Sample

April 27, 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.

```
[2]: headings = ['Building Identifier',
                  'Country',
                   'City',
                   'Quality / Stage of Data',
                   'Construction Date',
                   'Building Type',
                   'Gross Floor Area']
[3]: df = pd.read_excel('../Dataset/dataset.xlsx',header=1).drop('Unnamed: 0',axis=1)
[4]: df
[4]:
         Building Identifier Country City Quality / Stage of Data \
     0
                            1
                                   CA
                                       TOR
                                                              OOIFC
     1
                            2
                                   CA
                                       TOR
                                                              OOIFC
     2
                            3
                                       TOR
                                   CA
                                                              OOIFC
     3
                            4
                                   CA
                                       TOR
                                                              00IFC
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                                   CA
                                       TOR
                                                              OOIFC
     5
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                                   CA
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                                   CA TOR
                                                              OOIFC
                                      TOR
     7
                            8
                                   CA
                                                              00IFC
                            9
                                       TOR
                                                              OOIFC
     8
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     9
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     10
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                                   CA
                                       TOR
                                                              00IFC
     11
                           12
                                   CA
                                       TOR
                                                              00IFC
```

12	13	CA	TOR	00IFC
13	14	CA	TOR	OOIFC
14	15	CA	TOR	OOIFC
15	16	CA	TOR	OOIFC
16	17	CA	TOR	00IFC
17	18	CA	TOR	OOIFC
18	19	CA	TOR	OOIFC
19	20	CA	TOR	00IFC
20	21	CA	TOR	00IFC
21	22	CA	TOR	OOIFC
22	23	CA	TOR	OOIFC
23	24	CA	TOR	OOIFC
24	25	CA	TOR	OOIFC
25	26	CA	TOR	OOIFC
26	27	CA	WIN	OOIFC
27	28	CA	TOR	OOIFC
28	29	CA	TOR	OOIFC
29	30	CA	TOR	OOIFC
30	31	CA	TOR	OOIFC
31	32	CA	TOR	OOIFC
32	33	CA	TOR	OOIFC
33	34	CA	TOR	00IFC
34	3 4 35	CA		001FC 001FC
			TOR	
35	36	CA	TOR	00IFC
36	37	CA	TOR	00IFC
37	38	CA	TOR	00IFC
38	39	CA	TOR	OOIFC
39	40	US	NEW	00IFC
40	41	CA	TOR	00IFC
41	42	CA	TOR	OOIFC
42	43	CA	TOR	OOIFC
43	44	CA	TOR	OOIFC
44	45	CA	TOR	00IFC
45	46	CA	TOR	00IFC
46	47	CA	TOR	OOIFC
47	48	CA	RIC	OIARC
48	49	CA	TOR	OOIFC
49	50	CA	TOR	OOIFC
50	51	CA	TOR	00IFC
51	52	CA	TOR	OOIFC
52	53	CA	TOR	OOIFC
53	54	CA	TOR	OOIFC
54	55	CA	TOR	OOIFC
55	56	CA	TOR	OOIFC
56	57	CA	TOR	OOIFC
57	58	CA	TOR	OOIFC
58	59	CA	TOR	OIFBP

\

	Construction	Date	Building	Туре	Gross Floor Area
0		2021		SND	521.18
1		2021		SND	389.24
2		2021		SND	411.64
3		2021		SND	269.56
4		2011		OFF	11248.00
5		2011		APB	11317.00
6		2021		SND	445.99
7		2021		SND	438.45
8		2021		SND	714.07
9		2021		SND	343.24
10		2009		OFF	73083.00
11		1917		SMR	199.93
12		2021		SND	226.89
13		2021		SND	611.73
14		2021		SND	343.44
15		2021		SND	613.38
16		1969		SNR	413.72
17		1969		SNR	333.49
18		2021		SND	178.38
19		2021		SND	323.80
20		2020		SND	837.56
21		2021		SND	587.86
22		2021		SND	568.21
23		2021		SMD	234.73
24		2021		SND	294.84
25		2021		SND	496.77
26		2007		OFF	73600.00
27		2021		SND	643.30
28		2021		SND	701.61
29		2021		SMD	257.75
30		2021		SND	378.70
31		2021		SND	324.16
32		2020		SND	533.53
33		2020		SMD	254.05
34		2021		SND	423.03
35		2021		SND	328.16
36		2021		SND	421.59
37		2020		SND	628.59
38		2021		SND	464.51
39		2017		EDU	8983.00
40		2021		SND	346.14
41		1913		SNR	161.08
42		2021		SND	891.97
43		2021		SND	525.61

44	2021	SND	502.87		
45	2021	SND	379.18		
46	2021	SND	549.65		
47	2016	EDU	6819.00		
48	2020	SND	393.82		
49	2021	SND	648.14		
50	1988	INS	21934.00		
51	2018	APB	53146.02		
52	2018	MIX	33975.25		
53	2017	APB	69784.00		
54	2017	APB	39409.04		
55	2016	APB	53871.00		
56	2020	LNW	137.23		
57	2020	LNW	144.92		
58	2019	LNW	83.10		
59	2021	LNW	234.79		
	000_G2010.20.000_03 00	00.00_kg_1	000_B1010.20.000_03 (00 00.00_kg_1	\
0		NaN		NaN	
1		NaN		NaN	
2		NaN		NaN	
3		NaN		NaN	
4		13704.0		1.776816e+06	
5		NaN		1.514400e+06	
6		NaN		NaN	
7		NaN		NaN	
8		NaN		NaN	
9		NaN		NaN	
10		58008.0		4.029264e+06	
11		NaN		NaN	
12		NaN		NaN	
13		NaN		NaN	
14		NaN		NaN	
15		NaN		NaN	
16		NaN		NaN	
17		NaN		NaN	
18		NaN		NaN	
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21		NaN		NaN	
22		NaN		NaN	
23		NaN		NaN	
24		NaN		NaN	
25		NaN		NaN	
26		NaN		4.480680e+06	
27		NaN N-N		NaN	
28		NaN		NaN	

29	NaN		NaN	
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31	NaN		NaN	
32	NaN		NaN	
33	NaN		NaN	
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35	NaN		NaN	
36	NaN		NaN	
37	NaN		NaN	
38	NaN		NaN	
39	NaN		2.191431e+04	
40	NaN		NaN	
41	NaN		NaN	
42	NaN		NaN	
43	NaN		NaN	
44	NaN		NaN	
45	NaN		NaN	
46	NaN		NaN	
47	NaN		3.756000e+04	
48	NaN		NaN	
49	NaN		NaN	
50	NaN		NaN	
51	NaN		NaN	
52	NaN		NaN	
53	NaN		NaN	
54	NaN		NaN	
55	NaN		NaN	
56	NaN		NaN	
57	NaN		NaN	
58	NaN		NaN	
59	NaN		NaN	
	000_C1010.10.000_04 22 00.00_kg_1		000_B2010.10.000_07 46 16.00_kg_2	\
0	NaN	•••	NaN	
1	NaN		NaN	
2	NaN	•••	NaN	
3	NaN		NaN	
4	19397.560000	•••		
		•••	NaN	
5	53877.650000	•••	NaN	
6	NaN	•••	NaN	
7	NaN	•••	NaN	
8	NaN	•••	NaN	
9	NaN	•••	NaN	
10	562574.500000	•••	NaN	
11	NaN		NaN	
12	NaN	•••	NaN	
13	NaN	•••	NaN	
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14	NaN	•••	NaN
15	NaN	•••	NaN
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17	NaN	•••	NaN
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25	NaN		NaN
26	354208.227500		NaN
27	NaN		NaN
28	NaN		NaN
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32	NaN		NaN
33	NaN		NaN
34	NaN		NaN
35	NaN		NaN
36	NaN		NaN
37	NaN	•••	NaN
38	NaN	•••	NaN
39	8666.292723		NaN
40	NaN		NaN
41	NaN	•••	NaN
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43	NaN	•••	NaN
44	NaN	•••	NaN
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49	NaN	•••	NaN
50	NaN	•••	NaN
51	8194.250000	•••	NaN
52	191988.905000	•••	NaN
53	82694.400000		NaN
54	46298.790000	•••	NaN
55	422839.793489	•••	NaN
56	NaN	•••	NaN
57	NaN	•••	NaN
58	NaN	•••	NaN
59	NaN	•••	67.3

	001_B2010.80.000_07 27 00.00_kg_2	001 B2010 80 000 07 21 13 00 kg 2	\
0	NaN	NaN	`
1	NaN	NaN	
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4	NaN	NaN	
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6	NaN	NaN	
7	NaN	NaN	
8	NaN	NaN Nan	
9 10	NaN NaN	NaN NaN	
11	NaN NaN	NaN NaN	
12	NaN	NaN	
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14	NaN	NaN	
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16	NaN	NaN	
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19	NaN	NaN	
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21	NaN	NaN	
22	NaN	NaN	
23	NaN	NaN	
24	NaN	NaN	
25	NaN	NaN	
26	NaN	NaN Na N	
27 28	NaN NaN	NaN NaN	
29	NaN	NaN	
30	NaN	NaN	
31	NaN	NaN	
32	NaN	NaN	
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34	NaN	NaN	
35	NaN	NaN	
36	NaN	NaN	
37	NaN	NaN	
38	NaN	NaN	
39	NaN	NaN	
40	NaN	NaN	
41	NaN	NaN	
42	NaN	NaN Na N	
43	NaN NaN	NaN NaN	
44 45	NaN Nan	NaN NaN	
45	NaN	NaN	

