

# Explore\_Data\_EH

May 11, 2024

```
[1]: import pandas as pd
import numpy as np
import altair as alt

alt.renderers.enable('png')
```

```
[1]: RendererRegistry.enable('png')
```

## 1 UWV Exploratory Analysis

Read classification information

According to the metadata, there are types of data on sick leave percentages:

- Accumulated over all economic activities
- Split into different sectors using SBI 2008
- Split on company size (1-10, 10-100 and 100+)

```
[2]: classification = pd.read_csv('data/classificatie.csv', sep=';')
classification.info()

classification['Key'] = classification['Key'].astype('category');
classification.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 39 entries, 0 to 38
Data columns (total 3 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Key              39 non-null    object
1   Title            39 non-null    object
2   Description      39 non-null    object
dtypes: object(3)
memory usage: 1.0+ KB
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 39 entries, 0 to 38
Data columns (total 3 columns):
#   Column          Non-Null Count  Dtype
---  -
```

```

0   Key          39 non-null    category
1   Title        39 non-null    object
2   Description  39 non-null    object
dtypes: category(1), object(2)
memory usage: 2.1+ KB

```

```
[3]: classification
```

```

[3]:      Key                                     Title \
0   T001081      A-U Alle economische activiteiten
1   301000      A Landbouw, bosbouw en visserij
2   305700      B Delfstoffenwinning
3   300003      B-F Nijverheid en energie
4   307500      C Industrie
5   307610      10-12 Voedings-, genotmiddelenindustrie
6   317105      17-18 Papier- en grafische industrie
7   320005      19-22 Raffinaderijen en chemie
8   328110      24-30, 33 Metaal-elektro industrie
9   346600      D Energievoorziening
10  348000      E Waterbedrijven en afvalbeheer
11  350000      F Bouwnijverheid
12  300007      G-N Commerciële dienstverlening
13  354200      G Handel
14  354300      45 Autohandel en -reparatie
15  356900      46 Groothandel en handelsbemiddeling
16  371600      47 Detailhandel (niet in auto's)
17  383100      H Vervoer en opslag
18  383200      49 Vervoer over land
19  389100      I Horeca
20  391600      J Informatie en communicatie
21  396300      K Financiële dienstverlening
22  402000      L Verhuur en handel van onroerend goed
23  403300      M Specialistische zakelijke diensten
24  410200      N Verhuur en overige zakelijke diensten
25  415300      812 Schoonmaakbedrijven
26  300013      O-U Niet-commerciële dienstverlening
27  417400      O Openbaar bestuur en overheidsdiensten
28  419000      P Onderwijs
29  422400      Q Gezondheids- en welzijnszorg
30  422500      86 Gezondheidszorg
31  422600      861 Ziekenhuizen
32  425300      87 Verpleging en zorg met overnachting
33  426600      88 Welzijnszorg zonder overnachting
34  428100      R Cultuur, sport en recreatie
35  435500      S Overige dienstverlening
36  WP19078      1 tot 10 werkzame personen
37  WP19091      10 tot 100 werkzame personen

```

	Description
0	Alle economische activiteiten \nDeze categorie...
1	Landbouw, bosbouw en visserij \nDeze sectie om...
2	Winning van delfstoffen \nDeze sectie omvat: \...
3	Nijverheid en energie \nDeze categorie is een ...
4	Industrie \nDeze sectie omvat: \n- de mechanis...
5	Vervaardiging van voedingsmiddelen, dranken en...
6	Vervaardiging van papier, karton en papier- en...
7	Raffinaderijen, chemie, vervaardiging van farm...
8	Metaal-elektro industrie\nDeze categorie is ee...
9	Productie en distributie van en handel in elek...
10	Winning en distributie van water; afval- en af...
11	Bouwnijverheid \nDeze sectie omvat: \n- algeme...
12	Commerciële dienstverlening \nDeze categorie i...
13	Groot- en detailhandel; reparatie van auto's \...
14	Handel in en reparatie van auto's, motorfietse...
15	Groothandel en handelsbemiddeling (niet in aut...
16	Detailhandel (niet in auto's) \nTot de detailh...
17	Vervoer en opslag \nDeze sectie omvat: \n- het...
18	Vervoer over land \nDeze afdeling omvat: \n- p...
19	Logies-, maaltijd- en drankverstrekking \nDeze...
20	Informatie en communicatie \nDeze sectie omvat...
21	Financiële instellingen \nDeze sectie omvat: \...
22	Verhuur van en handel in onroerend goed \nDeze...
23	Advisering, onderzoek en overige specialistisc...
24	Verhuur van roerende goederen en overige zakel...
25	Reiniging \nDeze klasse omvat: \n- glazenwasse...
26	Niet-commerciële dienstverlening \nDeze catego...
27	Openbaar bestuur, overheidsdiensten en verplic...
28	Onderwijs \nDeze sectie omvat: \n- alle vormen...
29	Gezondheids- en welzijnszorg \nDeze sectie omv...
30	Gezondheidszorg \nDeze afdeling omvat de groep...
31	Ziekenhuizen en geestelijke gezondheids- en ve...
32	Verpleging, verzorging en begeleiding met over...
33	Maatschappelijke dienstverlening zonder overna...
34	Cultuur, sport en recreatie \nDeze sectie omva...
35	Overige dienstverlening \nDeze sectie omvat (a...
36	Het aantal /"/werkzame personen/" / bestaat uit...
37	Het aantal /"/werkzame personen/" / bestaat uit...
38	Het aantal /"/werkzame personen/" / bestaat uit...

