3. Matching

3.1 Import data

3.1.1 Import sample

```
In [3]: | import pandas as pd
sample = pd.read feather("downloads/sample 2019-04-20.feather")
```

In [4]: ▶ sample.head()

Out[4]:

	gvkey	name	SIC	NAICS	GICS_group	GICS_industry	GICS_sector	GICS_subindustry	execid	yι
o	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	06175	200
1	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	06175	200
2	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	06175	200
3	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	06175	200
4	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	13283	200

In [5]: ▶ len(sample)

Out[5]: 2676

```
In [6]:
         ▶ sample['vear'].value counts()
   Out[6]: 2006.0
                       237
            2007.0
                       234
            2013.0
                       231
            2012.0
                       229
            2008.0
                       224
            2010.0
                       223
            2011.0
                       222
            2009.0
                       221
            2016.0
                       210
            2014.0
                       209
            2015.0
                       206
            2017.0
                       196
            2018.0
                        32
            Name: year, dtype: int64
```

3.1.2 Remove empty observations

Empty observations are those for which no personnel was identified by compustat in any year, and thus year has received a nan when merging compustat data on companies with compustat data on personnel. We will remove those for now.

```
In [7]:  sum(pd.isna(sample['execid']))
    Out[7]: 2
In [8]:

■ sample[pd.isna(sample['execid'])]
    Out[8]:
                                        NAICS GICS_group GICS_industry GICS_sector GICS_subindustry exe
                    gvkey
                               May
               875 007127 Department 5311 452111
                                                     2550
                                                                255030
                                                                              25
                                                                                        25503010
                                                                                                 N
                           Stores Co
                              Sears
              1155 009563 Roebuck & 5311 452111
                                                     2550
                                                                255030
                                                                              25
                                                                                        25503010
                                                                                                 N
                                Co
In [9]:
          N len(sample)
    Out[9]: 2676
In [10]:

  | sample = sample[~pd.isna(sample['execid'])]

In [11]:
           len(sample)
   Out[11]: 2674
```

3.1.2 Import recalls

Out[12]:

	country	date	description	hazard	importer	incidents	link	name	remedy
0	China	March 14, 2019	This recall involves Mobile Warming Performanc	The lithium-ion battery can overheat, melt or	Tech Gear 5.7, Inc., of San Marcos, Calif.	Tech Gear 5.7 has received four reports of bat	https://cpsc.gov /Recalls /2019/Tech- Gear-5-7-Re	Mobile Warming Performance Heated Socks	Refund
1	China	March 12, 2019	The recall expansion involves lithium-ion batt	The lithiumion batteries can overheat, posing	HP Inc., of Palo Alto, Calif.	HP has received eight new reports of battery p	https://cpsc.gov /Recalls /2019/HP- Expands- Recal	Lithium-ion batteries for HP commercial notebo	Replace
2	Taiwan and China	March 14, 2019	This recall involves O'Brien Performer Pro Com	The skis can detach from the binding during a	O'Brien Watersports Inc., of Snoqualmie, Wash.	O'Brien Watersports has received three reports	https://cpsc.gov /Recalls /2019/OBrien- Waterspor	Performer Pro Combo water skis	Refund

Make sure data is a date column.

In [13]: M recalls['date'] = pd.to datetime(recalls['date'])

In [14]: M recalls.head(3) Out[14]:

reme	name	link	incidents	importer	hazard	description	date	country	
Refu	Mobile Warming Performance Heated Socks	https://cpsc.gov /Recalls /2019/Tech- Gear-5-7-Re	Tech Gear 5.7 has received four reports of bat	Tech Gear 5.7, Inc., of San Marcos, Calif.	The lithiumion battery can overheat, melt or	This recall involves Mobile Warming Performanc	2019-03-14	China	0
Repla	Lithium-ion batteries for HP commercial notebo	https://cpsc.gov /Recalls /2019/HP- Expands- Recal	HP has received eight new reports of battery p	HP Inc., of Palo Alto, Calif.	The lithiumion batteries can overheat, posing	The recall expansion involves lithium-ion batt	2019-03-12	China	1
Refu	Performer Pro Combo water skis	https://cpsc.gov /Recalls /2019/OBrien- Waterspor	O'Brien Watersports has received three reports	O'Brien Watersports Inc., of Snoqualmie, Wash.	The skis can detach from the binding during a	This recall involves O'Brien Performer Pro Com	2019-03-14	Taiwan and China	2

3.2 Clean company names

3.2.1 Make everything lowercase.

```
In [17]: N sample['name clean'] = sample['name clean'].str.lower()
```

3.2.2 Remove special characters

```
In [18]: N sample['name clean'] = sample['name clean'].str.replace('[^\w\s]'. '')
```

3.2.3 Remove resulting double spaces

```
In [19]: N sample['name clean'] = sample['name clean'].str.replace(' '.')
```

3.2.4 Remove abbreviations like Inc, Co, etc.

We add a space to the end of the strings to be able to only remove full words. Otherwise, removing "co" would mess up occurances of "corp".

Remove trailing whitespace.

```
In [22]: | sample['name clean'] = sample['name clean'].str.strip()
```

Inspect results.