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57	NaN	NaN	
58	NaN	NaN	
59	37.3	112.67	
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31	NaN	NaN
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48	NaN	NaN
49	NaN	NaN
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51	NaN	NaN
52	NaN	NaN
53	NaN	NaN
54	NaN	NaN
55	NaN	NaN
56	NaN	NaN
57	NaN	NaN
58	NaN	NaN
59	2655.54	277.59
	OB1_A5020.10.000_06 11 00.00_kg_1	OB1_A5020.10.000_09 21 16.00_kg_1 \
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52	NaN	NaN
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54	NaN	NaN
55	NaN	NaN
56	NaN	NaN
57	NaN	NaN
58	NaN	NaN
59	889.66	854.98
	300.00	331.00
	000 01010 10 000 07 01 12 00 1 1	OOD B3010 OO OOO O7 O1 12 OO 1 1
0	000_C1010.10.000_07 21 13.00_kg_1	00R_B3010.90.000_07 21 13.00_kg_1
0	NaN	NaN

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1	NaN	NaN
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54	NaN	NaN
55	NaN	NaN
56	NaN	NaN
57	NaN	NaN
58	NaN	NaN
59	127.47	420.29
	00R_B1020.20.000_07 51 13.00_kg_1	
0	naN	
1	NaN	
2	NaN	
3	NaN	
4	NaN	
5	NaN	
6	NaN	
7	NaN	
8	NaN	
9	NaN	
10	NaN	
11	NaN	
12	NaN	
13	NaN	
14	NaN	
15	Nan Nan	
16	NaN NaN	
17	NaN	
18 19	NaN NaN	
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	NaN Na N	
21	NaN Nan	
22	NaN Nan	
23	NaN Nan	
24	NaN Nan	
25	NaN Na N	
26	NaN Na N	
27	NaN Na N	
28	NaN Nan	
29	NaN	
30	NaN	
31	NaN	
32	NaN	

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     58
                                        NaN
     59
                                     315.22
     [60 rows x 2090 columns]
[5]: name_conversion = pd.read_csv('name_conversion.csv')
     building_name_conversion = pd.read_csv('building_type_name_conversion.csv')
[6]: building_name_map = {k['Building Code']:k['Building Type'] for _,k in_
      →building_name_conversion.iterrows()}
[7]: name_map = {k.Code:k.Category for _,k in name_conversion.iterrows()}
[8]: 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',
```

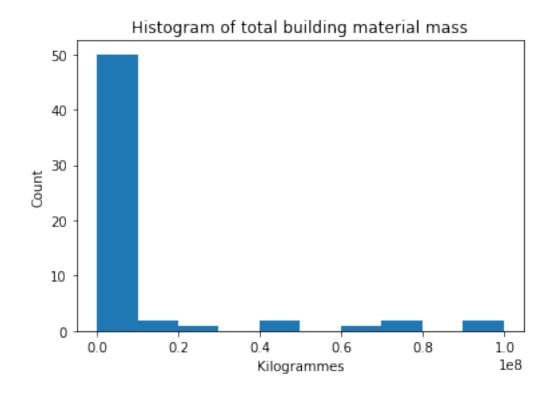
```
'Columns Supporting Roofs': 'CSR',
    'Roof Girders and Beams': 'RGB',
    'Roof Trusses':'ORT',
    'Roof Joists':'ORJ',
    'Parking Bumpers':'OPB',
    'Precast Concrete Stair Treads': 'PCS',
    'Roof Curbs':'ORC',
    'Exterior Wall Construction': 'EWC',
    'Composite Decking':'CPD',
    'Cast-in-Place 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.

```
[9]: 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.xlabel('Kilogrammes')
plt.ylabel('Count');
```

[9]: Text(0, 0.5, 'Count')



3 2. Investigate a specific material

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

```
[10]: cols = [d for d in df.columns if '03 31 00' in d]
[11]: 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)
[12]:
      concrete_df
[12]:
          Building Identifier Country City Quality / Stage of Data \
      0
                                    CA
                                        TOR
                                                                OOIFC
      1
                             2
                                    CA
                                        TOR
                                                                OOIFC
                             3
      2
                                    CA
                                        TOR
                                                                OOIFC
      3
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                                    CA
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                                        TOR
                                                                OOIFC
                                    CA
      6
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                                    CA
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      7
                                        TOR
                             8
                                    CA
                                                                OOIFC
                             9
                                    CA
                                        TOR
                                                                OOIFC
```

9	10	CA	TOR	OOIFC
10	11	CA	TOR	OOIFC
11	12	CA	TOR	OOIFC
12	13	CA	TOR	OOIFC
13	14	CA	TOR	OOIFC
14	15	CA	TOR	OOIFC
15	16	CA	TOR	OOIFC
16	17	CA	TOR	OOIFC
17	18	CA	TOR	OOIFC
18	19	CA	TOR	OOIFC
19	20	CA	TOR	OOIFC
20	21	CA	TOR	OOIFC
21	22	CA	TOR	OOIFC
22	23	CA	TOR	OOIFC
23	24	CA	TOR	OOIFC
24	25	CA	TOR	OOIFC
25	26	CA	TOR	OOIFC
26	27	CA	WIN	OOIFC
27	28	CA	TOR	OOIFC
28	29	CA	TOR	OOIFC
29	30	CA	TOR	OOIFC
30	31	CA	TOR	OOIFC
31	32	CA	TOR	OOIFC
32	33	CA	TOR	OOIFC
33	34	CA	TOR	00IFC
34	35	CA	TOR	00IFC
35	36	CA	TOR	00IFC
36	37	CA	TOR	OOIFC
37	38	CA	TOR	OOIFC
38	39	CA	TOR	00IFC
39	40	US	NEW	OOIFC
40	41	CA	TOR	OOIFC
41	42	CA	TOR	OOIFC
42	43	CA	TOR	OOIFC
43	44	CA	TOR	OOIFC
44	45	CA	TOR	OOIFC
45	46	CA	TOR	00IFC
46	47	CA	TOR	00IFC
47	48	CA	RIC	OIARC
48	49	CA	TOR	OOIFC
49	50	CA	TOR	OOIFC
50	51	CA	TOR	OOIFC
51	52	CA	TOR	OOIFC
52	53	CA	TOR	00IFC
53	54	CA	TOR	00IFC
54	55	CA	TOR	OOIFC
55	56	CA	TOR	OOIFC

56 57 58 59		57 CA 58 CA 59 CA 60 CA	TOR TOR			00IFC 00IFC 0IFBP 0IFBP	
0 1 2 3 4 5 6 7 8 9 10 11 12	Construction Date 2021 2021 2021 2021 2021 2011 2011 2021 2021 2021 2021 2021 2021 2021 2020		SND SND SND OFF APB SND	Gross	Floor Area 521.18 389.24 411.64 269.56 11248.00 11317.00 445.99 438.45 714.07 343.24 73083.00 199.93 226.89	Foundations 1.709236e+05 1.082862e+05 1.909299e+05 6.736923e+04 0.000000e+00 1.295202e+05 1.174431e+05 1.927680e+05 9.564723e+04 0.000000e+00 9.927316e+04 5.835472e+04	\
13 14 15 16 17 18	2021 2021 2021 1969 1969 2021		SND SND SNR SNR SNR		611.73 343.44 613.38 413.72 333.49 178.38	2.061282e+05 1.436814e+05 1.789777e+05 9.293583e+04 1.186380e+05 6.408230e+04	
19 20 21 22 23	2021 2020 2021 2021 2021		SND SND SND SND SMD		323.80 837.56 587.86 568.21 234.73	4.733438e+04 2.605656e+05 2.455371e+05 1.415184e+05 8.560216e+04	
242526272829	2021 2021 2007 2021 2021 2021		SND OFF SND SND SMD		294.84 496.77 73600.00 643.30 701.61 257.75	9.718853e+04 1.810933e+05 8.183304e+04	
30 31 32 33 34 35	2021 2021 2020 2020 2021 2021		SND SND SND SMD SND SND		378.70 324.16 533.53 254.05 423.03 328.16	1.477228e+05 1.188635e+05 1.627046e+05 8.882102e+04 9.980270e+04 1.238544e+05	
36 37 38 39 40	2021 2020 2021 2017 2021		SND SND SND EDU SND		421.59 628.59 464.51 8983.00 346.14	1.760423e+05 2.298828e+05 1.886381e+05 0.000000e+00 9.748630e+04	

