                                                 00IFC
      022
                           TOR
                                                 00IFC
                                                                      2021
      023
                           TOR
                                                 00IFC
                                                                      2021
      025
                           TOR
                                                 00IFC
                                                                      2021
      026
                           TOR
                                                 00IFC
                                                                      2021
      028
                           TOR
                                                 OOIFC
                                                                      2021
      029
                           TOR
                                                 00IFC
                                                                      2021
      031
                           TOR
                                                 00IFC
                                                                      2021
      032
                           TOR
                                                 00IFC
                                                                      2021
```

033 035 036 037 038 039 041 043 044	TOR	001FC 001FC 001FC 001FC 001FC 001FC 001FC 001FC 001FC 001FC	2020 2021 2021 2021 2020 2021 2021 2021
049	TOR	00IFC	2020
050	TOR	00IFC	2021
Building Identifier 001 002 003 004 007 008 009 010 013 014 015 016 019 020 021 022 023 025 026 028 029 031 032 033 035 036	Building Type SND	Gross Floor Area 521.18 389.24 411.64 269.56 445.99 438.45 714.07 343.24 226.89 611.73 343.44 613.38 178.38 323.80 837.56 587.86 568.21 294.84 496.77 643.30 701.61 378.70 324.16 533.53 423.03 328.16	Conveying \
037	SND	421.59	0.0
038	SND	628.59	0.0
039	SND	464.51	0.0
041	SND	346.14	0.0
043	SND	891.97	0.0

```
044
                                                                0.0
                                 SND
                                                 525.61
045
                                 SND
                                                 502.87
                                                                0.0
046
                                 SND
                                                 379.18
                                                                0.0
047
                                                                0.0
                                 SND
                                                 549.65
049
                                 SND
                                                 393.82
                                                                0.0
050
                                 SND
                                                                0.0
                                                 648.14
Building Identifier Exterior Horizontal Enclosures
001
                                              11.137992
002
                                               5.461939
003
                                               3.786074
004
                                               6.503479
007
                                              11.933511
800
                                              12.707195
009
                                              12.865930
010
                                               4.300619
013
                                              12.424245
014
                                               5.140200
015
                                               6.494467
016
                                              13.090524
019
                                               9.782438
020
                                               9.824569
021
                                              13.521848
022
                                               6.949783
023
                                              12.754287
025
                                               3.650542
026
                                               5.352985
028
                                              11.769043
029
                                              11.799093
031
                                               5.522739