The results look promising, but there might be some missmatches for gap. JCPenney might also need some alternative names (e.g., jcpenney), so we will remove those for now. Another entry to pay attention to is "staples" which might also yield missmatches.

3.2.5 Drop ambiguous

In [26]: ■ sample.head() Out[26]: SIC NAICS GICS_group GICS_industry GICS_sector GICS_subindustry execid Best Buy 0 002184 5731 443142 2550 255040 25 25504020 06175 200 Co Inc **Best** Buy 1 002184 5731 443142 2550 255040 25 25504020 06175 200 Co Inc Best Buy 2 002184 5731 443142 2550 255040 25 25504020 06175 200 Co Inc Best Buy 3 002184 5731 443142 2550 255040 25 25504020 06175 200 Co 5731 443142 2550 25 25504020 13283 200 4 002184 255040 Co

3.3 Clean recall data

We look for matches in the retailer column.

3.3.1 Make everything lowercase

3.3.2 Remove special characters

3.3.3 Remove resulting double spaces

```
In [29]: M recalls['retailer'] = recalls['retailer'].str.replace(' '. ' ')
```

3.4 Check for complete cases

Some companies do not have executives registered in compustat in some years. We want to have all possible combinations in the dataset to match with the recalls. We create otherwise empty columns for those observations.

```
In [30]:
            companies = sample['name_clean'].unique()
              print(companies)
               ['best buy' 'officemax' 'circuit city' 'target' 'dillards'
                'dollar general' 'family dollar' 'macys' 'genuine parts' 'home depot' 'sears holdings' 'l brands' 'lowes companies' 'nordstrom' 'autonation' 'ross' 'rs legacy' 'toys r us' 'foot locker' 'tjx companies' 'big lots'
                'tiffany' 'office depot' 'signet jewelers' 'staples' 'autozone' 'kohls'
                'bed bath beyond' 'oreilly automotive' 'petsmart' 'urban outfitters'
                 'tractor supply' 'dollar tree' 'abercrombie fitch' 'carmax' 'gamestop'
                'advance auto parts' 'lkq' 'ulta beauty']
In [31]:
            years = sample['year'].unique()
               print(vears)
               [2006, 2007, 2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017,
                2018.]
In [32]:
            ▶ rows_before = len(sample)
               print(rows before)
               2525
In [33]:

    import itertools as it

               combinations = list(it.product(companies, years))
               combinations[:5]
    Out[33]: [('best buy', 2006.0),
                ('best buy', 2007.0),
('best buy', 2008.0),
('best buy', 2009.0),
                ('best buy', 2010.0)]
In [34]: ▶ missing combination = [True not in
                                           ((sample['name clean'] == combination[0]) & (sample['ye
                                           for combination in combinations
               sum(missing combination)
   Out[34]: 0
```

Fortunately, there are no missing cases. We don't have to add any dummy rows and can use the number of rows for each company-year set to see the number of managers/executives.

3.5 Find companies in recalls

3.5.1 Run testrun

Seems to be working as expected. In the next step, we want to find all matches per company-year observation.

```
recalls['vear'] = recalls['date'].dt.vear
In [39]:
In [40]:
          pd.value counts(recalls[test]['vear'])
   Out[40]: 2016.0
            2013.0
                       6
            2012.0
                       5
                      5
            2014.0
            2015.0
                      4
            2017.0
                       4
            2018.0
                      2
            2011.0
                      1
            Name: year, dtype: int64
```

3.5.2 Find all matches

8 of 10

```
In [45]:
           recalls matched
    Out[45]:
                      recalls
                                name_clean
                 year
               2016.0
                          9
                                   best buy
               2013.0
                          6
                                   best buy
               2012.0
                          5
                                   best buy
               2014.0
                          5
                                   best buy
               2015.0
                                   best buy
               2017.0
                                   best buy
               2018.0
                          2
                                   best buy
               2011.0
                                   best buy
In [46]:
           ▶ len(recalls matched)
    Out[46]: 125
          3.6 Merge
           ▶ len(sample)
    Out[47]: 2525

  | recalls matched = pd.merge(sample. recalls matched. on=['name clean'. 'vear'].

          We accurately report that we have not found recalls where the value is NA.

  | recalls matched['recalls'] = recalls matched['recalls'].fillna(0)

In [49]:
```

In [50]: M recalls matched.head()

Out[50]:

gvkey name SIC NAICS GICS_group GICS_industry GICS_sector GICS_subindustry execid years.

	gvkey	name	SIC	NAICS	GICS_group	GICS_industry	GICS_sector	GICS_subindustry	execid	yι
0	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	06175	200
1	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	06175	200
2	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	06175	200
		Best								

In [51]: ▶ len(recalls matched)

Out[51]: 2525

In [52]: ▶ recalls matched

Out[52]:

	gvkey	name	SIC	NAICS	GICS_group	GICS_industry	GICS_sector	GICS_subindustry	execid
0	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	06175
1	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	06175
2	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	06175
3	002184	Best Buy Co Inc	5731	443142	2550	255040	25	25504020	06175
		D 1							

3.6 Save to feather