41		1913	SNR	161.08	5.362299e+04
42		2021	SND	891.97	2.157609e+05
43		2021	SND	525.61	2.567725e+05
44		2021	SND	502.87	1.372402e+05
45		2021	SND	379.18	1.437386e+05
46		2021	SND	549.65	1.435894e+05
47		2016	EDU	6819.00	0.000000e+00
48		2020	SND	393.82	7.294707e+04
49		2021	SND	648.14	2.216331e+05
50		1988	INS	21934.00	0.000000e+00
51		2018	APB	53146.02	1.115822e+07
52		2018	MIX	33975.25	4.220040e+06
53		2017	APB	69784.00	7.912944e+06
54		2017	APB	39409.04	9.350736e+06
55		2016	APB	53871.00	1.627512e+06
56		2020	LNW	137.23	3.111394e+04
57		2020	LNW	144.92	3.241172e+04
58		2019	LNW	83.10	3.347723e+04
59		2021	LNW	234.79	8.400714e+04
					0.100.110 01
	Subgrade End	clasuras	Slabs-On-Grade	Substructure	Interior \
^	pubgrade Lin			Dubbulactare	•
0		0.0	6.721219e+04		0.0
1		0.0	3.576043e+04		0.0
2		0.0	3.246461e+04		0.0
3		0.0	1.595211e+04		0.0
4		0.0	0.000000e+00		0.0
5		0.0	0.000000e+00		0.0
6		0.0	3.521918e+04		0.0
7		0.0	4.289057e+04		0.0
8		0.0	8.446873e+04		0.0
9		0.0	2.033114e+04		0.0
10					
11		0.0	0.000000e+00		0.0
		0.0	0.000000e+00 1.971760e+04		
12					0.0
12 13		0.0	1.971760e+04		0.0
13		0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04		0.0 0.0 0.0 0.0
13 14		0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04		0.0 0.0 0.0 0.0 0.0
13 14 15		0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04 4.219445e+04		0.0 0.0 0.0 0.0 0.0
13 14 15 16		0.0 0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04 4.219445e+04 3.376814e+04		0.0 0.0 0.0 0.0 0.0 0.0
13 14 15 16 17		0.0 0.0 0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04 4.219445e+04 3.376814e+04 2.622366e+04		0.0 0.0 0.0 0.0 0.0 0.0
13 14 15 16 17 18		0.0 0.0 0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04 4.219445e+04 3.376814e+04 2.622366e+04 2.343862e+04		0.0 0.0 0.0 0.0 0.0 0.0 0.0
13 14 15 16 17		0.0 0.0 0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04 4.219445e+04 3.376814e+04 2.622366e+04		0.0 0.0 0.0 0.0 0.0 0.0
13 14 15 16 17 18		0.0 0.0 0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04 4.219445e+04 3.376814e+04 2.622366e+04 2.343862e+04		0.0 0.0 0.0 0.0 0.0 0.0 0.0
13 14 15 16 17 18 19		0.0 0.0 0.0 0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04 4.219445e+04 3.376814e+04 2.622366e+04 2.343862e+04 2.368485e+04		0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0
13 14 15 16 17 18 19 20 21		0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04 4.219445e+04 3.376814e+04 2.622366e+04 2.343862e+04 2.368485e+04 6.344851e+04 6.865710e+04		0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0
13 14 15 16 17 18 19 20 21 22		0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04 4.219445e+04 3.376814e+04 2.622366e+04 2.343862e+04 2.368485e+04 6.344851e+04 6.865710e+04 6.684690e+04		0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0
13 14 15 16 17 18 19 20 21 22 23		0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04 4.219445e+04 3.376814e+04 2.622366e+04 2.343862e+04 2.368485e+04 6.344851e+04 6.865710e+04 1.294360e+04		0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0
13 14 15 16 17 18 19 20 21 22		0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0	1.971760e+04 1.435987e+04 4.140039e+04 2.246836e+04 4.219445e+04 3.376814e+04 2.622366e+04 2.343862e+04 2.368485e+04 6.344851e+04 6.865710e+04 6.684690e+04		0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0

26	0.0	0.00000e	+00	0.0
27	0.0	5.230228e	+04	0.0
28	0.0	6.233222e	+04	0.0
29	0.0	1.211886e	+04	0.0
30	0.0	3.514722e	+04	0.0
31	0.0	2.011968e	+04	0.0
32	0.0	3.674638e	+04	0.0
33	0.0	1.160387e	+04	0.0
34	0.0	3.329286e	+04	0.0
35	0.0	1.931159e	+04	0.0
36	0.0	3.304437e	+04	0.0
37	0.0	5.528816e	+04	0.0
38	0.0	2.866777e	+04	0.0
39	0.0	0.00000e	+00	0.0
40	0.0	2.237098e	+04	0.0
41	0.0	1.235658e	+04	0.0
42	0.0	5.949332e	+04	0.0
43	0.0	3.378685e	+04	0.0
44	0.0	3.951047e		0.0
45	0.0	2.913799e		0.0
46	0.0	3.506390e		0.0
47	0.0	0.00000e		0.0
48	0.0	3.364275e		0.0
49	0.0	6.099032e		0.0
50	0.0	0.000000e		0.0
51	2728008.0	3.647520e		11033448.0
52	1705680.0	3.834720e		5400288.0
53	3246168.0	1.407000e		14052000.0
54	3567720.0	9.045840e		7607280.0
55 56	3438168.0	7.174800e		22907184.0
56	0.0	1.439848e		0.0
57 50	0.0	2.000253e		0.0
58 50	0.0	5.412759e+ 1.962799e+		0.0
59	0.0	1.9627996	FU4	0.0
	Substructure Related A	Activities	Superstructure	\
0	Substitution netated is	0.0	1.938810e+03	`
1		0.0	1.397610e+03	
2		0.0	1.528710e+02	
3		0.0	1.212090e+01	
4		0.0	0.000000e+00	
5		0.0	0.000000e+00	
6		0.0	5.332590e+02	
7		0.0	1.970790e+03	
8		0.0	4.049670e+03	
9		0.0	9.440170e+02	
10		0.0	0.000000e+00	

11	0.0	0.000000e+00
12	0.0	9.785830e+02
13	0.0	5.381500e+02
14	0.0	0.000000e+00
15	0.0	0.000000e+00
16	0.0	0.000000e+00
17	0.0	7.514840e+03
18	0.0	0.000000e+00
19	0.0	2.111800e+03
20	0.0	3.270810e+03
21	0.0	2.533580e+03
22	0.0	6.016340e+02
23		1.827610e+03
	0.0	
24	0.0	5.977480e+02
25	0.0	2.540900e+03
26	0.0	0.000000e+00
27	0.0	7.189470e+02
28	0.0	2.276420e+02
29	0.0	1.587900e+03
30	0.0	1.096510e+04
31	0.0	5.530400e+03
32	0.0	1.360980e+03
33	0.0	2.177290e+03
34	0.0	6.524310e+02
35	0.0	3.944150e+03
36	0.0	4.401230e+02
37	0.0	8.518740e+02
38	0.0	2.593160e+03
39	0.0	0.000000e+00
40	0.0	2.360810e+02
41	0.0	0.000000e+00
42	0.0	8.599660e+02
43	0.0	1.038810e+03
44	0.0	4.881840e+02
45	0.0	1.267510e+03
46	0.0	1.154890e+03
47	0.0	0.000000e+00
48	0.0	1.835120e+02
49	0.0	1.041320e+03
50	0.0	0.000000e+00
51	133464.0	2.780006e+07
52	112872.0	2.226535e+07
53	169896.0	3.204622e+07
54	276264.0	1.483577e+07
55	93048.0	3.239134e+07
56	0.0	0.000000e+00
57	0.0	0.000000e+00
٠.	5.0	1.300000.00

58 59				.0		00000e+00 00000e+00		
	Exterior	Vertical	Enclosures	Exte	rior	Horizontal	Enclosures	\
0			0.0				0.0	·
1			0.0				0.0	
2			0.0				0.0	
3			0.0				0.0	
4			0.0				0.0	
5			0.0				0.0	
6			0.0				0.0	
7			0.0				0.0	
8			0.0				0.0	
9			0.0				0.0	
10			0.0				0.0	
11			0.0				0.0	
12			0.0				0.0	
13			0.0				0.0	
14			0.0				0.0	
15 16			0.0				0.0	
17			0.0				0.0	
18			0.0				0.0	
19			0.0				0.0	
20			0.0				0.0	
21			0.0				0.0	
22			0.0				0.0	
23			0.0				0.0	
24			0.0				0.0	
25			0.0				0.0	
26			0.0				0.0	
27			0.0				0.0	
28			0.0				0.0	
29			0.0				0.0	
30			0.0				0.0	
31			0.0				0.0	
32			0.0				0.0	
33			0.0				0.0	
34			0.0				0.0	
35 36			0.0				0.0	
36 37			0.0				0.0	
38			0.0				0.0	
39			0.0				0.0	
40			0.0				0.0	
41			0.0				0.0	
42			0.0				0.0	