032
                                               5.361174
033
                                               8.494907
035
                                              11.102019
036
                                              10.234937
037
                                              12.223172
038
                                              10.408758
039
                                               4.118745
041
                                              11.787081
043
                                              10.710312
044
                                              18.918490
045
                                               6.014586
046
                                               6.169302
047
                                              11.310711
049
                                              16.116861
050
                                               9.684756
```

Building Identifier Exterior Vertical Enclosures Foundations  $\setminus$ 

001	136.	939623	335.649367		
002		018253	281.318698		
003		450370	464.462195		
004		215196	255.359136		
007		325975	295.116668		
008		552921	269.468463		
009		310510	276.917123		
010		632241	283.893850		
013		668275	261.874926		
014		332008	343.714248		
015	147.	104280	424.099610		
016	156.	986570	298.537712		
019	112.	523711	371.149916		
020	186.	570501	148.769711		
021	91.	689386	317.583491		
022	94.	557055	428.185321		
023	83.	789887	255.012975		
025		856507	261.274626		
026		883144	251.725837		
028		949693	156.365248		
029		418023	266.164355		
031		214896	403.602589		
032		512918	377.853541		
033		518430	309.062696		
035	154.	072547	243.607664		
036	184.	202156	388.744353		
037	158.	716507	424.443503		
038	136.	076590	369.744859		
039	151.	068033	412.845205		
041	146.	479339	287.564257		
043	213.	677214	245.205806		
044	109.	529933	498.010299		
045		481074	278.679758		
046		418003			
047		866168			
049		069509			
050		118584			
030	131.	110004	347.187490		
Duilding Identifies	Interior Construction	Tn+onic	m Finiahoa	Dlumbina	\
•	Interior Construction	interio		Plumbing	\
001	16.482129		6.202080	0.0	
002	12.248343		4.491260	0.0	
003	15.931829		3.030369	0.0	
004	4.574132		2.920482	0.0	
007	19.773909		4.539900	0.0	
008	10.683759		4.767511	0.0	
009	18.937583		4.898301	0.0	
010	17.891930		6.753884	0.0	

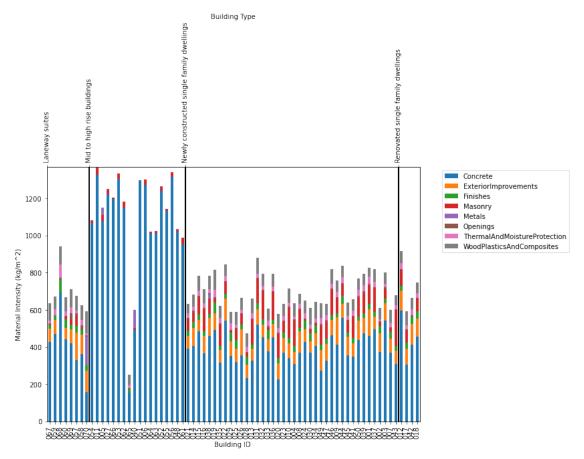
013	17.256393	4.154604	0.0
014	13.258982	5.577869	0.0
015	18.195449	5.729880	0.0
016	17.589067	5.763898	0.0
019	19.638502	7.549843	0.0
020	18.186467	3.384055	0.0
021	17.799752		
022	19.088554		
023	23.268519		
025	20.047035		
026	14.370613		
028	16.010229		
029	23.078653		
031	19.181898		0.0
032	24.166732		
032	34.027695		
035			
	16.390809		
036	7.854953		
037	16.125050		
038	16.271010		
039	15.108900		
041	19.523228		
043	20.691791		
044	19.155639		
045	22.485115		
046	16.651076		
047	20.753973		
049	22.332639		
050	23.995586	3.705203	0.0
~	Site Improvements Sl	abs-On-Grade Specia	al Construction \
001	0.0	273.972401	0.0
002	0.0	192.874465	0.0
003	0.0	170.733356	0.0
004	0.0	124.186526	0.0
007	0.0	153.061618	0.0
008	0.0	211.910108	0.0
009	0.0	266.709576	0.0
010	0.0	138.510228	0.0
013	0.0	129.263543	0.0
014	0.0	165.513154	0.0
015	0.0	129.532248	0.0
016	0.0	166.414337	0.0
019	0.0	223.398638	0.0
020	0.0	158.178114	0.0
021	0.0	143.282268	0.0
022	0.0	237.918968	0.0
V			

023	0.0	199.364347	0.0
025	0.0	131.174185	0.0
026	0.0	242.284758	0.0
028	0.0	152.407914	0.0
029	0.0	169.419640	0.0
031	0.0	179.868896	0.0
032	0.0	132.696247	0.0
033	0.0	135.390288	0.0
035	0.0	147.458950	0.0
036	0.0	128.887840	0.0
037	0.0	147.225241	0.0
038	0.0	186.334547	0.0
039	0.0	145.273403	0.0
041	0.0	139.821081	0.0
043	0.0	138.994603	0.0
044	0.0	139.646277	0.0
045	0.0	182.059329	0.0
046	0.0	158.446049	0.0
047	0.0	154.805714	0.0
049	0.0	198.860705	0.0
050	0.0	199.209464	0.0
D :11: T1			
Bullding Identifier	Subgrade Enclosures	Substructure Interior	\
001	9.652903	7.521547	
001 002	9.652903 6.851955	7.521547 11.871041	
002 003	6.851955 11.298572	11.871041 8.277288	
002 003 004	6.851955 11.298572 4.351465	11.871041 8.277288 20.070275	
002 003 004 007	6.851955 11.298572 4.351465 9.478642	11.871041 8.277288 20.070275 5.575509	
002 003 004 007 008	6.851955 11.298572 4.351465 9.478642 4.218921	11.871041 8.277288 20.070275 5.575509 1.817270	
002 003 004 007 008 009	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687	