So we need to make a tree out of this with T001081 at the top

“Tree T001081 +> 301000 +> 300003 +>305700 +>307500

[ ]:

## 1.1 Read data

The CSV file contains four columns (ID, BedrijfskenmerkenSBI2008, Perioden and Ziekteverzuimpercentage\_1) There are rows with only a spaces and a period which seem to indicate missing data.

```
[4]: uwv = pd.read_csv('data/80072ned_UntypedDataSet_10052024_144733.csv', sep=';', na_values='')
```

```
uwv.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 5460 entries, 0 to 5459
```

```
Data columns (total 4 columns):
```

#	Column	Non-Null Count	Dtype
0	ID	5460 non-null	int64
1	BedrijfskenmerkenSBI2008	5460 non-null	object
2	Perioden	5460 non-null	object
3	Ziekteverzuimpercentage_1	5150 non-null	float64

```
dtypes: float64(1), int64(1), object(2)
```

```
memory usage: 170.8+ KB
```

## 1.2 Data fixing

Split the “Perioden” into three columns: Jaar (Year), RijType (is this row a quarterly row), Volgnummer (for quarterly rows, this indicates which quarter).

```
[5]: uwv[['Jaar', 'VerzuimType', 'Kwartaalnummer']] = uwv['Perioden'].str.extract(r'(\d+)(KW|JJ)(\d+)', expand=True)
```

```
uwv['Jaar'] = uwv['Jaar'].astype(int)
```

```
uwv['Kwartaalnummer'] = uwv['Kwartaalnummer'].astype(int)
```

```
uwv['VerzuimType'] = uwv['VerzuimType'].astype('category')
```

```
uwv['Kwartaal'] = uwv.apply(lambda row: row['Jaar'] * 10 + row['Kwartaalnummer'], axis=1)
```

```
uwv['BedrijfskenmerkenSBI2008'] = uwv['BedrijfskenmerkenSBI2008'].astype('category')
```

```
uwv
```

```
[5]:
```

	ID	BedrijfskenmerkenSBI2008	Perioden	Ziekteverzuimpercentage_1	\
0	0	T001081	1996KW01	5.5	
1	1	T001081	1996KW02	4.6	
2	2	T001081	1996KW03	4.0	
3	3	T001081	1996KW04	4.7	
4	4	T001081	1996JJ00	4.7	

...	...	...	...	...
5455	5455	WP19098	2023KW01	6.5
5456	5456	WP19098	2023KW02	5.7
5457	5457	WP19098	2023KW03	5.5
5458	5458	WP19098	2023KW04	6.4
5459	5459	WP19098	2023JJ00	6.0

	Jaar	VerzuimType	Kwartaalnummer	Kwartaal
0	1996	KW	1	19961
1	1996	KW	2	19962
2	1996	KW	3	19963
3	1996	KW	4	19964
4	1996	JJ	0	19960

...	...	...	...	...
5455	2023	KW	1	20231
5456	2023	KW	2	20232
5457	2023	KW	3	20233
5458	2023	KW	4	20234
5459	2023	JJ	0	20230

[5460 rows x 8 columns]

[ ]:

### 1.3 Get train and test

We will use 2022 and up as the final test data. All prior tot 2022 will be training data. To test the trained model, we wil use 2021. So we get three splits:

- All data prior to 2021 is the real train data. This is the data to perform exploratory data analysis on.
- All data from 2021 wil be the test set to test our trained models on.
- When we are really done, 2022 and onwards will be the final test set.

Additionally, we will only use the quarterly numbers (Verzuimtype = 'KW')

```
[6]: uwv_test = uwv[(uwv['Jaar'] >= 2022) & (uwv['VerzuimType'] == 'KW')]
      uwv_train = uwv[(uwv['Jaar'] < 2021) & (uwv['VerzuimType'] == 'KW')]
      uwv_train_test = uwv[(uwv['Jaar'] == 2021) & (uwv['VerzuimType'] == 'KW')]
```

```
[7]: uwv_train.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 3900 entries, 0 to 5443
Data columns (total 8 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   ID                                    3900 non-null   int64
1   BedrijfskenmerkenSBI2008             3900 non-null   category
```

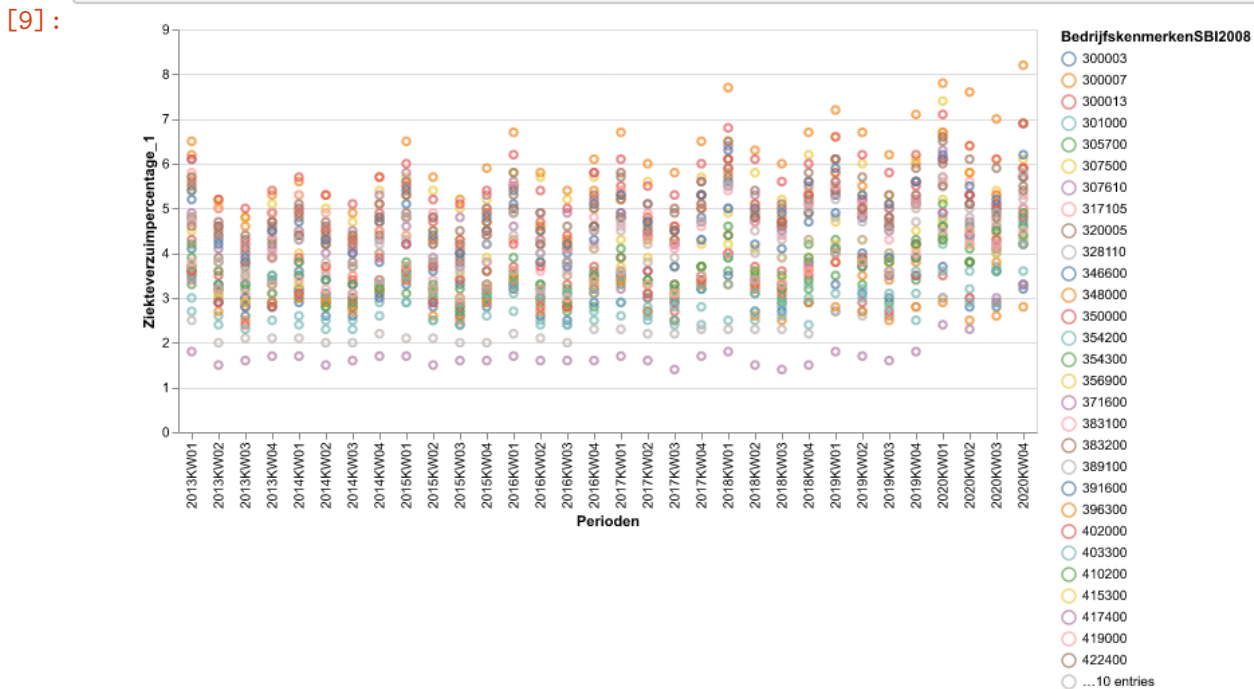
```

2   Perioden                3900 non-null   object
3   Ziekteverzuimpercentage_1 3652 non-null   float64
4   Jaar                    3900 non-null   int64
5   VerzuimType              3900 non-null   category
6   Kwartaalnummer           3900 non-null   int64
7   Kwartaal                 3900 non-null   int64
dtypes: category(2), float64(1), int64(4), object(1)
memory usage: 222.4+ KB