43			0.0		0.0	
44			0.0		0.0	
45			0.0		0.0	
46			0.0		0.0	
47			0.0		0.0	
48			0.0		0.0	
49			0.0		0.0	
50			0.0		0.0	
51		72	7896.0		537984.0	
52		40	5408.0		392400.0	
53		32	8032.0		799872.0	
54		11	9088.0		0.0	
55		15	9336.0		0.0	
56			0.0		0.0	
57			0.0		0.0	
58			0.0		0.0	
59			0.0		0.0	
	Interior	Construction		Plumbing	-	\
0		0.0	0.0	0.0	0.0	
1		0.0	0.0	0.0	0.0	
2		0.0	0.0	0.0	0.0	
3		0.0	0.0	0.0	0.0	
4		0.0	0.0	0.0	0.0	
5		0.0	0.0	0.0	0.0	
6		0.0	0.0	0.0	0.0	
7		0.0	0.0	0.0	0.0	
8 9		11307.2	0.0	0.0	0.0	
		0.0	0.0	0.0	0.0	
10 11		0.0	0.0	0.0	0.0	
12		0.0	0.0	0.0	0.0	
13		0.0	0.0	0.0	0.0	
14		0.0	0.0	0.0	0.0	
15		0.0	0.0	0.0	0.0	
16		0.0	0.0	0.0	0.0	
17		0.0	0.0	0.0	0.0	
18		0.0	0.0	0.0	0.0	
19		0.0	0.0	0.0	0.0	
20		0.0	0.0	0.0	0.0	
21		0.0	0.0	0.0	0.0	
22		0.0	0.0	0.0	0.0	
23		0.0	0.0	0.0	0.0	
24		0.0	0.0	0.0	0.0	
25		0.0	0.0	0.0	0.0	
26		0.0	0.0	0.0	0.0	
27		0.0	0.0	0.0	0.0	
				0.0	0.0	

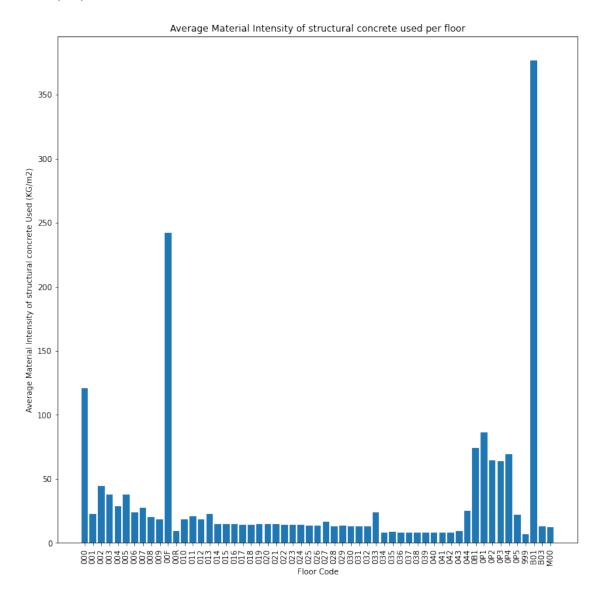
28	0.0	0.0	0.0	0.0
29	0.0	0.0	0.0	0.0
30	0.0	0.0	0.0	0.0
31	0.0	0.0	0.0	0.0
32	0.0	0.0	0.0	0.0
33	0.0	0.0	0.0	0.0
34	0.0	0.0	0.0	0.0
35	0.0	0.0	0.0	0.0
36	0.0	0.0	0.0	0.0
37	0.0	0.0	0.0	0.0
38	0.0	0.0	0.0	0.0
39	0.0	0.0	0.0	0.0
40	0.0	0.0	0.0	0.0
41	0.0	0.0	0.0	0.0
42	0.0	0.0	0.0	0.0
43	0.0	0.0	0.0	0.0
44	0.0	0.0	0.0	0.0
45	0.0	0.0	0.0	0.0
46	0.0	0.0	0.0	0.0
47	0.0	0.0	0.0	0.0
48	0.0	0.0	0.0	0.0
49	0.0	0.0	0.0	0.0
50	0.0	0.0	0.0	0.0
51	6816696.0	2494560.0	0.0	80592.0
52	5893176.0	1829328.0	48816.0	62280.0
53	9050592.0	2304480.0	172032.0	0.0
54	5180976.0	861888.0	130152.0	0.0
55	5604960.0	1664448.0	0.0	220992.0
56	0.0	0.0	0.0	0.0
57	0.0	0.0	0.0	0.0
58	0.0	0.0	0.0	0.0
59	0.0	0.0	0.0	0.0

Site Improvements

	1
0	0.0
1	0.0
2	0.0
3	0.0
4	0.0
5	0.0
6	0.0
7	0.0
8	0.0
9	0.0
10	0.0
11	0.0
12	0.0

13	0.0
14 15	0.0
16	0.0
17	0.0
18 19	0.0
20	0.0
21	0.0
22 23	0.0
24	0.0
25 26	0.0
27	0.0
28	0.0
29 30	0.0
31	0.0
32	0.0
33 34	0.0
35	0.0
36	0.0
37 38	0.0
39	0.0
40 41	0.0
42	0.0
43	0.0
44 45	0.0
46	0.0
47	0.0
48 49	0.0
50	0.0
51	0.0
52 53	0.0 18384.0
54	97560.0
55 56	0.0
57	0.0
58	0.0
59	0.0

[13]: Text(0.5, 0, 'Floor Code')



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

```
[14]: f = lambda x: name_map[re.split('[_\.\]',x)[1][0:3]] #This function takes in a__

-full column name and returns only the Level 3 MasterFormat code.

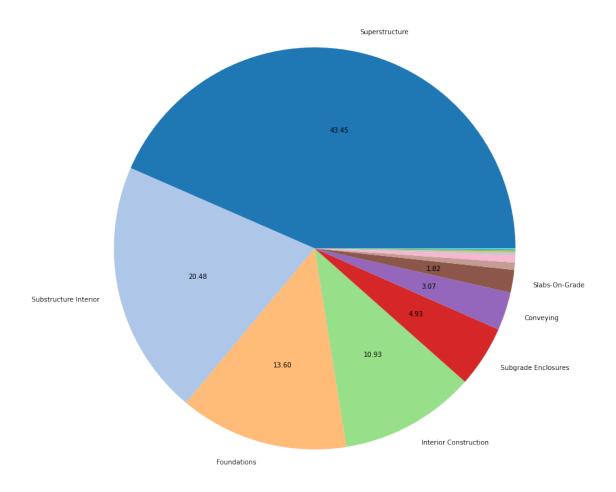
concrete_df = df[cols].groupby(f,axis=1).sum()
```

```
[15]: concrete_df.mean().sort_values(ascending=False)
```

```
[15]: Superstructure
                                         2.156826e+06
      Substructure Interior
                                         1.016670e+06
      Foundations
                                         6.750260e+05
      Interior Construction
                                         5.426285e+05
      Subgrade Enclosures
                                         2.447624e+05
      Conveying
                                         1.525784e+05
      Slabs-On-Grade
                                         9.043012e+04
     Exterior Vertical Enclosures
                                         2.899600e+04
     Exterior Horizontal Enclosures
                                         2.883760e+04
     Substructure Related Activities
                                         1.309240e+04
     Special Construction
                                         6.064400e+03
     Plumbing
                                         5.850000e+03
     Site Improvements
                                         1.932400e+03
      dtype: float64
```

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

```
[16]: 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_\]
        if v > 30000 else '' for k,v in to_plot.items()])
    plt.ylabel('')
    plt.title('Percentage of total steel used in each building element category');
    # plt.legend(loc='center left',bbox_to_anchor=(-0.20, 0.75));
    plt.tight_layout();
```



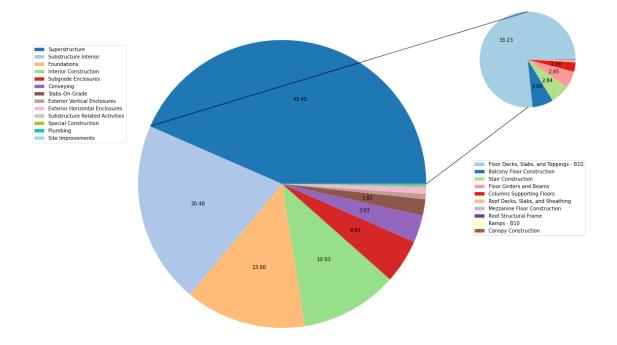
3.2 Pie version B: external legend with slice labels

```
[17]: 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]] \
   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='Paired',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')
```



