

Project_2_1

July 28, 2020

1 Project 2 Exercise 1. Audiology databases preprocessing

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from os import walk
import re
```

1.1 1. Read files

```
[2]: data_path = './data'
```

```
[3]: for root, dirs, files in walk(data_path):
    all_files = [data_path+'/'+file for file in files if file.split('.')[1] in_
↳ ['data', 'names']]

for file_num in range(len(all_files)):
    print(f'#{file_num} ->', all_files[file_num])
```

```
#0 -> ./data/audiology.names
#1 -> ./data/audiology.data
#2 -> ./data/audiology.standardized.data
#3 -> ./data/audiology.standardized.names
```

```
[4]: # in here we choose the index from above result. for example here we want data_
↳ file with index 2
# and data names file with index 3

# open names file
names_file = [line.strip() for line in open(all_files[3], 'r').readlines()]

print_flag = False
attrs = []
for line in names_file:
    if 'Missing attributes' in line:
        break
```

```

if print_flag:
    attr = re.findall("(\w*)\(*\)?:", line)

    if len(attr) > 0:
        attrs.append(attr[0])

if 'Attribute' in line:
    print_flag = True

attrs.insert(-1, 'indentifier')

# print all attributes
print('attributes count: ', len(attrs), '\n'*2)
print(attrs)

```

attributes count: 71

```

['age_gt_60', 'air', 'airBoneGap', 'ar_c', 'ar_u', 'bone', 'boneAbnormal',
'bser', 'history_buzzing', 'history_dizziness', 'history_fluctuating',
'history_fullness', 'history_heredit', 'history_nausea', 'history_noise',
'history_recruitment', 'history_ringing', 'history_roaring', 'history_vomiting',
'late_wave_poor', 'm_at_2k', 'm_cond_lt_1k', 'm_gt_1k', 'm_m_gt_2k', 'm_m_sn',
'm_m_sn_gt_1k', 'm_m_sn_gt_2k', 'm_m_sn_gt_500', 'm_p_sn_gt_2k', 'm_s_gt_500',
'm_s_sn', 'm_s_sn_gt_1k', 'm_s_sn_gt_2k', 'm_s_sn_gt_3k', 'm_s_sn_gt_4k',
'm_sn_2_3k', 'm_sn_gt_1k', 'm_sn_gt_2k', 'm_sn_gt_3k', 'm_sn_gt_4k',
'm_sn_gt_500', 'm_sn_gt_6k', 'm_sn_lt_1k', 'm_sn_lt_2k', 'm_sn_lt_3k',
'middle_wave_poor', 'mod_gt_4k', 'mod_mixed', 'mod_s_mixed', 'mod_s_sn_gt_500',
'mod_sn', 'mod_sn_gt_1k', 'mod_sn_gt_2k', 'mod_sn_gt_3k', 'mod_sn_gt_4k',
'mod_sn_gt_500', 'notch_4k', 'notch_at_4k', 'o_ar_c', 'o_ar_u', 's_sn_gt_1k',
's_sn_gt_2k', 's_sn_gt_4k', 'speech', 'static_normal', 'tymp',
'viith_nerve_signs', 'wave_V_delayed', 'waveform_ItoV_prolonged', 'indentifier',
'class']

```

```

[5]: data_df = pd.read_csv(all_files[2])
data_df.columns = attrs
data_df = data_df.replace('?', np.NaN)
data_df

```