```

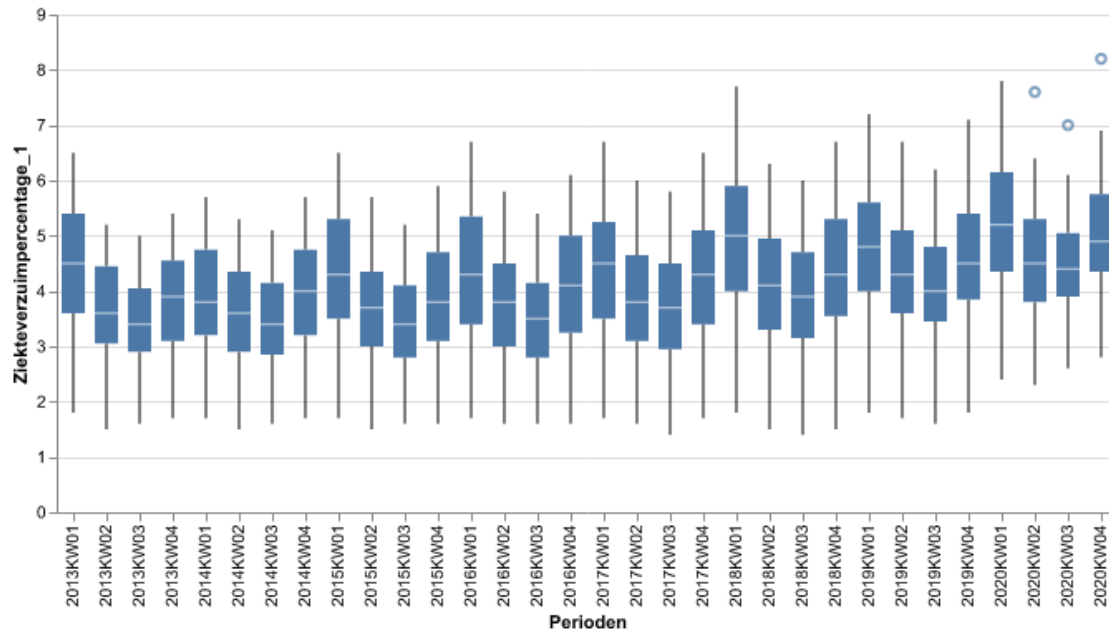
```
[8]: uwv_train_2012plus = uwv_train[uwv_train['Jaar'] > 2012]
```

```
[9]: alt.Chart(uwv_train_2012plus).mark_point().encode(
      x='Perioden',
      y='Ziekteverzuimpercentage_1',
      color='BedrijfskenmerkenSBI2008'
    )
```



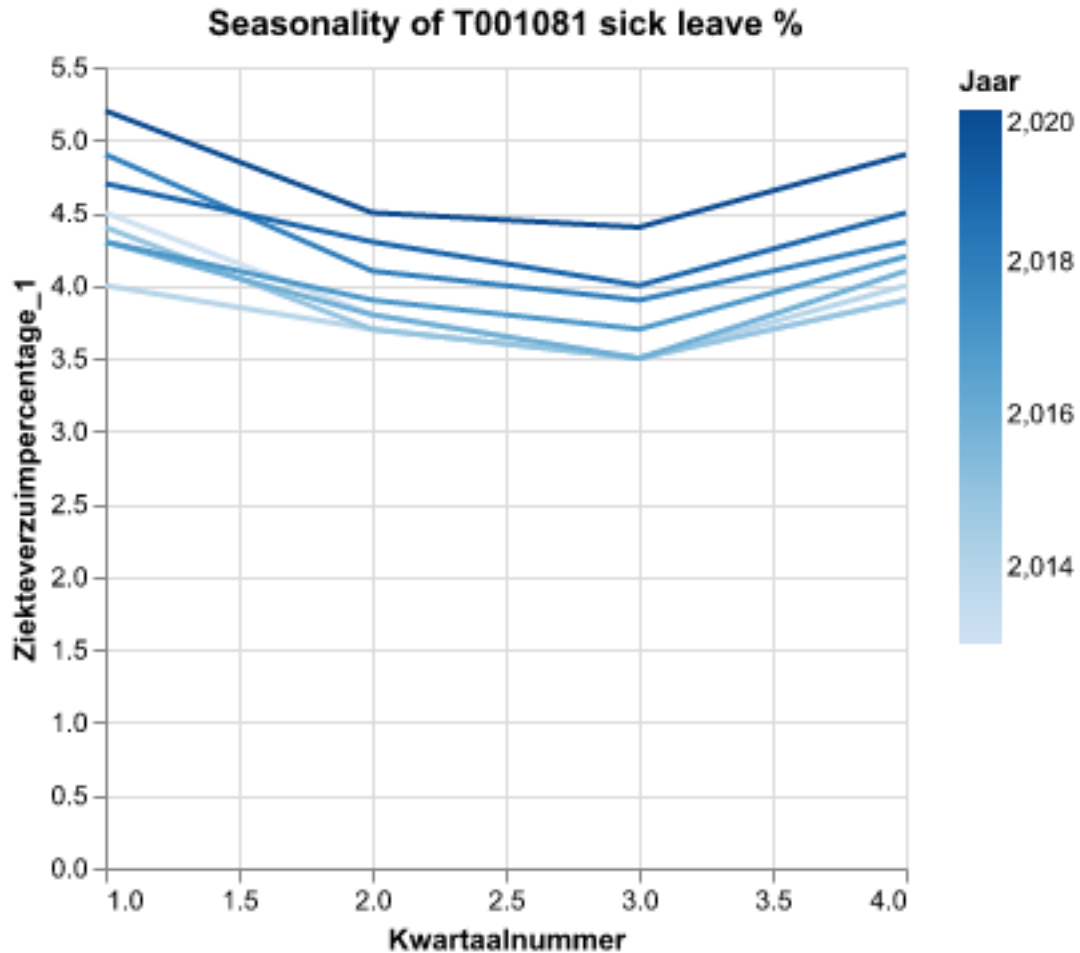
```
[10]: alt.Chart(uwv_train_2012plus).mark_boxplot().encode(
      x='Perioden',
      y='Ziekteverzuimpercentage_1',
    )
```