002 003 004 007 008	6.851955 11.298572 4.351465 9.478642 4.218921	11.871041 8.277288 20.070275 5.575509 1.817270	
002 003 004 007 008 009	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687	
002 003 004 007 008 009	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759	
002 003 004 007 008 009 010 013	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162	
002 003 004 007 008 009 010 013 014	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240	
002 003 004 007 008 009 010 013 014 015	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152	
002 003 004 007 008 009 010 013 014 015 016	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.0000000	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309	
002 003 004 007 008 009 010 013 014 015	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152	
002 003 004 007 008 009 010 013 014 015 016	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.0000000	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309	
002 003 004 007 008 009 010 013 014 015 016	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.000000 4.617006	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309 11.946436	
002 003 004 007 008 009 010 013 014 015 016 019 020 021 022	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.000000 4.617006 7.131170 7.959752	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309 11.946436 8.875410 9.098153	
002 003 004 007 008 009 010 013 014 015 016 019 020 021	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.000000 4.617006 7.131170 7.959752 6.339651	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309 11.946436 8.875410 9.098153 11.209887	
002 003 004 007 008 009 010 013 014 015 016 019 020 021 022 023 025	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.000000 4.617006 7.131170 7.959752 6.339651 7.469048	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309 11.946436 8.875410 9.098153 11.209887 3.895085	
002 003 004 007 008 009 010 013 014 015 016 019 020 021 022 023 025 026	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.000000 4.617006 7.131170 7.959752 6.339651 7.469048 9.448689	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309 11.946436 8.875410 9.098153 11.209887 3.895085 4.154656	
002 003 004 007 008 009 010 013 014 015 016 019 020 021 022 023 025 026 028	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.000000 4.617006 7.131170 7.959752 6.339651 7.469048 9.448689 0.000000	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309 11.946436 8.875410 9.098153 11.209887 3.895085 4.154656 11.506782	
002 003 004 007 008 009 010 013 014 015 016 019 020 021 022 023 025 026	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.000000 4.617006 7.131170 7.959752 6.339651 7.469048 9.448689	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309 11.946436 8.875410 9.098153 11.209887 3.895085 4.154656	
002 003 004 007 008 009 010 013 014 015 016 019 020 021 022 023 025 026 028	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.000000 4.617006 7.131170 7.959752 6.339651 7.469048 9.448689 0.000000	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309 11.946436 8.875410 9.098153 11.209887 3.895085 4.154656 11.506782	
002 003 004 007 008 009 010 013 014 015 016 019 020 021 022 023 025 026 028 029 031	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.000000 4.617006 7.131170 7.959752 6.339651 7.469048 9.448689 0.000000 11.919460 7.509119	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309 11.946436 8.875410 9.098153 11.209887 3.895085 4.154656 11.506782 8.789598 10.575300	
002 003 004 007 008 009 010 013 014 015 016 019 020 021 022 023 025 026 028 029	6.851955 11.298572 4.351465 9.478642 4.218921 8.902623 9.601245 3.818403 7.722754 9.135529 4.868508 0.000000 4.617006 7.131170 7.959752 6.339651 7.469048 9.448689 0.000000 11.919460	11.871041 8.277288 20.070275 5.575509 1.817270 25.192687 7.744759 9.532825 6.168162 5.601240 9.004152 8.758309 11.946436 8.875410 9.098153 11.209887 3.895085 4.154656 11.506782 8.789598	

035	0.000000		10.013415	
036	4.762839		19.086997	
037	9.538939		12.833857	
038	6.039206		7.143042	
039	9.071017		12.485838	
041	7.568785		12.011677	
043	4.540919		10.725241	
044	6.720435		8.275280	
045	6.092739		10.878686	
046	9.489156		13.750663	
047	6.042229		8.345960	
049	6.057127		5.861907	
050	7.221222		8.240307	
Building Identifier	Substructure Related	Activities	Superstructure	\
001		0.0	30.228003	
002		0.0	26.271523	
003		0.0	23.756286	
004		0.0	30.396721	