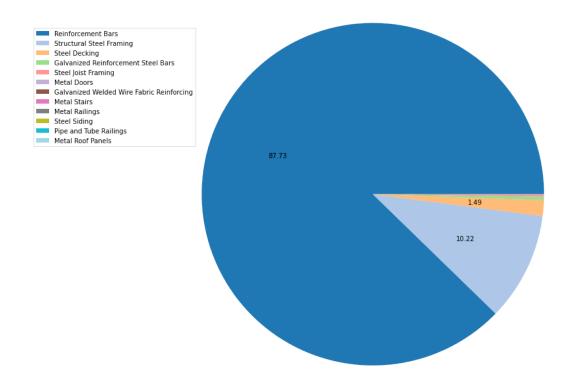
We can produce a pie chart for a single building, also.

```
[18]: mf_codes = pd.read_csv('mf_name_conversion.csv')
[19]: 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)]
```

```
[20]: 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)
[21]: f = lambda x: mf_codes[mf_codes.Code == str.replace(re.split('_',x)[2],'00','').

strip('.')].values[0][0]
      steel_df = df[columns_to_keep].groupby(f,axis=1).sum()
[22]: (steel_df>0).sum(axis=1).sort_values()
[22]: 15
            1
      42
            1
      22
            1
      36
            1
      7
            1
      34
            1
      31
            1
      35
            1
      55
            2
      58
            2
      40
            2
      41
            2
      1
            2
      43
            2
      24
            2
            2
      23
      21
            2
            2
      20
            2
      54
      44
            2
      17
            2
      16
            2
            2
      30
            2
      14
      45
            2
            2
      12
      11
            2
      32
            2
      9
            2
      33
            2
            2
      3
      18
            2
            3
      0
```

```
52
           3
      53
           3
           3
      56
      46
           3
           3
      39
      29
           3
      37
           3
      28
           3
      27
           3
      26
           3
      25
           3
      13
           3
      10
           3
      2
           3
      38
           3
           3
      5
      6
           3
           3
      8
      57
           4
      4
           4
      49
           4
      50
           4
      48
           4
      47
           4
      19
      51
           4
      59
           4
      dtype: int64
[23]: 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 MasterFormatu
      plt.tight_layout();
      plt.savefig('steel_composition_pie.pdf')
```



```
destrip('.')].values[0][0] + '/' + x.split('_')[0]

tdf = df[columns_to_keep].groupby(f,axis=1).sum().iloc[47,:]

tdf = tdf[tdf>0]

[25]:

from collections import defaultdict

todf = defaultdict(dict)

for (a,b),c in zip(tdf.keys().str.split('/'),tdf.values):

    todf[a][b] = c

toplot = pd.DataFrame(todf)

toplot.plot.bar(figsize=(12,12));

plt.xlabel('Floor Number')

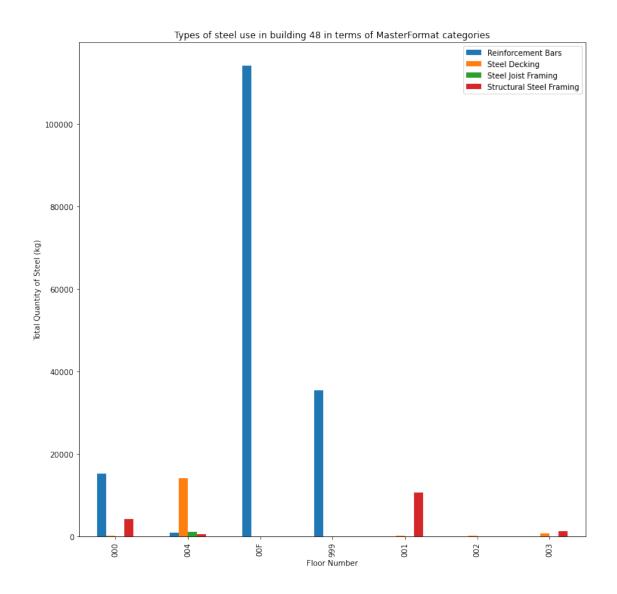
plt.ylabel('Total Quantity of Steel (kg)')

plt.title('Types of steel use in building 48 in terms of MasterFormat_

decategories')

plt.savefig('bar_steel_onebuildingtype_byfloor.pdf')
```

[24]: f = lambda x: mf_codes[mf_codes.Code == str.replace(re.split('_',x)[2],'00','').



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

```
[26]:
                         Gross Floor Area Structural Concrete/000 \
      Construction Date
      1913
                                                       0.000000e+00
                                161.080000
      1917
                                199.930000
                                                       0.000000e+00
      1969
                                373.605000
                                                       0.000000e+00
      1988
                             21934.000000
                                                       0.00000e+00
      2007
                             73600.000000
                                                       0.00000e+00
      2009
                             73083.000000
                                                       0.00000e+00
```

2011 2016 2017 2018 2019 2020 2021	11282.500000 30345.000000 39392.013333 43560.635000 83.100000 418.528571 445.404444	0.000000e+00 3.595656e+06 4.084352e+06 5.893680e+06 0.000000e+00 4.914431e+03 0.000000e+00	
Constantion Date	Structural Concrete/001	Structural Concrete/002	\
Construction Date 1913	1044 380000	0.0	
	1944.380000	0.0	
1917	4972.300000	0.0	
1969	7262.220500	0.0	
1988	0.000000	0.0	
2007	0.000000	0.0	
2009	0.000000	0.0	
2011	0.00000	0.0	
2016	0.00000	2680512.0	
2017	0.00000	989280.0	
2018	0.00000	1511892.0	
2019	0.000000	0.0	
2020	4923.690714	0.0	
2021	11399.123858	0.0	
	Structural Concrete/003	Structural Concrete/004	\
Construction Date	Structural Concrete/003	Structural Concrete/004	\
Construction Date	Structural Concrete/003	Structural Concrete/004	\
			\
1913	0.0	0.0	\
1913 1917	0.0	0.0 0.0	\
1913 1917 1969	0.0 0.0 0.0	0.0 0.0 0.0	\
1913 1917 1969 1988	0.0 0.0 0.0 0.0	0.0 0.0 0.0 0.0	\
1913 1917 1969 1988 2007	0.0 0.0 0.0 0.0	0.0 0.0 0.0 0.0 0.0	\
1913 1917 1969 1988 2007 2009	0.0 0.0 0.0 0.0 0.0	0.0 0.0 0.0 0.0 0.0	\
1913 1917 1969 1988 2007 2009 2011	0.0 0.0 0.0 0.0 0.0 0.0	0.0 0.0 0.0 0.0 0.0 0.0	\
1913 1917 1969 1988 2007 2009 2011 2016	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1686228.0	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1057032.0	\
1913 1917 1969 1988 2007 2009 2011 2016 2017	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1686228.0 1232336.0	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1057032.0 778480.0	\
1913 1917 1969 1988 2007 2009 2011 2016 2017 2018	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1686228.0 1232336.0 1347936.0	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1057032.0 778480.0 1323132.0	\
1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1686228.0 1232336.0 1347936.0	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1057032.0 778480.0 1323132.0 0.0	
1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019 2020	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1686228.0 1232336.0 1347936.0 0.0 0.0	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1057032.0 778480.0 1323132.0 0.0 0.0	\
1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019 2020	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1686228.0 1232336.0 1347936.0 0.0 0.0	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1057032.0 778480.0 1323132.0 0.0 0.0	\
1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019 2020 2021	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1686228.0 1232336.0 1347936.0 0.0 0.0	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1057032.0 778480.0 1323132.0 0.0 0.0	\
1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019 2020 2021	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1686228.0 1232336.0 1347936.0 0.0 0.0	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1057032.0 778480.0 1323132.0 0.0 0.0 0.0	\
1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019 2020 2021 Construction Date 1913 1917	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1686228.0 1232336.0 1347936.0 0.0 0.0 0.0 0.0	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1057032.0 778480.0 1323132.0 0.0 0.0 0.0 0.0	\
1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019 2020 2021 Construction Date 1913 1917 1969	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1686228.0 1232336.0 1347936.0 0.0 0.0 0.0 0.0	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1057032.0 778480.0 1323132.0 0.0 0.0 0.0 0.0 0.0	\
1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019 2020 2021 Construction Date 1913 1917	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1686228.0 1232336.0 1347936.0 0.0 0.0 0.0 0.0	0.0 0.0 0.0 0.0 0.0 0.0 0.0 1057032.0 778480.0 1323132.0 0.0 0.0 0.0 0.0	\

2009 2011 2016 2017 2018 2019 2020 2021	0.0 0.0 1056780.0 683496.0 2164812.0 0.0 0.0	0.0 0.0 1129680.0 679376.0 969060.0 0.0 0.0	
	Structural Concrete/007	Structural Concrete/008	\
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	1809852.0	857976.0	•••
2017	632520.0	651080.0	•••
2018	752208.0	734688.0	•••
2019	0.0	0.0	•••
2020	0.0	0.0	•••
2021	0.0	0.0	•••
	Structural Concrete/OB1	Structural Concrete/OP1	\
Construction Date			
1913	0.000000	0.0	
1917	0.000000	0.0	
1969	0.000000	0.0	
1988	0.000000	0.0	
2007	0.000000	0.0	
2009	0.000000	0.0	
2011			
	0.000000	0.0	
2016	0.000000 0.000000	0.0 2206668.0	
2016 2017		0.0	
	0.000000	0.0 2206668.0	
2017	0.00000 0.00000	0.0 2206668.0 3402456.0	
2017 2018	0.000000 0.000000 0.000000	0.0 2206668.0 3402456.0 3713916.0	
2017 2018 2019	0.000000 0.000000 0.000000 5412.758585	0.0 2206668.0 3402456.0 3713916.0 0.0	
2017 2018 2019 2020	0.000000 0.000000 0.000000 5412.758585 0.000000	0.0 2206668.0 3402456.0 3713916.0 0.0 0.0	\
2017 2018 2019 2020	0.000000 0.000000 0.000000 5412.758585 0.000000 545.221944	0.0 2206668.0 3402456.0 3713916.0 0.0 0.0	\
2017 2018 2019 2020 2021	0.000000 0.000000 0.000000 5412.758585 0.000000 545.221944	0.0 2206668.0 3402456.0 3713916.0 0.0 0.0	\
2017 2018 2019 2020 2021 Construction Date	0.000000 0.000000 0.000000 5412.758585 0.000000 545.221944 Structural Concrete/OP2	0.0 2206668.0 3402456.0 3713916.0 0.0 0.0 0.0 Structural Concrete/OP3	\
2017 2018 2019 2020 2021 Construction Date 1913	0.000000 0.000000 0.000000 5412.758585 0.000000 545.221944 Structural Concrete/OP2	0.0 2206668.0 3402456.0 3713916.0 0.0 0.0 0.0 Structural Concrete/OP3	\
2017 2018 2019 2020 2021 Construction Date 1913 1917	0.000000 0.000000 0.000000 5412.758585 0.000000 545.221944 Structural Concrete/OP2 0.0 0.0	0.0 2206668.0 3402456.0 3713916.0 0.0 0.0 0.0 Structural Concrete/OP3	\