```

[5]:   age_gt_60      air airBoneGap  ar_c  ar_u      bone boneAbnormal \
0         f  moderate          f  normal  normal      NaN          t
1         t    mild          t   NaN  absent    mild          t
2         t    mild          t   NaN  absent    mild          f
3         t    mild          f  normal  normal    mild          t
4         t    mild          f  normal  normal    mild          t
..      ...      ...      ...      ...      ...      ...

```

194	t	mild	f	absent	normal	mild	t
195	t	mild	f	normal	absent	mild	f
196	f	normal	f	normal	normal	unmeasured	f
197	t	mild	f	normal	normal	unmeasured	f
198	t	normal	f	normal	normal	unmeasured	f

	bser	history_buzzing	history_dizziness	...	s_sn_gt_2k	s_sn_gt_4k	\
0	NaN	f	f	...	f	f	
1	NaN	f	f	...	f	f	
2	NaN	f	f	...	f	f	
3	NaN	f	f	...	f	f	
4	NaN	f	f	...	f	f	
..	
194	NaN	f	f	...	f	f	
195	NaN	f	f	...	f	f	
196	degraded	f	f	...	f	f	
197	NaN	f	f	...	f	f	
198	NaN	f	f	...	f	f	

	speech	static_normal	tymp	viith_nerve_signs	wave_V_delayed	\
0	normal	t	a	f	f	
1	normal	t	as	f	f	
2	normal	t	b	f	f	
3	good	t	a	f	f	
4	very_good	t	a	f	f	
..	
194	very_good	t	a	f	f	
195	very_good	t	c	f	f	
196	normal	f	a	f	f	
197	very_good	t	a	f	f	
198	normal	t	a	f	f	

	waveform_ItoV_prolonged	indentifier	class
0	f	p2	cochlear_unknown
1	f	p3	mixed_cochlear_age_fixation
2	f	p4	mixed_cochlear_age_otitis_media
3	f	p5	cochlear_age
4	f	p6	cochlear_age
..
194	f	p196	cochlear_age
195	f	p197	mixed_cochlear_age_otitis_media
196	f	p198	possible_brainstem_disorder
197	f	p199	cochlear_age
198	f	p200	cochlear_age

[199 rows x 71 columns]

1.2 2. Columns and Rows counts

```
[6]: print('Columns counts: ', len(data_df.columns))
      print('Row counts: ', len(data_df))
```

Columns counts: 71

Row counts: 199

1.3 3 & 4 . Miss values count

```
[7]: pd.DataFrame(data_df.isna().any())
```

```
[7]:
```

	0
age_gt_60	False
air	False
airBoneGap	False
ar_c	True
ar_u	True
...	...
viith_nerve_signs	False
wave_V_delayed	False
waveform_ItoV_prolonged	False
indentifier	False
class	False

[71 rows x 1 columns]

```
[8]: pd.DataFrame(data_df.isna().sum())
```

```
[8]:
```

	0
age_gt_60	0
air	0
airBoneGap	0
ar_c	4
ar_u	3
...	..
viith_nerve_signs	0
wave_V_delayed	0
waveform_ItoV_prolonged	0
indentifier	0
class	0

[71 rows x 1 columns]

1.4 5. Fill missing values by imputing them

```
[9]: cols_with_missing = [col for col in data_df.columns if data_df[col].isnull().
    ↪any()]
print('Columns with missing values: ', cols_with_missing)

# for missing values we can use simple imputer to fill missing values...
# from sklearn.impute import SimpleImputer
# my_imputer = SimpleImputer()
# imputed_data_df = pd.DataFrame(my_imputer.fit_transform(data_df))
# imputed_data_df.columns = data_df.columns
# data_df = imputed_data_df

# Actually!!! ther is an easier way to handle missing values...
# we can just simply DROP them! :)))

data_df = data_df.drop(cols_with_missing, axis=1)
data_df
```

Columns with missing values: ['ar_c', 'ar_u', 'bone', 'bser', 'o_ar_c', 'o_ar_u', 'speech']