[10]:



```
[11]: alt.  
      ↪Chart(uwv_train_2012plus[uwv_train_2012plus['BedrijfskenmerkenSBI2008']=='T001081']).  
      ↪mark_line().encode(  
        x='Kwartaalnummer',  
        y='Ziekteverzuimpercentage_1',  
        color='Jaar'  
      ).properties(title='Seasonality of T001081 sick leave %')
```

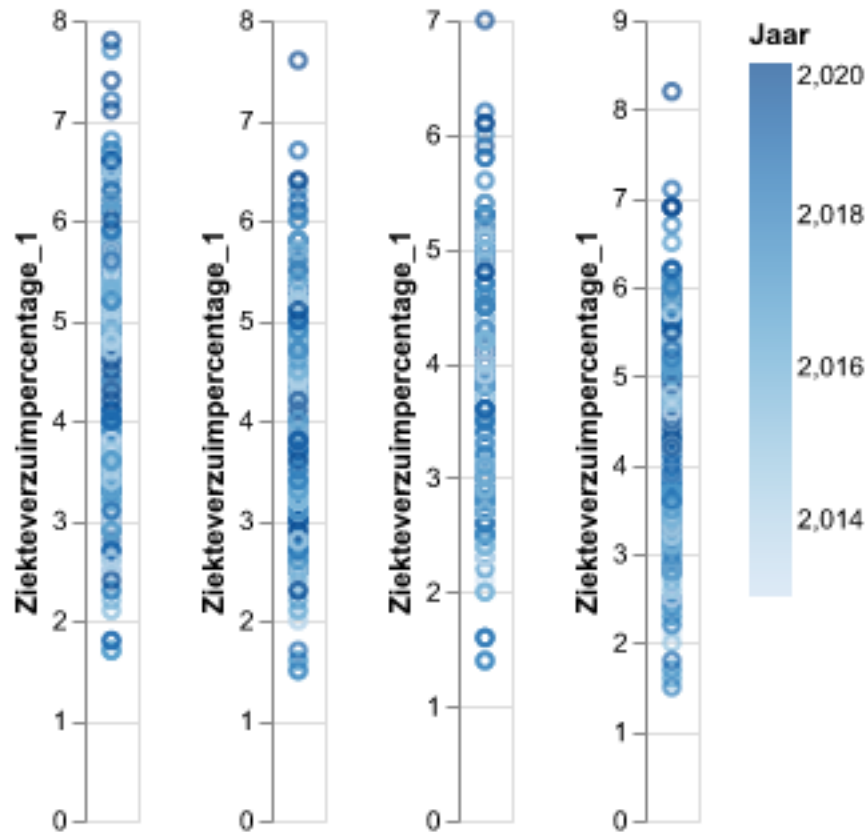
[11]:



```
[12]: alt.Chart(uwv_train_2012plus[uwv_train_2012plus['Kwartaalnummer'] == 1]).
      ↪mark_point().encode(
        y='Ziekteverzuimpercentage_1',
        color='Jaar'
      ) | alt.Chart(uwv_train_2012plus[uwv_train_2012plus['Kwartaalnummer'] == 2]).
      ↪mark_point().encode(
        y='Ziekteverzuimpercentage_1',
        color='Jaar'
      ) | alt.Chart(uwv_train_2012plus[uwv_train_2012plus['Kwartaalnummer'] == 3]).
      ↪mark_point().encode(
        y='Ziekteverzuimpercentage_1',
        color='Jaar'
      ) | alt.Chart(uwv_train_2012plus[uwv_train_2012plus['Kwartaalnummer'] == 4]).
      ↪mark_point().encode(
        y='Ziekteverzuimpercentage_1',
        color='Jaar'
      )
      )
```