2007 2009 2011 2016 2017 2018 2019 2020 2021	0.0 0.0 0.0 1715028.0 2513320.0 2637060.0 0.0 0.0	0.0 0.0 0.0 1596444.0 2469984.0 2756916.0 0.0 0.0	
	Structural Concrete/OP4	Structural Concrete/OP5	\
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	0.0	0.0	
2017	1895672.0	508328.0	
2018	4093284.0	0.0	
2019	0.0	0.0	
2020	0.0	0.0	
2020	0.0	0.0	
2021	0.0	0.0	
			\
2021 Construction Date	0.0	0.0	\
2021 Construction Date 1913	0.0 Structural Concrete/999 0.0	0.0 Structural Concrete/B01 64035.190000	\
Construction Date 1913 1917	0.0 Structural Concrete/999 0.0 0.0	0.0 Structural Concrete/B01 64035.190000 114018.460000	\
2021 Construction Date 1913	0.0 Structural Concrete/999 0.0	0.0 Structural Concrete/B01 64035.190000	\
Construction Date 1913 1917	0.0 Structural Concrete/999 0.0 0.0	0.0 Structural Concrete/B01 64035.190000 114018.460000	\
2021 Construction Date 1913 1917 1969	0.0 Structural Concrete/999 0.0 0.0 0.0	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000	\
2021 Construction Date 1913 1917 1969 1988	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000	\
2021 Construction Date 1913 1917 1969 1988 2007	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000 0.0000000	\
2021 Construction Date 1913 1917 1969 1988 2007 2009	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0 0.0	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000 0.000000 0.000000	\
2021 Construction Date 1913 1917 1969 1988 2007 2009 2011	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0 0.0 0.0	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000 0.000000 0.000000 0.000000	\
2021 Construction Date 1913 1917 1969 1988 2007 2009 2011 2016	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0 0.0 0.0 0.0 155076.0	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000 0.000000 0.0000000 0.000000	\
2021 Construction Date 1913 1917 1969 1988 2007 2009 2011 2016 2017	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 155076.0 7736.0	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000 0.0000000 0.0000000 0.000000	\
2021 Construction Date 1913 1917 1969 1988 2007 2009 2011 2016 2017 2018	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0 0.0 0.0 0.0	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000 0.000000 0.000000 0.000000 0.000000	\
2021 Construction Date 1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 155076.0 7736.0 558516.0 0.0	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000 0.000000 0.000000 0.000000 0.000000	\
2021 Construction Date 1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019 2020 2021	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 155076.0 7736.0 558516.0 0.0 0.0	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000 0.000000 0.000000 0.000000 0.000000	\
Construction Date 1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019 2020 2021 Construction Date	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0 0.0 0.0 0.0 155076.0 7736.0 558516.0 0.0 0.0 0.0 Structural Concrete/B03	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000 0.000000 0.000000 0.000000 0.000000	\
Construction Date 1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019 2020 2021 Construction Date 1913	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0 0.0 0.0 0.0 155076.0 7736.0 558516.0 0.0 0.0 0.0 Structural Concrete/B03	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000 0.0000000 0.0000000 0.000000	
Construction Date 1913 1917 1969 1988 2007 2009 2011 2016 2017 2018 2019 2020 2021 Construction Date	0.0 Structural Concrete/999 0.0 0.0 0.0 0.0 0.0 0.0 0.0 155076.0 7736.0 558516.0 0.0 0.0 0.0 Structural Concrete/B03	0.0 Structural Concrete/B01 64035.190000 114018.460000 132278.015000 0.000000 0.000000 0.000000 0.000000 0.000000	

1988	0.00000	0.0
2007	0.00000	0.0
2009	0.00000	0.0
2011	0.00000	0.0
2016	0.00000	82056.0
2017	0.00000	0.0
2018	0.00000	597624.0
2019	0.00000	0.0
2020	988.177143	0.0
2021	0.00000	0.0

[13 rows x 58 columns]

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

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

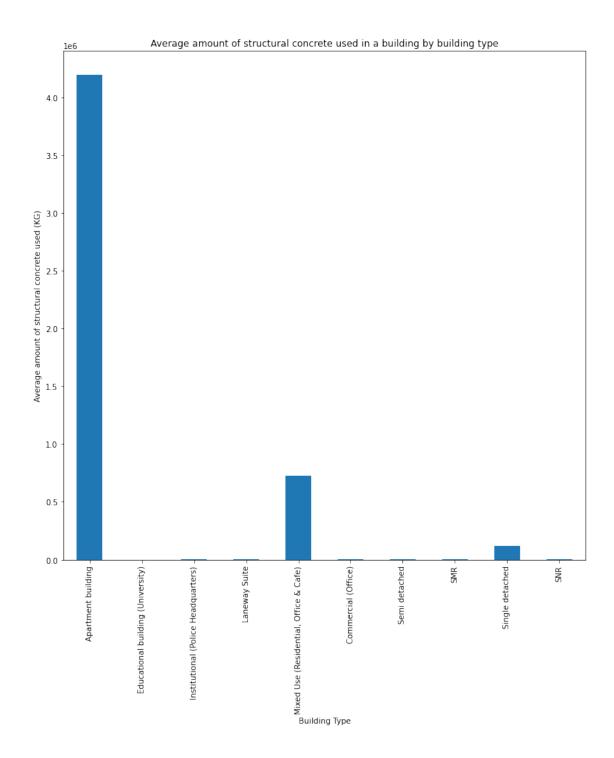
→rename(index=building_name_map).plot(kind='bar',figsize=(12,12))

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

plt.title('Average amount of structural concrete used in a building by building

→type');
```

[27]: Text(0.5, 1.0, 'Average amount of structural concrete used in a building by building type')



4 3. Uncertainty by Building Type

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

```
[28]: 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])
[29]: from collections import Counter
[30]: for k,v in uncertainty_level.items():
          uncertainty_level[k] = Counter(v) #Construct a Counter object per building_
       \hookrightarrow type
[31]: uncertainty_level
[31]: {'SND': Counter({'1': 1662, '2': 641, '4': 293}),
       'OFF': Counter({'1': 494, '3': 307}),
       'APB': Counter({'1': 1171, '2': 1, '3': 971}),
       'SMR': Counter({'1': 21, '2': 27, '4': 8}),
       'SNR': Counter({'1': 58, '2': 70, '4': 56}),
       'SMD': Counter({'1': 170, '2': 34, '4': 19}),
       'EDU': Counter({'1': 93, '3': 24, '2': 6}),
       'INS': Counter({'1': 90, '3': 77, '2': 1}),
       'MIX': Counter({'1': 363, '3': 276}),
       'LNW': Counter({'2': 46, '1': 142, '4': 19})}
     Next, we aggregate columns by the purporse of the material and uncertainty combined, and report
     the average by building type.
[32]: 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 ⊔
      by_function_df = pd.concat([df[headings[1:]],df[cols].groupby(f,axis=1).
       \rightarrowsum()],axis=1)
[33]: by_function_df.groupby('Building Type').mean().rename(index=building_name_map).

¬drop(['Construction Date'],axis=1).round(2)
[33]:
                                               Gross Floor Area Interiors/1 \
      Building Type
      Apartment building
                                                       45505.41
                                                                    5330644.8
      Educational building (University)
                                                        7901.00
                                                                          0.0
      Institutional (Police Headquarters)
                                                                          0.0
                                                       21934.00
      Laneway Suite
                                                                          0.0
                                                          150.01
      Mixed Use (Residential, Office & Cafe)
                                                       33975.25
                                                                    5893176.0
```

Commercial (Office) Semi detached SMR Single detached SNR	52643.67 248.84 199.93 478.40 302.76		0.0 0.0 0.0 305.6 0.0	
Building Type Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite Mixed Use (Residential, Office & Cafe) Commercial (Office) Semi detached SMR Single detached SNR	Services/1 1525512.0 0.0 0.0 0.0 1878144.0 0.0 0.0 0.0 0.0 0.0	21949118.40 0.00 0.00 0.00	0.00 0.00 0.00 13.19	
Building Type Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite Mixed Use (Residential, Office & Cafe) Commercial (Office) Semi detached SMR Single detached SNR	Sitework/1 23188.8 0.0 0.0 0.0 0.0 0.0 0.0 0.0			
Building Type Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite Mixed Use (Residential, Office & Cafe) Commercial (Office) Semi detached SMR Single detached SNR	Special Con	astruction And	603	on/1 \ 16.8
Building Type	Substructur	re/1 Substruc	cture/2 \	

Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite	20539176.00 0.00 0.00 58217.18	0.00 0.00 0.00 44.81
Mixed Use (Residential, Office & Cafe)	11822352.00	0.00
Commercial (Office)	0.00	0.00
Semi detached	97640.84	0.00
SMR	110089.90	8900.86
Single detached	180887.68	5347.87
SNR	93180.79	19334.28
	Substructure/4	
Building Type	Substructure/4	
Building Type Apartment building	Substructure/4 0.00	
~ **	·	
Apartment building	0.00	
Apartment building Educational building (University)	0.00	
Apartment building Educational building (University) Institutional (Police Headquarters)	0.00 0.00 0.00	
Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite	0.00 0.00 0.00 1850.97	
Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite Mixed Use (Residential, Office & Cafe)	0.00 0.00 0.00 1850.97 0.00	
Apartment building Educational building (University) Institutional (Police Headquarters) Laneway Suite Mixed Use (Residential, Office & Cafe) Commercial (Office)	0.00 0.00 0.00 1850.97 0.00 0.00	

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

0.00

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

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

→code (kg)')

display(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().

→round(0))
```