007		0.0	39.906513	
008		0.0	39.907474	
009		0.0	38.291591	
010		0.0	35.370538	
013		0.0	35.355314	
014		0.0	33.388004	
015		0.0	39.370016	
016		0.0	40.958564	
019		0.0	63.006044	
020		0.0	36.597047	
021		0.0	28.734226	
022		0.0	37.457583	
023		0.0	36.265538	
025		0.0	30.389475	
026		0.0	43.728928	
028		0.0	35.393414	
029		0.0	39.408113	
031		0.0	82.392236	
032		0.0	46.380703	
033		0.0	25.469871	
035		0.0	35.666107	
036		0.0	49.284461	
037		0.0	34.035382	
038		0.0	47.065025	
039		0.0	37.921434	
041		0.0	27.740220	
043		0.0	29.045531	
044		0.0	33.265489	

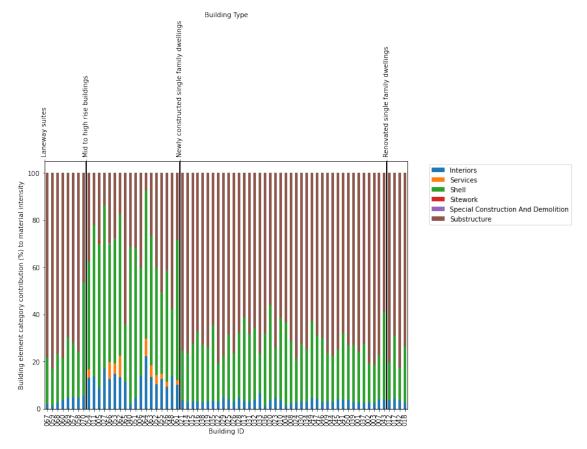
```
045
                                                          0.0
                                                                     37.265275
      046
                                                          0.0
                                                                     46.860447
      047
                                                          0.0
                                                                     31.152827
      049
                                                          0.0
                                                                     49.899420
      050
                                                          0.0
                                                                     38.021046
      Building Identifier Water And Gas Mitigation
      001
                                                   0.0
      002
                                                   0.0
      003
                                                   0.0
      004
                                                   0.0
      007
                                                   0.0
      800
                                                   0.0
      009
                                                   0.0
      010
                                                   0.0
      013
                                                   0.0
      014
                                                   0.0
      015
                                                   0.0
      016
                                                   0.0
      019
                                                   0.0
      020
                                                   0.0
      021
                                                   0.0
      022
                                                   0.0
      023
                                                   0.0
      025
                                                   0.0
      026
                                                   0.0
      028
                                                   0.0
      029
                                                   0.0
      031
                                                   0.0
      032
                                                   0.0
      033
                                                   0.0
      035
                                                   0.0
      036
                                                   0.0
      037
                                                   0.0
      038
                                                   0.0
      039
                                                   0.0
      041
                                                   0.0
      043
                                                   0.0
                                                   0.0
      044
      045
                                                   0.0
      046
                                                   0.0
      047
                                                   0.0
      049
                                                   0.0
      050
                                                   0.0
[38]: master_format_convert = {v:k for k,v in {
          'Concrete':'03',
```

```
'Masonry':'04',
          'Metals':'05',
          'WoodPlasticsAndComposites':'06',
          'ThermalAndMoistureProtection':'07',
          'Finishes':'09',
          'Openings':'08',
          'Earthwork': '31',
          'ExteriorImprovements':'32'
     }.items() }
[39]: f = lambda x: master_format_convert[re.split('[\.\]',x)[4]]
     toplot = pd.concat([df[headings[1:]],df_mi[kilogram_columns].groupby(f,axis=1).
      [40]: building type map = dict(building name conversion[['Building Code', 'Type']].