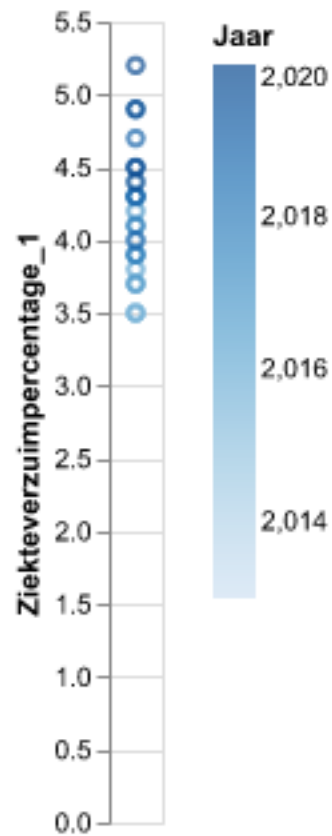
[12]:



```
[13]: alt.  
      ↪ Chart(uwv_train_2012plus[uwv_train_2012plus['BedrijfskenmerkenSBI2008']=='T001081']).  
      ↪ mark_point().encode(  
          x=alt.X(alt.repeat("column"), type='ordinal'),  
          y='Ziekteverzuimpercentage_1',  
          color='Jaar'  
      ).repeat(  
          row=['Kwartaalnummer']  
      ).properties(  
          title='Seasonality of T001081 sick leave %'  
      )
```

[13]:

Seasonality of T001081 sick leave %



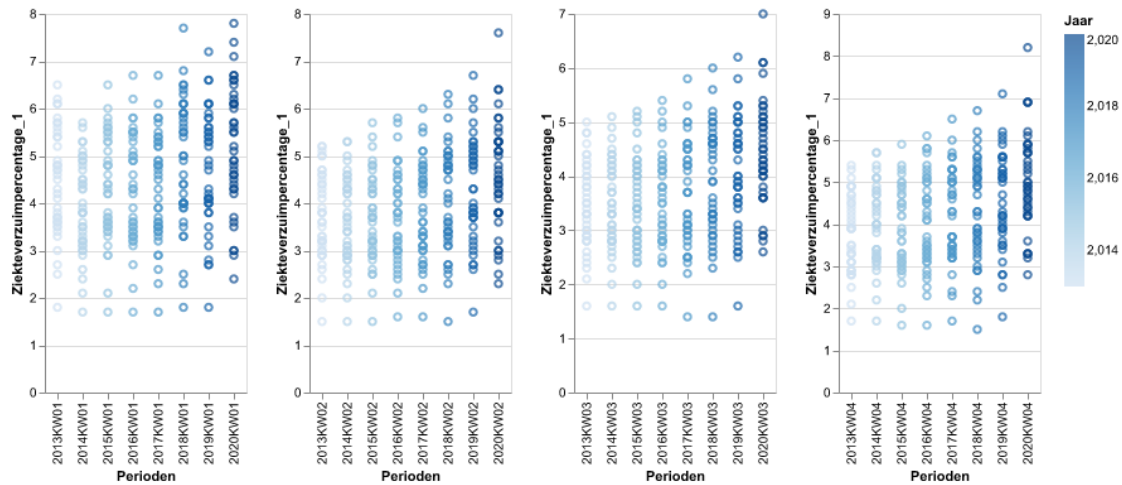
```
[14]: alt.Chart(uwv_train_2012plus[uwv_train_2012plus['Kwartaalnummer'] == 1]).
      ↪mark_point().encode(
        x='Perioden',
        y='Ziekteverzuimpercentage_1',
        color='Jaar'
    ) | alt.Chart(uwv_train_2012plus[uwv_train_2012plus['Kwartaalnummer'] == 2]).
      ↪mark_point().encode(
        x='Perioden',
        y='Ziekteverzuimpercentage_1',
        color='Jaar'
    ) | alt.Chart(uwv_train_2012plus[uwv_train_2012plus['Kwartaalnummer'] == 3]).
      ↪mark_point().encode(
        x='Perioden',
        y='Ziekteverzuimpercentage_1',
        color='Jaar'
    ) | alt.Chart(uwv_train_2012plus[uwv_train_2012plus['Kwartaalnummer'] == 4]).
      ↪mark_point().encode(
        x='Perioden',
```

```

y='Ziekteverzuimpercentage_1',
color='Jaar'
)

```

[14]:



```

[15]: uwv_train_2012plus_T =
↳ uwv_train_2012plus[uwv_train_2012plus['BedrijfskenmerkenSBI2008'] ==
↳ 'T001081']
uwv_train_2012plus_T

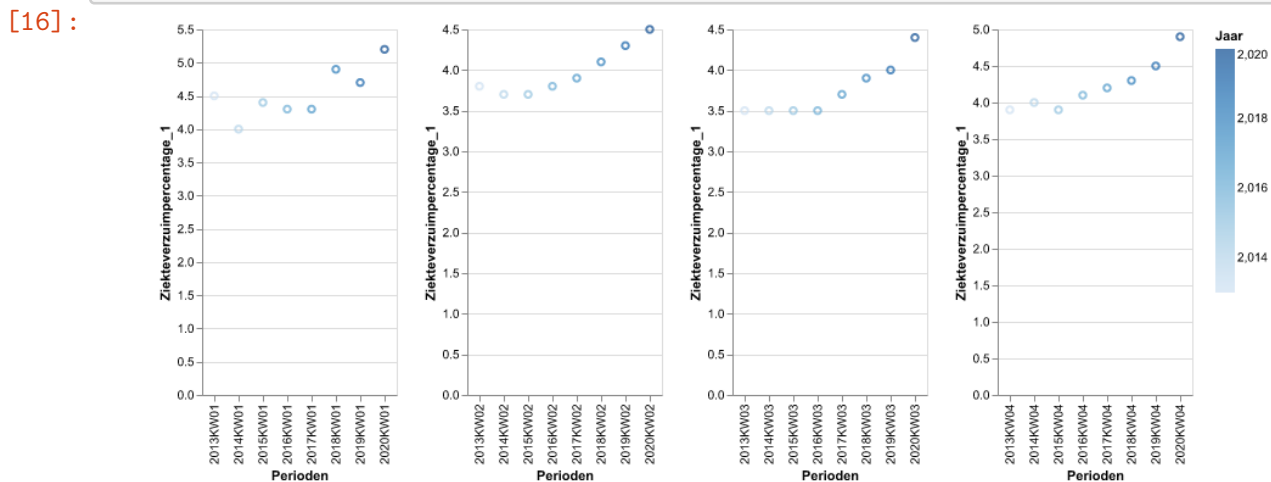
```