Average amount of material used per building, by year and uncertainty code (kg)

	1	2	3	4
Construction Date				
1913	90840.0	8231.0	0.0	7754.0
1917	137657.0	24862.0	0.0	20786.0
1969	124421.0	90335.0	0.0	32662.0
1988	27569611.0	2134.0	832782.0	0.0
2007	89001799.0	0.0	2715390.0	0.0
2009	96695376.0	0.0	2989172.0	0.0
2011	12969995.0	436848.0	393956.0	0.0
2016	37771429.0	837792.0	988332.0	0.0
2017	40843021.0	0.0	1112868.0	0.0
2018	54609667.0	0.0	1537632.0	0.0

SNR

2019	53403.0	2376.0	0.0	0.0
2020	222826.0	26710.0	0.0	28228.0
2021	243613.0	27693.0	0.0	41449.0

5 4. Material Intensity

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

```
[35]: kilogram_columns = [d for d in df.columns if 'kg' in d]
      df_mi = df[kilogram_columns].div(df['Gross Floor Area'],axis=0)
[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)
      f = lambda x: name_map[re.split('[_\.\ ]',x)[1][0:3]]
      pd.concat([df[headings[1:]],df_mi[kilogram_columns].groupby(f,axis=1).
       [36]:
         Country City Quality / Stage of Data Construction Date Building Type
      0
              CA
                 TOR
                                        OOIFC
                                                             2021
                                                                             SND
                 TOR
                                                             2021
      1
              CA
                                        OOIFC
                                                                             SND
      2
                                                             2021
              CA
                 TOR
                                        OOIFC
                                                                             SND
      3
              CA
                  TOR
                                        OOIFC
                                                             2021
                                                                             SND
      6
              CA
                 TOR
                                        OOIFC
                                                             2021
                                                                             SND
      7
              CA TOR
                                        OOIFC
                                                             2021
                                                                             SND
              CA
                  TOR
                                        OOIFC
                                                             2021
      8
                                                                            SND
      9
              CA
                  TOR
                                        OOIFC
                                                             2021
                                                                             SND
      12
              CA
                  TOR
                                        OOIFC
                                                             2021
                                                                             SND
                                                             2021
      13
              CA
                  TOR
                                        OOIFC
                                                                             SND
      14
              CA
                  TOR
                                        OOIFC
                                                             2021
                                                                             SND
      15
              CA
                  TOR
                                        OOIFC
                                                             2021
                                                                             SND
      18
              CA
                  TOR
                                        OOIFC
                                                             2021
                                                                             SND
      19
              CA
                  TOR
                                        OOIFC
                                                             2021
                                                                             SND
                  TOR
                                                             2020
      20
              CA
                                        OOIFC
                                                                             SND
                 TOR
                                        OOIFC
                                                             2021
      21
              CA
                                                                            SND
      22
              CA
                 TOR
                                        OOIFC
                                                             2021
                                                                             SND
      24
              CA
                                                             2021
                  TOR
                                        OOIFC
                                                                            SND
              CA
                 TOR
                                        OOIFC
                                                             2021
      25
                                                                             SND
              CA TOR
                                                             2021
      27
                                        00IFC
                                                                             SND
      28
              CA
                  TOR
                                        OOIFC
                                                             2021
                                                                            SND
      30
              CA TOR
                                        00IFC
                                                             2021
                                                                             SND
              CA
                 TOR
                                        OOIFC
                                                             2021
      31
                                                                            SND
      32
              CA
                 TOR
                                        OOIFC
                                                             2020
                                                                             SND
      34
              CA
                                                             2021
                  TOR
                                        OOIFC
                                                                             SND
      35
              CA
                  TOR
                                        OOIFC
                                                             2021
                                                                             SND
      36
              CA
                  TOR
                                        OOIFC
                                                             2021
                                                                             SND
      37
              CA
                  TOR
                                        OOIFC
                                                             2020
                                                                             SND
```

38 40 42 43 44 45 46 48 49	CA CA CA CA CA CA CA CA CA	TOR		001FC 001FC 001FC 001FC 001FC 001FC 001FC 001FC		2021 2021 2021 2021 2021 2021 2021 2020 2021		SND SND SND SND SND SND SND SND
	Gross F	loor Area	Conveying	Exterior	Horizontal		\	
0		521.18	0.0			11.137992		
1		389.24	0.0			5.461939		
2		411.64	0.0			3.786074		
3		269.56	0.0			6.503479		
6		445.99	0.0			11.933511		
7		438.45	0.0			12.707195		
8		714.07	0.0			12.865930		
9		343.24	0.0			4.300619		
12		226.89	0.0			12.424245		
13		611.73	0.0			5.140200		
14		343.44	0.0			6.494467		
15		613.38	0.0			13.090524		
18		178.38	0.0			9.782438		
19		323.80	0.0			9.824569		
20		837.56	0.0			13.521848		
21		587.86	0.0			6.949783		
22		568.21	0.0			12.754287		
24		294.84	0.0			3.650542		
25		496.77	0.0			5.352985		
27		643.30	0.0			11.769043		
28		701.61	0.0			11.799093		
30		378.70	0.0			5.522739		
31		324.16	0.0			5.361174		
32		533.53	0.0			8.494907		
34		423.03	0.0			11.102019		
35		328.16	0.0			10.234937		
36		421.59	0.0			12.223172		
37		628.59	0.0			10.408758		
38		464.51	0.0			4.118745		
40		346.14	0.0			11.787081		
42		891.97	0.0			10.710312		
43		525.61	0.0			18.918490		
44		502.87	0.0			6.014586		
45		379.18	0.0			6.169302		
46		549.65	0.0			11.310711		
48		393.82	0.0			16.116861		

	Exterior	Vertical	Enclosures	Foundations	•••	Interio	or Finishes \	\
0			136.939623	335.649367	•••		8.309413	•
1			69.018253	281.318698	•••		6.490936	
2			101.450370	464.462195	•••		4.574905	
3			188.215196	255.359136	•••		8.510443	
6			61.325975	295.116668	•••		6.391063	
7			130.552921	269.468463	•••		6.584780	
8			104.310510	276.917123	•••		6.563894	
9			210.632241	283.893850	•••		8.940907	
12			186.668275	261.874926	•••		6.134611	
13			102.332008	343.714248	•••		7.638991	
14			147.104280	424.099610	•••		7.860800	
15			156.986570	298.537712	•••		8.068881	
18			112.523711	371.149916	•••		9.551856	
19			186.570501	148.769711	•••		9.483653	
20			91.689386	317.583491	•••		7.152371	
21			94.557055	428.185321	•••		6.754074	
22			83.789887	255.012975	•••		7.860492	
24			127.856507	261.274626	•••		4.807604	
25			89.883144	251.725837			5.921358	
27			83.949693	156.365248	•••		8.492430	
28			53.418023	266.164355	•••		7.952623	
30			164.214896	403.602589	•••		7.221059	
31			190.512918	377.853541	•••		6.597902	
32			68.518430	309.062696	•••		6.648595	
34			154.072547	243.607664	•••		4.717349	
35			184.202156	388.744353	•••		5.648226	
36			158.716507	424.443503	•••		5.625641	
37			136.076590	369.744859	•••		5.699975	
38			151.068033	412.845205	•••		7.621364	
40			146.479339	287.564257	•••		7.916204	
42			213.677214	245.205806	•••		7.577250	
43			109.529933	498.010299	•••		7.954358	
44			91.481074	278.679758	•••		4.564488	
45			172.418003	391.303861	•••		6.339432	
46			127.866168	266.468237	•••		6.701647	
48			140.069509	188.980245	•••		10.629628	
49			131.118584	347.187490	•••		5.089382	
	Plumbing	Sito Tm	provements	Slabs-On-Grad	0	Special	Construction	\
0	0.0	DICE IIII	0.0	273.97240		pheciai	0.0	`
1	0.0		0.0	192.87446			0.0	
2	0.0		0.0	170.73335			0.0	
3	0.0		0.0	124.18652			0.0	
6	0.0		0.0	153.06161			0.0	
J	0.0		0.0	100.00101			0.0	