      →values)
     toplot['Building Type'] = toplot['Building Type'].replace(building_type_map)
     toplot = toplot.sort_values('Building Type')
[41]: set(df['Building Type'].values)
[41]: {'APB', 'EDU', 'INS', 'LNW', 'MIX', 'OFF', 'SMD', 'SMR', 'SND', 'SNR', 'TWN'}
[42]: fig, ax = plt.subplots(figsize=(10,7))
     cols = toplot.columns[5:]
     margin_bottom = np.zeros(len(toplot))
     cmap = plt.get cmap('tab10')
     for num, col in enumerate(cols):
         values = toplot[col].values
         toplot[col].plot.bar(x='Year',y='Value', ax=ax, stacked=True,
                                         bottom = margin_bottom, color=cmap(num),__
      →label=col)
         margin bottom += values
     plt.legend(bbox_to_anchor=(1.05, 1), loc='upper left')
     plt.ylabel('Material Intensity (kg/m^2)')
     plt.xlabel('Building ID ')
     ax2 = ax.twiny()
     ax2.set_xlim(0, len(toplot))
     ax2.set_xticks([k for k,v in enumerate(toplot['Building Type'].values) if v !=_
      →toplot['Building Type'].values[k-1] or k==0])
     for tick in ax2.get xticklabels():
         tick.set_rotation(90)
```



Newly constructed single family dwellings Renovated single family dwellings		461.18 277.06		
Building Identifier Building Type	Exterior	Improvem	ents Fini	shes \
Laneway suites		8	1.80 3	35.26
Mid to high rise buildings				0.99
Newly constructed single family dwellings				31.17
Renovated single family dwellings				33.64
Renovated Single Tamily dwellings		10	0.50	00.04
Building Identifier Building Type	Masonry	Metals	Openings	\
Laneway suites	11.68	18.65	12.42	
Mid to high rise buildings		8.57		
Newly constructed single family dwellings	83.77			
Renovated single family dwellings	55.31	0.74	5.84	
Building Identifier Building Type	ThermalAndMoistureProtection \			
Laneway suites			28.	81
Mid to high rise buildings				97
Newly constructed single family dwellings	25.63			
Renovated single family dwellings	26.98			
Building Identifier Building Type	WoodPlas	ticsAndC	omposites	Total MI
Laneway suites			89.25	690.83
Mid to high rise buildings			3.02	1102.96
Newly constructed single family dwellings			68.82	699.22
Renovated single family dwellings			64.59	730.36
Std Dev Material Intensity:				
Building Identifier Building Type	Gross Fl	oor Area	Concrete	· \
Laneway suites		50.64	150.56	3
Mid to high rise buildings		32439.52	302.20	)
Newly constructed single family dwellings		168.17	82.14	<u> </u>
Renovated single family dwellings		117.28	120.26	3
Building Identifier Building Type	Exterior	Improvem	ents Fini	.shes \
Laneway suites	43.53 17.14		7.14	
Mid to high rise buildings	0.00 4.22			
Newly constructed single family dwellings			9.40	
Renovated single family dwellings			2.94	6.38
TOTAL TARGET TARILLY AWOLLINGS		_	2.01	
Building Identifier Building Type	Masonry	Metals	Openings	\

```
Laneway suites
                                                  20.51
                                                          52.54
                                                                    11.58
     Mid to high rise buildings
                                                  11.42
                                                          23.39
                                                                      1.66
     Newly constructed single family dwellings
                                                  49.26
                                                           3.35
                                                                      2.21
     Renovated single family dwellings
                                                  37.88
                                                           0.86
                                                                      1.43
     Building Identifier
                                                ThermalAndMoistureProtection \
     Building Type
     Laneway suites
                                                                        19.63
     Mid to high rise buildings
                                                                         4.13
     Newly constructed single family dwellings
                                                                        6.14
     Renovated single family dwellings
                                                                        5.44
     Building Identifier
                                                WoodPlasticsAndComposites Total MI
     Building Type
                                                                    17.69
     Laneway suites
                                                                             107.85
     Mid to high rise buildings
                                                                    12.82
                                                                             280.73
     Newly constructed single family dwellings
                                                                    11.58
                                                                              95.96
     Renovated single family dwellings
                                                                     6.55
                                                                             140.02
[45]: df_mi = df[kilogram_columns].div(df['Gross Floor Area'],axis=0)
[46]: df_mi = df[kilogram_columns].div(df['Gross Floor Area'],axis=0)
      df_mi = df_mi.div(df_mi.sum(axis=1),axis=0) * 100
      f = lambda x: name_map[re.split('[\.\]',x)[1][0]]
      toplot = pd.concat([df[headings[1:]],df_mi[kilogram_columns].groupby(f,axis=1).