[15]:	ID	BedrijfskenmerkenSBI2008	Perioden	Ziekteverzuimpercentage_1	Jaar	\
	85	T001081	2013KW01	4.5	2013	
	86	T001081	2013KW02	3.8	2013	
	87	T001081	2013KW03	3.5	2013	
	88	T001081	2013KW04	3.9	2013	
	90	T001081	2014KW01	4.0	2014	
	91	T001081	2014KW02	3.7	2014	
	92	T001081	2014KW03	3.5	2014	
	93	T001081	2014KW04	4.0	2014	
	95	T001081	2015KW01	4.4	2015	
	96	T001081	2015KW02	3.7	2015	
	97	T001081	2015KW03	3.5	2015	
	98	T001081	2015KW04	3.9	2015	
	100	T001081	2016KW01	4.3	2016	
	101	T001081	2016KW02	3.8	2016	
	102	T001081	2016KW03	3.5	2016	
	103	T001081	2016KW04	4.1	2016	
	105	T001081	2017KW01	4.3	2017	
	106	T001081	2017KW02	3.9	2017	
	107	T001081	2017KW03	3.7	2017	
	108	T001081	2017KW04	4.2	2017	

110	110	T001081	2018KW01	4.9	2018
111	111	T001081	2018KW02	4.1	2018
112	112	T001081	2018KW03	3.9	2018
113	113	T001081	2018KW04	4.3	2018
115	115	T001081	2019KW01	4.7	2019
116	116	T001081	2019KW02	4.3	2019
117	117	T001081	2019KW03	4.0	2019
118	118	T001081	2019KW04	4.5	2019
120	120	T001081	2020KW01	5.2	2020
121	121	T001081	2020KW02	4.5	2020
122	122	T001081	2020KW03	4.4	2020
123	123	T001081	2020KW04	4.9	2020

	VerzuimType	Kwartaalnummer	Kwartaal
85	KW	1	20131
86	KW	2	20132
87	KW	3	20133
88	KW	4	20134
90	KW	1	20141
91	KW	2	20142
92	KW	3	20143
93	KW	4	20144
95	KW	1	20151
96	KW	2	20152
97	KW	3	20153
98	KW	4	20154
100	KW	1	20161
101	KW	2	20162
102	KW	3	20163
103	KW	4	20164
105	KW	1	20171
106	KW	2	20172
107	KW	3	20173
108	KW	4	20174
110	KW	1	20181
111	KW	2	20182
112	KW	3	20183
113	KW	4	20184
115	KW	1	20191
116	KW	2	20192
117	KW	3	20193
118	KW	4	20194
120	KW	1	20201
121	KW	2	20202
122	KW	3	20203
123	KW	4	20204

```
[16]: alt.Chart(uwv_train_2012plus_T[(uwv_train_2012plus_T['Kwartaalnummer'] == 1) ]).
      ↪mark_point().encode(
          x='Perioden',
          y='Ziekteverzuimpercentage_1',
          color='Jaar'
      ) | alt.Chart(uwv_train_2012plus_T[uwv_train_2012plus_T['Kwartaalnummer'] == 1
      ↪2]).mark_point().encode(
          x='Perioden',
          y='Ziekteverzuimpercentage_1',
          color='Jaar'
      ) | alt.Chart(uwv_train_2012plus_T[uwv_train_2012plus_T['Kwartaalnummer'] == 1
      ↪3]).mark_point().encode(
          x='Perioden',
          y='Ziekteverzuimpercentage_1',
          color='Jaar'
      ) | alt.Chart(uwv_train_2012plus_T[uwv_train_2012plus_T['Kwartaalnummer'] == 1
      ↪4]).mark_point().encode(
          x='Perioden',
          y='Ziekteverzuimpercentage_1',
          color='Jaar'
      )
      ↪)
```



```
[17]: from pandas.plotting import lag_plot
      import matplotlib.pyplot as plt

      fig, axes = plt.subplots(nrows=2, ncols=4, figsize=(20,10))

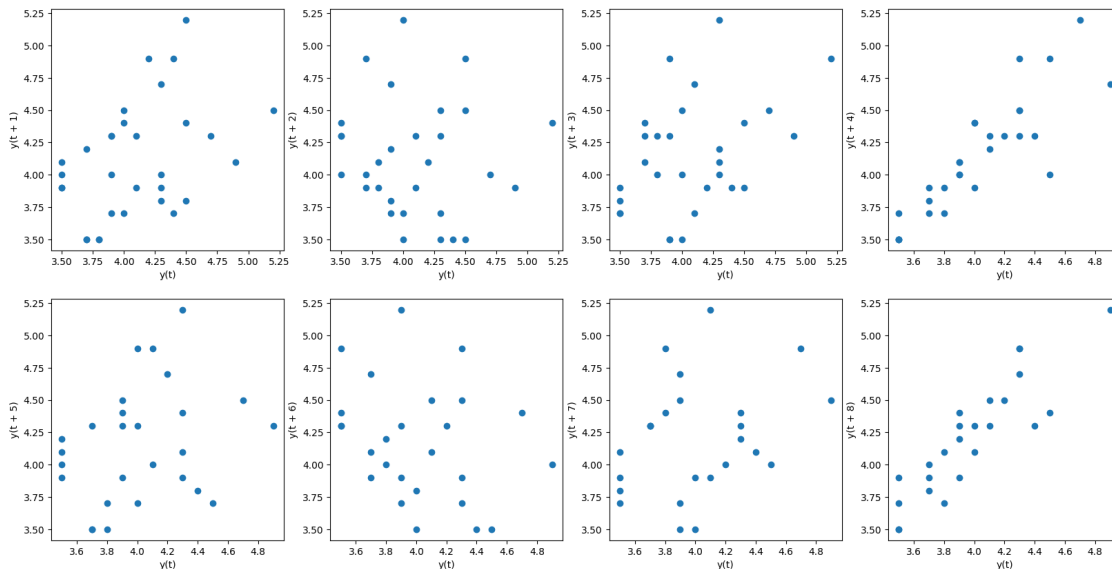
      lag_plot(uwv_train_2012plus_T['Ziekteverzuimpercentage_1'], lag=1, ax=axes[0,
      ↪0])
```

```

lag_plot(uwv_train_2012plus_T['Ziekteverzuimpercentage_1'], lag=2, ax=axes[0,1])
lag_plot(uwv_train_2012plus_T['Ziekteverzuimpercentage_1'], lag=3, ax=axes[0,2])
lag_plot(uwv_train_2012plus_T['Ziekteverzuimpercentage_1'], lag=4, ax=axes[0,3])
lag_plot(uwv_train_2012plus_T['Ziekteverzuimpercentage_1'], lag=5, ax=axes[1,0])
lag_plot(uwv_train_2012plus_T['Ziekteverzuimpercentage_1'], lag=6, ax=axes[1,1])
lag_plot(uwv_train_2012plus_T['Ziekteverzuimpercentage_1'], lag=7, ax=axes[1,2])
lag_plot(uwv_train_2012plus_T['Ziekteverzuimpercentage_1'], lag=8, ax=axes[1,3])

plt.show()