7	0.0		0.0	211.910108		0.0
8	0.0		0.0	266.709576		0.0
9	0.0		0.0	138.510228		0.0
12	0.0		0.0	129.263543		0.0
13	0.0		0.0	165.513154		0.0
14	0.0		0.0	129.532248		0.0
15	0.0		0.0	166.414337		0.0
18	0.0		0.0	223.398638		0.0
19	0.0		0.0	158.178114		0.0
20	0.0		0.0	143.282268		0.0
21	0.0		0.0	237.918968		0.0
22	0.0		0.0	199.364347		0.0
24	0.0		0.0	131.174185		0.0
25	0.0		0.0	242.284758		0.0
27	0.0		0.0	152.407914		0.0
28	0.0		0.0	169.419640		0.0
30	0.0		0.0	179.868896		0.0
31	0.0		0.0	132.696247		0.0
32	0.0		0.0	135.390288		0.0
34	0.0		0.0	147.458950		0.0
35	0.0		0.0	128.887840		0.0
36	0.0		0.0	147.225241		0.0
37	0.0		0.0	186.334547		0.0
38	0.0		0.0	145.273403		0.0
40	0.0		0.0	139.821081		0.0
42	0.0		0.0	138.994603		0.0
43	0.0		0.0	139.646277		0.0
44	0.0		0.0	182.059329		0.0
45	0.0		0.0	158.446049		0.0
46	0.0		0.0	154.805714		0.0
48	0.0		0.0	198.860705		0.0
49	0.0		0.0	199.209464		0.0
	Subgrade End	closures	Substructi	ire Interior	\	
0	•	9.652903	Dubb ti uc ti	0.000000	1	
1		6.851955		0.000000		
2		1.298572		0.000000		
3		4.351465		0.000000		
6		9.478642		0.054452		
7		4.218921		0.000000		
8		3.902623		0.000000		
9		9.601245		0.000000		
12		3.818403		0.935612		
13	7	7.722754		0.000000		
14	Ş	9.135529		0.000000		
15	4	4.868508		0.467438		
18	(0.00000		0.000000		

19	4.617006	0.000000
20	7.131170	0.000000
21	7.959752	0.000000
22	6.339651	0.000000
24	7.469048	0.000000
25	9.448689	0.078017
27	0.000000	0.096759
28	11.919460	0.000000
30	7.509119	0.330172
31	5.073992	0.000000
32	8.867868	0.000000
34	0.00000	0.000000
35	4.762839	0.000000
36	9.538939	0.000000
37	6.039206	1.461249
38	9.071017	0.000000
40	7.568785	0.394416
42	4.540919	0.371810
43	6.720435	0.000000
44	6.092739	0.000000
45	9.489156	0.195110
46	6.042229	0.499896
48	6.057127	1.647329
49	7.221222	1.208104
	Substructure Related Activities	Superstruc
Λ	0.0	30.22

	Substructure	${\tt Related}$	Activities	Superstructure	Water	${\tt And}$	Gas	Mitigation
0			0.0	30.228003				0.0
1			0.0	26.271523				0.0
2			0.0	23.756286				0.0
3			0.0	30.517748				0.0
6			0.0	39.906513				0.0
7			0.0	39.907474				0.0
8			0.0	38.291591				0.0
9			0.0	35.370538				0.0
12			0.0	35.355314				0.0
13			0.0	33.388004				0.0
14			0.0	39.370016				0.0
15			0.0	40.958564				0.0
18			0.0	63.006044				0.0
19			0.0	36.597047				0.0
20			0.0	28.734226				0.0
21			0.0	37.457583				0.0
22			0.0	36.265538				0.0
24			0.0	30.389475				0.0
25			0.0	43.728928				0.0
27			0.0	35.393414				0.0
28			0.0	39.408113				0.0

```
30
                                  0.0
                                                                                0.0
                                             82.392236
31
                                  0.0
                                             46.380703
                                                                                0.0
32
                                  0.0
                                                                                0.0
                                             25.469871
34
                                  0.0
                                             35.666107
                                                                                0.0
35
                                  0.0
                                             49.404111
                                                                                0.0
36
                                  0.0
                                             34.035382
                                                                                0.0
37
                                  0.0
                                             47.065025
                                                                                0.0
38
                                  0.0
                                             37.921434
                                                                                0.0
                                  0.0
                                                                                0.0
40
                                             27.740220
42
                                  0.0
                                             29.045531
                                                                                0.0
                                  0.0
                                                                                0.0
43
                                             33.265489
44
                                  0.0
                                             37.265275
                                                                                0.0
45
                                  0.0
                                             46.860447
                                                                                0.0
46
                                  0.0
                                                                                0.0
                                             31.152827
48
                                  0.0
                                             49.899420
                                                                                0.0
49
                                  0.0
                                                                                0.0
                                             38.021046
```

[37 rows x 21 columns]

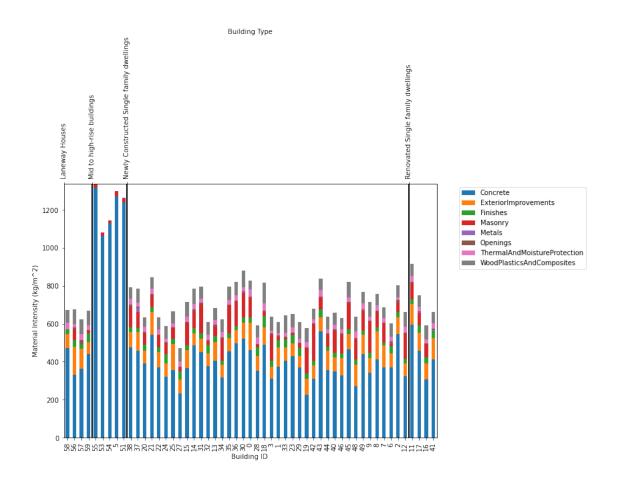
```
[38]: f = lambda x: master_format_convert[re.split('[_\.\]',x)[4]] toplot = pd.concat([df[headings[1:]],df_mi[kilogram_columns].groupby(f,axis=1).

→sum()],axis=1).sort_values(['Building Type'])
```

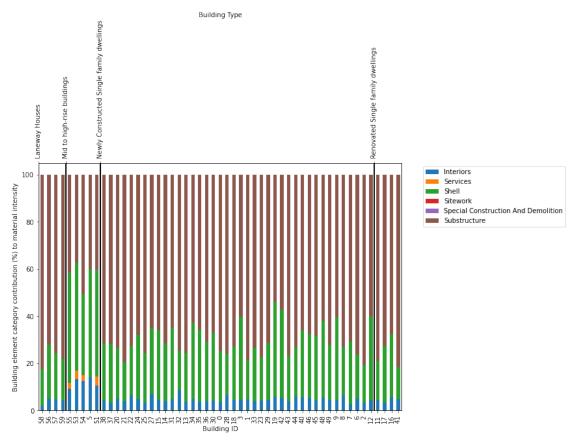
```
types_to_keep = ['APB','SND','SNR','SMR','SMD','ADU','SEC','ROW','LNW']
toplot = toplot[toplot['Building Type'].isin(types_to_keep)]
building_type_map = {
    'APB':'Mid to high-rise buildings',
    'SND':'Newly Constructed Single family dwellings',
    'SNR':'Renovated Single family dwellings',
    'SMD':'Newly Constructed Single family dwellings',
    'SMR':'Renovated Single family dwellings',
    'SMR':'Newly Constructed Single family dwellings',
    'SEC':'Newly Constructed Single family dwellings',
    'ROW':'Newly Constructed Single family dwellings',
    'LNW':'Laneway Houses'
```

```
toplot['Building Type'] = toplot['Building Type'].replace(building_type_map)
toplot = toplot.sort_values('Building Type')
```

```
[40]: fig, ax = plt.subplots(figsize=(10,7))
      cols = toplot.columns[6:]
      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)
      ax2.set_xticklabels([v for k,v in enumerate(toplot['Building Type'].values) if
      →v != toplot['Building Type'].values[k-1] or k==0])
      ax2.set_xlabel("Building Type")
      plt.grid(color='black',linewidth=2)
      plt.show()
```



```
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()
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)
ax2.set_xticklabels([v for k,v in enumerate(toplot['Building Type'].values) if
→v != toplot['Building Type'].values[k-1] or k==0])
ax2.set_xlabel("Building Type")
plt.grid(color='black',linewidth=2)
plt.show()
```



```
[43]: f = lambda x: name_map[re.split('[_\.\]',x)[1][0]] + '/' + re.split('[_\.\_
       \rightarrow]',x)[-1]
      toplot = df_mi[kilogram_columns].groupby(f,axis=1).sum()
[44]: df_mi = df[kilogram_columns].div(df['Gross Floor Area'],axis=0)
      df_mi = df_mi.div(df_mi.sum(axis=1),axis=0)
      f = lambda x: name_map[re.split('[\.\]',x)[1][0]] + '/' + re.split('[\.\_
      \rightarrow]',x)[-1]
      toplot = df_mi[kilogram_columns].groupby(f,axis=1).sum()
      for i in range(1,5):
          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_u
      \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'] += 1
      toplot_out['index'] = toplot_out['index'].astype('str')
[45]: 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')
[46]: from matplotlib.lines import Line2D
      fig, ax = plt.subplots(figsize=(7,15))
      ax.set_xlim(1,5)
      ax.set_ylim(1,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 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)_\toplot_out[v != toplot_out['Building Type'].values[k-1] or k==0])
ax2.set_ylabel("Building Type")

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