      →sum()],axis=1).sort_values('Building Type')
      toplot['Building Type'] = toplot['Building Type'].replace(building_type_map)
      toplot = toplot.sort values('Building Type')
      fig, ax = plt.subplots(figsize=(10,7))
      cols = toplot.columns[5:]
      margin_bottom = np.zeros(len(toplot))
      cmap = plt.get_cmap('tab10')
      for num, col in enumerate(cols):
          values = toplot[col].values
          toplot[col].plot.bar(x='Year',y='Value', ax=ax, stacked=True,
                                          bottom = margin_bottom, color=cmap(num),__
      →label=col)
          margin_bottom += values
      plt.legend(bbox_to_anchor=(1.05, 1), loc='upper left')
      plt.xlabel('Building ID')
      plt.ylabel('Building element category contribution (%) to material intensity')
      ax2 = ax.twiny()
```



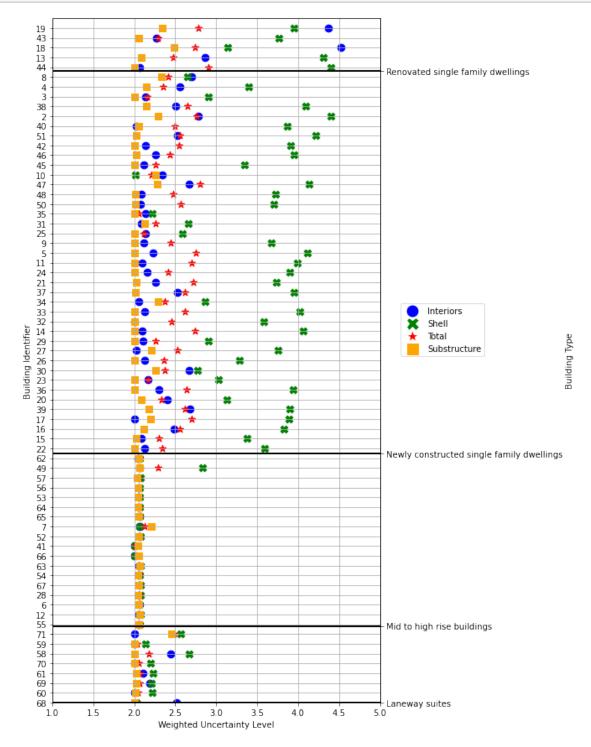
```
for i in range (1,6):
          toplot[f'Total/{i}'] = 0
      for k,v in toplot.iteritems():
          toplot[f'Total/{k.split("/")[1]}'] += v
      toplot_out = deepcopy(toplot)
      for k,v in toplot.iteritems():
          toplot_out[k] = (v/toplot[[c for c in toplot.columns if k.split('/')[0] in_
      \rightarrowc]].sum(axis=1)) * int(k.split('/')[1])
      f = lambda x: x.split('/')[0]
      toplot_out = pd.concat([df['Building Type'],toplot_out.groupby(f,axis=1).
      →sum()],axis=1).sort_values('Building Type')
      toplot out = toplot out.reset index()
      toplot_out['index'] = toplot_out['index'].astype('int') + 1
      toplot_out['index'] = toplot_out['index'].astype('str')
[49]: # toplot_out = toplot_out[toplot_out['Building Type'].isin(types_to_keep)]
      toplot_out['Building Type'] = toplot_out['Building Type'].
      →replace(building_type_map)
      toplot_out = toplot_out.sort_values('Building Type')
[50]: from matplotlib.lines import Line2D
      fig, ax = plt.subplots(figsize=(7,15))
      ax.set_xlim(1,5)
      ax.set_ylim(0,len(toplot_out))
      # ax.set_yticks(toplot_out['index'])
      handles = []
      for v,m,c in ...
      →[('Interiors','o','blue'),('Shell','X','green'),('Total','*','red'),('Substructure','s','or
          ax.scatter(x=toplot_out[v].values,y=toplot_out['index'].values, marker=m,__
       \rightarrowcolor=c, s=75)
          handles.append(
              Line2D([0], [0], marker=m, color='w', label=v,
                                    markerfacecolor=c, markersize=15)
      plt.legend(handles=handles,bbox_to_anchor=(1.05, 0.5), loc='lower left')
      plt.ylabel('Building Identifier')
      plt.xlabel('Weighted Uncertainty Level')
      plt.grid()
      ax2 = ax.twinx()
      ax2.set_ylim(0, len(toplot_out))
      ax2.set_yticks([k-1.5 for k,v in enumerate(toplot_out['Building Type'].values)_
      →if v != toplot_out['Building Type'].values[k-1] or k==0])
      # for tick in ax2.get yticklabels():
            tick.set rotation(90)
```

```
ax2.set_yticklabels([v for k,v in enumerate(toplot_out['Building Type'].values)

if v != toplot_out['Building Type'].values[k-1] or k==0])

ax2.set_ylabel("Building Type")

plt.grid(color='black',linewidth=2)
```



```
toplot_out
[51]:
         index
                                               Building Type
                                                                Interiors
                                                                            Services \
                                              Laneway suites
      18
             68
                                                                 2.520976
                                                                                 0.0
      17
             60
                                              Laneway suites
                                                                                 0.0
                                                                 2.000000
      16
             69
                                              Laneway suites
                                                                 2.190950
                                                                                 0.0
      15
             61
                                              Laneway suites
                                                                 2.106514
                                                                                 0.0
      11
            70
                                              Laneway suites
                                                                 2.005075
                                                                                 0.0
                 Newly constructed single family dwellings
      34
            44
                                                                                 0.0
                                                                 2.065774
      28
             13
                         Renovated single family dwellings
                                                                                 0.0
                                                                 2.868511
                         Renovated single family dwellings
      66
                                                                                 0.0
             18
                                                                 4.523878
                         Renovated single family dwellings