```



```

[18]: import math

start_lag = 0
lag_length = 21

lagged_autocorrelation = pd.DataFrame()
lagged_autocorrelation['lag'] = range(start_lag, lag_length)

white_noise_border = 1.96 / math.
    sqrt(len(uwv_train_2012plus_T['Ziekteverzuimpercentage_1']))

```

```

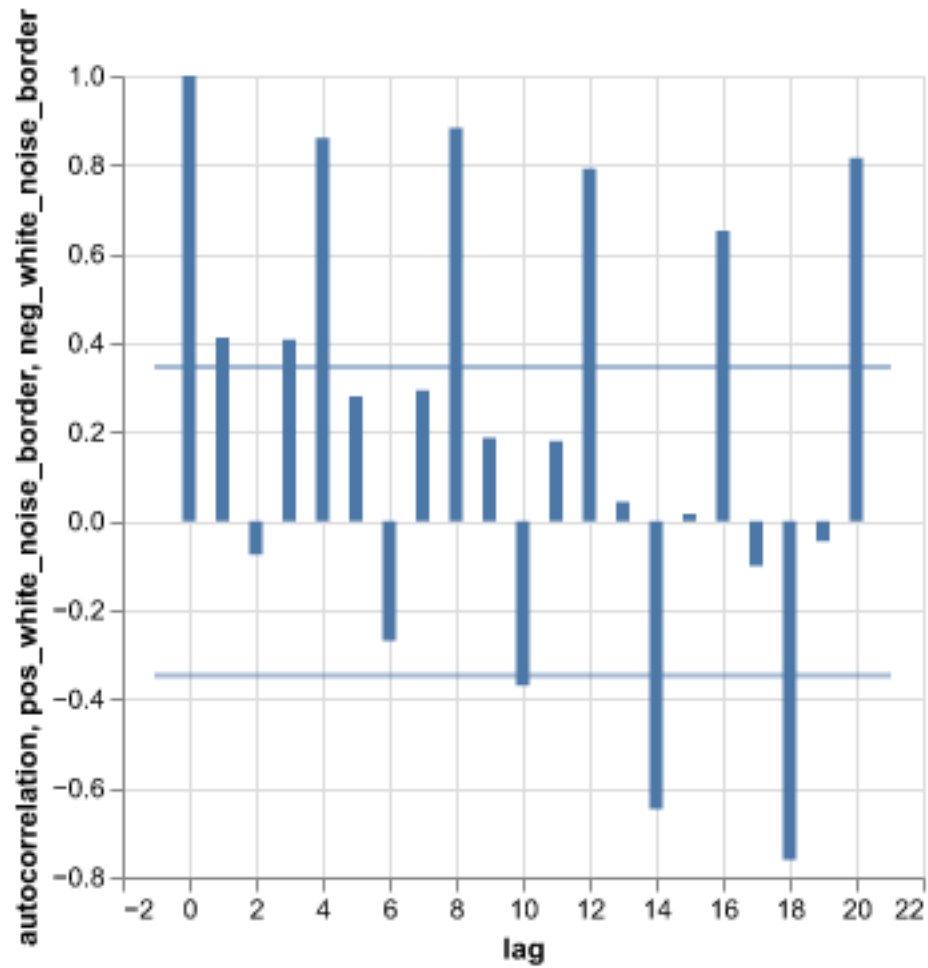
wn_border = pd.DataFrame()
wn_border['lag'] = range(start_lag - 1, lag_length + 1)
wn_border['pos_white_noise_border'] = [white_noise_border for _ in
    ↪range(start_lag - 1, lag_length + 1)]
wn_border['neg_white_noise_border'] = [-white_noise_border for _ in
    ↪range(start_lag - 1, lag_length + 1)]

lagged_autocorrelation['autocorrelation'] =
    ↪[uwv_train_2012plus_T['Ziekteverzuimpercentage_1'].autocorr(lag=lag) for lag
    ↪in lagged_autocorrelation['lag']]

alt.Chart(lagged_autocorrelation).mark_bar().encode(
    x='lag',
    y='autocorrelation',
) + alt.Chart(wn_border).mark_line(strokeDash=[1,1]).encode(
    x='lag',
    y='pos_white_noise_border',
) + alt.Chart(wn_border).mark_line(strokeDash=[1,1]).encode(
    x='lag',
    y='neg_white_noise_border'
)