```
[9]:
```

	age_gt_60	air	airBoneGap	boneAbnormal	history_buzzing	\
0	f	moderate	f	t	f	
1	t	mild	t	t	f	
2	t	mild	t	f	f	
3	t	mild	f	t	f	
4	t	mild	f	t	f	
..	
194	t	mild	f	t	f	
195	t	mild	f	f	f	
196	f	normal	f	f	f	
197	t	mild	f	f	f	
198	t	normal	f	f	f	

	history_dizziness	history_fluctuating	history_fullness	history_heredit	\
0	f	f	f	f	
1	f	f	f	f	
2	f	f	f	f	
3	f	f	f	f	
4	f	f	f	f	
..	
194	f	f	f	f	
195	f	f	f	f	
196	f	f	f	f	
197	f	f	f	f	
198	f	f	f	f	

	history_nausea	...	s_sn_gt_1k	s_sn_gt_2k	s_sn_gt_4k	static_normal	tymp	\
0	f	...	f	f	f		t	a
1	f	...	f	f	f		t	as
2	f	...	f	f	f		t	b
3	f	...	f	f	f		t	a
4	f	...	f	f	f		t	a
..	
194	f	...	f	f	f		t	a
195	f	...	f	f	f		t	c
196	t	...	f	f	f	f	f	a