      67
             43
                                                                 2.275307
                                                                                 0.0
                         Renovated single family dwellings
      68
             19
                                                                 4.371953
                                                                                 0.0
                     Sitework
                                Special Construction And Demolition
             Shell
                                                                        Substructure
      18
          2.025651
                           0.0
                                                                   0.0
                                                                             2.000000
                           0.0
                                                                   0.0
      17
          2.222478
                                                                             2.009786
      16
          2.206190
                           0.0
                                                                   0.0
                                                                             2.023037
                                                                   0.0
      15
          2.229146
                           0.0
                                                                             2.024653
          2.200763
                           0.0
                                                                   0.0
                                                                             2.000000
      11
      . .
      34
          4.396133
                           0.0
                                                                   0.0
                                                                             2.005594
      28
          4.306551
                           0.0
                                                                   0.0
                                                                             2.082720
                           0.0
                                                                   0.0
          3.139931
                                                                             2.480406
      66
                                                                   0.0
      67
          3.763229
                           0.0
                                                                             2.056058
      68
          3.946027
                           0.0
                                                                   0.0
                                                                             2.342662
              Total
      18
          2.015289
      17
          2.042591
      16
          2.064759
      15
          2.063625
      11
          2.051110
      . .
      34
          2.900427
      28
          2.475825
      66
          2.744405
      67
          2.294809
      68
          2.777552
```

[70 rows x 9 columns]

### 6 Additional Characteristics

6.1 1. Count number of floors in a given building based on position relative to the ground.

```
[52]: from collections import Counter import re import eeweather BUILDING_ID = '043' #As an example, select building 043 building_data = df.loc[BUILDING_ID]
```

```
[53]: seen = set()
      c = Counter()
      for k,v in building data.items():
          floor = k.split('_')[0]
          if floor in seen or v!=v or 'kg' not in k:
              continue
          seen.add(floor)
      for x in seen:
          parts = re.split('([A-Z])',x)
          parts = [p for p in parts if p!='']
          parts = [int(p) if p.isdigit() else p for p in parts]
          if 'B' in parts:
              c.update(['Basement'])
          elif 'R' in parts:
              c.update(['Roof'])
          elif 0 in parts:
              c.update(['Ground'])
          else:
              c.update(['Above Ground'])
      print(f'Floors relative to ground for building {BUILDING_ID}:')
      for k,v in c.items():
          print(f'{k}: {v} floor(s)')
```

Floors relative to ground for building 043:

Above Ground: 2 floor(s)
Basement: 1 floor(s)
Ground: 1 floor(s)
Roof: 1 floor(s)

### 6.2 2. Get climate conditions for a given building

This code retrieves local climate zones for a given building.

```
[54]: from geopy.geocoders import Nominatim
  locator = Nominatim(user_agent="ConstructionDataset")
  name_map = {
    'TOR':'Toronto',
```

```
'WIN':'Winnipeg',
          'NEW': 'New York',
          'RIC': 'Richmond',
          'MIS':'Mississuaga'
      }
      location = locator.geocode(f'{name_map[building_data.City]}, {building_data.
       →Country}')
[55]: ranked_stations = eeweather.rank_stations(location.latitude,location.longitude)
      ranked_stations = ranked_stations[~ranked_stations.iecc_climate_zone.isnull()]
      station, warnings = eeweather.select_station(ranked_stations)
[56]: print(f'Climate zones for building {BUILDING_ID}:')
      for k,v in station.climate_zones.items():
          if v is None:
              continue
          print(f'{k}: {v}')
     Climate zones for building 043:
     iecc_climate_zone: 5
     iecc_moisture_regime: A
     ba_climate_zone: Cold
 []:
```