```

[18]:



```
[19]: moving_average = pd.DataFrame()

moving_average['quarter'] = uwv_train_2012plus_T['Perioden']
moving_average['sick'] = uwv_train_2012plus_T['Ziekteverzuimpercentage_1']

for window in range(3, 16, 2):
    moving_average[f'{window}-MA'] =
    ↳uwv_train_2012plus_T['Ziekteverzuimpercentage_1'].rolling(window,
    ↳center=True).mean()

moving_average
```

```
[19]:
```

	quarter	sick	3-MA	5-MA	7-MA	9-MA	11-MA	13-MA	\
85	2013KW01	4.5	NaN	NaN	NaN	NaN	NaN	NaN	
86	2013KW02	3.8	3.933333	NaN	NaN	NaN	NaN	NaN	
87	2013KW03	3.5	3.733333	3.94	NaN	NaN	NaN	NaN	
88	2013KW04	3.9	3.800000	3.78	3.842857	NaN	NaN	NaN	



90	2014KW01	4.0	3.866667	3.72	3.771429	3.922222	NaN	NaN
91	2014KW02	3.7	3.733333	3.82	3.857143	3.833333	3.863636	NaN
92	2014KW03	3.5	3.733333	3.92	3.885714	3.800000	3.809091	3.900000
93	2014KW04	4.0	3.966667	3.86	3.828571	3.844444	3.854545	3.846154
95	2015KW01	4.4	4.033333	3.82	3.814286	3.888889	3.881818	3.823077
96	2015KW02	3.7	3.866667	3.90	3.900000	3.866667	3.845455	3.869231
97	2015KW03	3.5	3.700000	3.96	3.942857	3.844444	3.854545	3.900000
98	2015KW04	3.9	3.900000	3.84	3.871429	3.911111	3.909091	3.892308
100	2016KW01	4.3	4.000000	3.80	3.828571	3.944444	3.945455	3.892308
101	2016KW02	3.8	3.866667	3.92	3.914286	3.888889	3.918182	3.946154
102	2016KW03	3.5	3.800000	4.00	3.971429	3.888889	3.900000	4.015385
103	2016KW04	4.1	3.966667	3.92	3.942857	3.966667	4.009091	3.992308
105	2017KW01	4.3	4.100000	3.90	3.928571	4.077778	4.063636	4.007692
106	2017KW02	3.9	3.966667	4.04	4.085714	4.055556	4.063636	4.069231
107	2017KW03	3.7	3.933333	4.20	4.171429	4.066667	4.063636	4.130769
108	2017KW04	4.2	4.266667	4.16	4.142857	4.155556	4.145455	4.130769
110	2018KW01	4.9	4.400000	4.16	4.142857	4.222222	4.218182	4.146154
111	2018KW02	4.1	4.300000	4.28	4.257143	4.222222	4.209091	4.223077
112	2018KW03	3.9	4.100000	4.38	4.342857	4.233333	4.227273	4.307692
113	2018KW04	4.3	4.300000	4.26	4.314286	4.322222	4.345455	4.323077
115	2019KW01	4.7	4.433333	4.24	4.257143	4.433333	4.418182	4.361538
116	2019KW02	4.3	4.333333	4.36	4.414286	4.388889	4.436364	4.453846
117	2019KW03	4.0	4.266667	4.54	4.500000	4.422222	4.436364	NaN
118	2019KW04	4.5	4.566667	4.50	4.514286	4.533333	NaN	NaN
120	2020KW01	5.2	4.733333	4.52	4.542857	NaN	NaN	NaN
121	2020KW02	4.5	4.700000	4.70	NaN	NaN	NaN	NaN
122	2020KW03	4.4	4.600000	NaN	NaN	NaN	NaN	NaN
123	2020KW04	4.9	NaN	NaN	NaN	NaN	NaN	NaN

15-MA

85	NaN
86	NaN
87	NaN
88	NaN
90	NaN
91	NaN
92	NaN
93	3.866667
95	3.840000
96	3.873333
97	3.900000
98	3.886667
100	3.900000
101	3.980000
102	4.020000
103	4.013333
105	4.006667

```
106 4.073333
107 4.126667
108 4.133333
110 4.146667
111 4.240000
112 4.306667
113 4.326667
115 4.366667
116      NaN
117      NaN
118      NaN
120      NaN
121      NaN
122      NaN
123      NaN
```

```
[20]: charts = [alt.Chart(moving_average).mark_line().encode(x='quarter', y='sick')]

for window in range(3, 16, 2):
    charts.append(alt.Chart(moving_average).mark_line().encode(x='quarter',
        y=f'{window}-MA'))

alt.vconcat(*charts)
```

```
[20]:
```



```
[21]: from statsmodels.tsa.seasonal import STL

slp = uwv_train_2012plus_T['Ziekteverzuimpercentage_1']
slp.index = uwv_train_2012plus_T['Perioden']
slp
```

```
[21]: Perioden
2013KW01    4.5
2013KW02    3.8
2013KW03    3.5
2013KW04    3.9
2014KW01    4.0
2014KW02    3.7
2014KW03    3.5
2014KW04    4.0
2015KW01    4.4
2015KW02    3.7
2015KW03    3.5
2015KW04    3.9
2016KW01    4.3
2016KW02    3.8
2016KW03    3.5
2016KW04    4.1
2017KW01    4.3
2017KW02    3.9
2017KW03    3.7
2017KW04    4.2
2018KW01    4.9
2018KW02    4.1
2018KW03    3.9
2018KW04    4.3
2019KW01    4.7
2019KW02    4.3
2019KW03    4.0
2019KW04    4.5
2020KW01    5.2
2020KW02    4.5
2020KW03    4.4
2020KW04    4.9
Name: Ziekteverzuimpercentage_1, dtype: float64
```

```
[22]: plt.rc("font", size=6)
stl = STL(slp, period=4)
res = stl.fit()
```

```
fig = res.plot()
fig.autofmt_xdate()
```