197	f	...	f	f	f		t	a
198	f	...	f	f	f		t	a

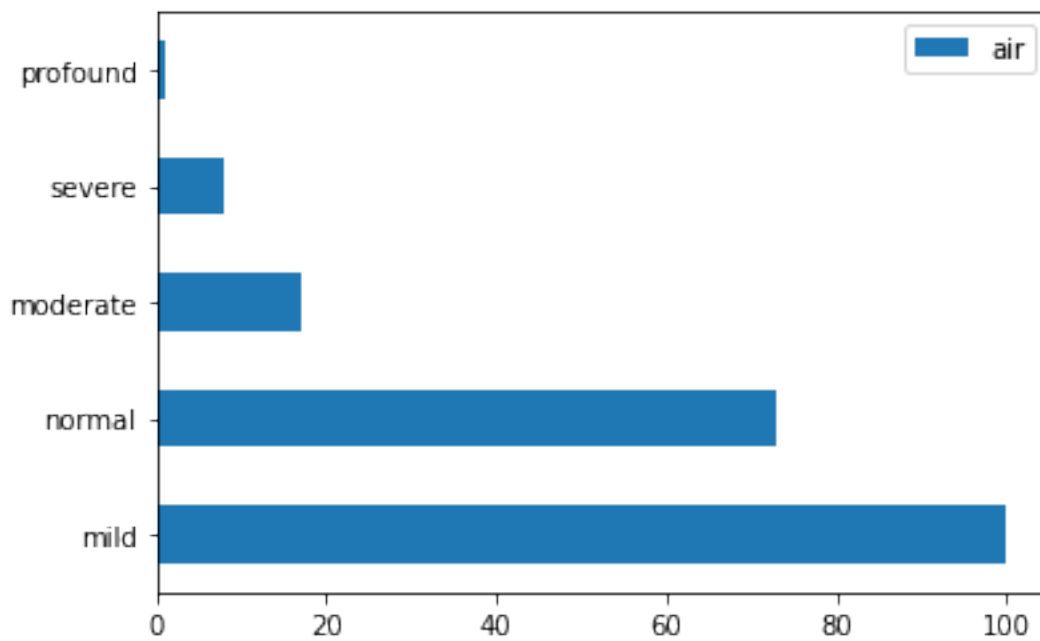
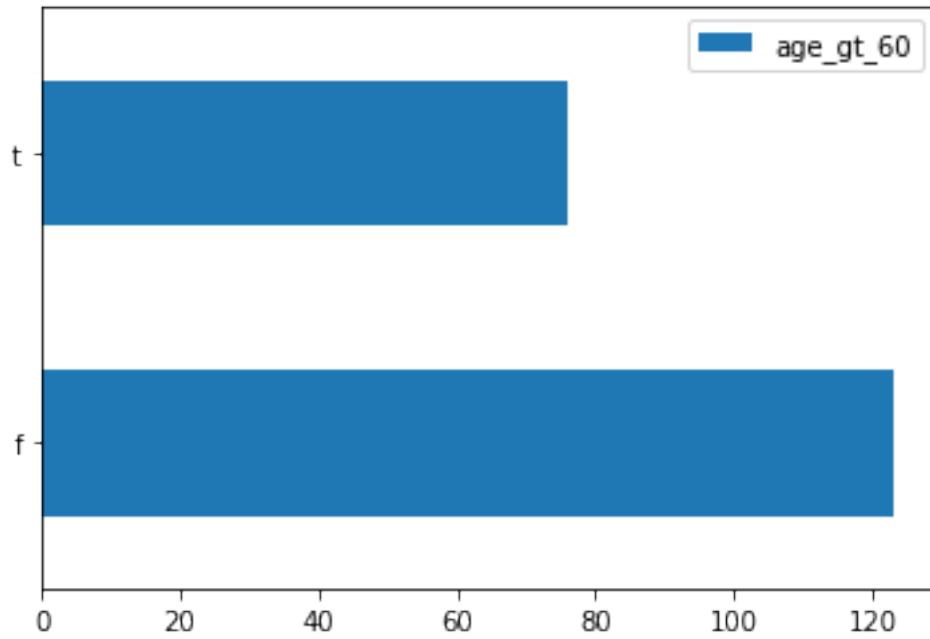
	viith_nerve_signs	wave_V_delayed	waveform_ItoV_prolonged	indentifier	\
0	f	f	f	p2	
1	f	f	f	p3	
2	f	f	f	p4	
3	f	f	f	p5	
4	f	f	f	p6	
..	
194	f	f	f	p196	
195	f	f	f	p197	
196	f	f	f	p198	
197	f	f	f	p199	
198	f	f	f	p200	

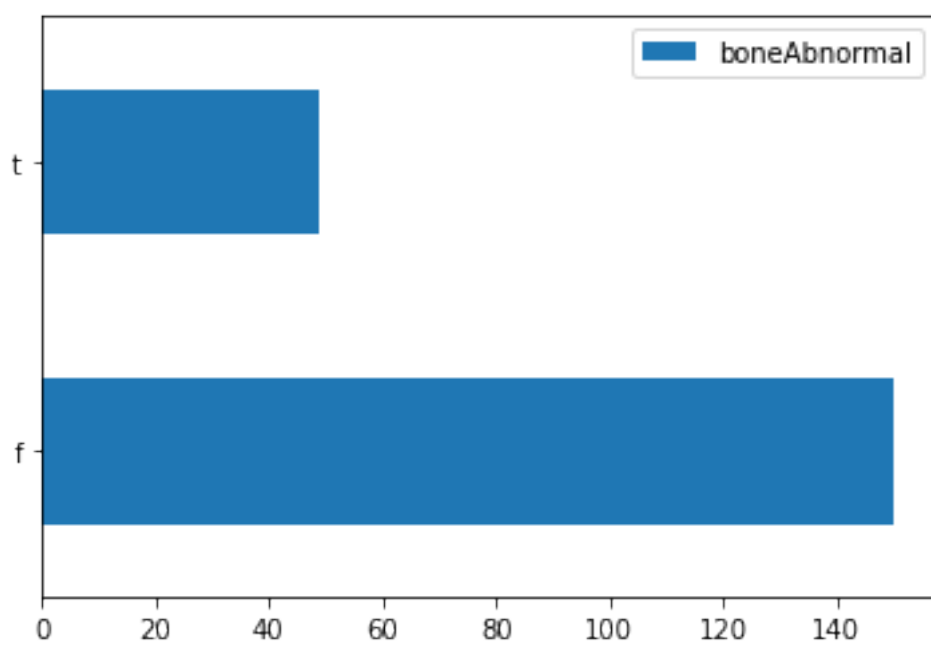
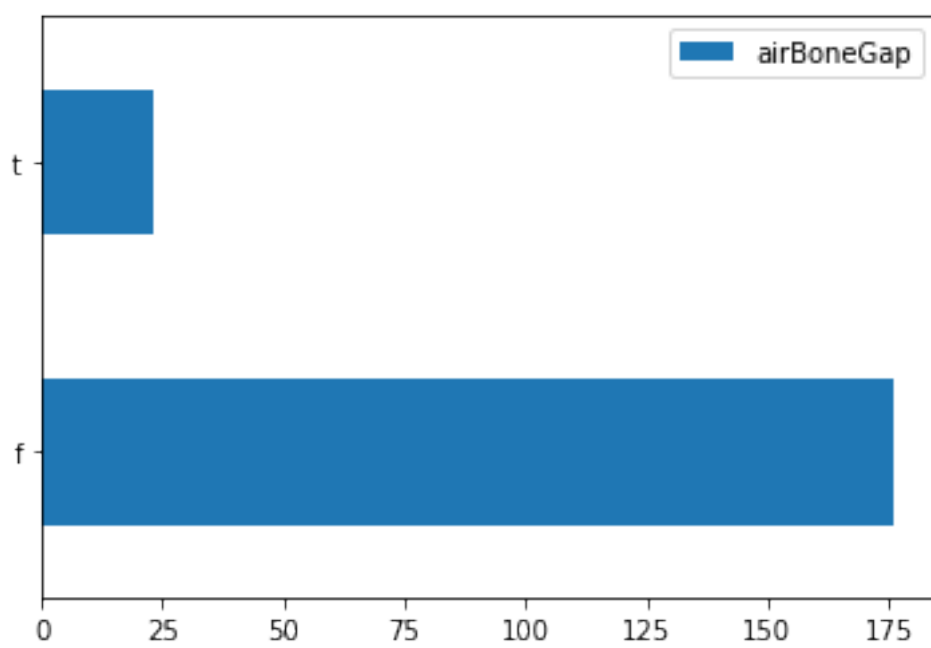
	class
0	cochlear_unknown
1	mixed_cochlear_age_fixation
2	mixed_cochlear_age_otitis_media
3	cochlear_age
4	cochlear_age
..	...
194	cochlear_age
195	mixed_cochlear_age_otitis_media
196	possible_brainstem_disorder
197	cochlear_age
198	cochlear_age

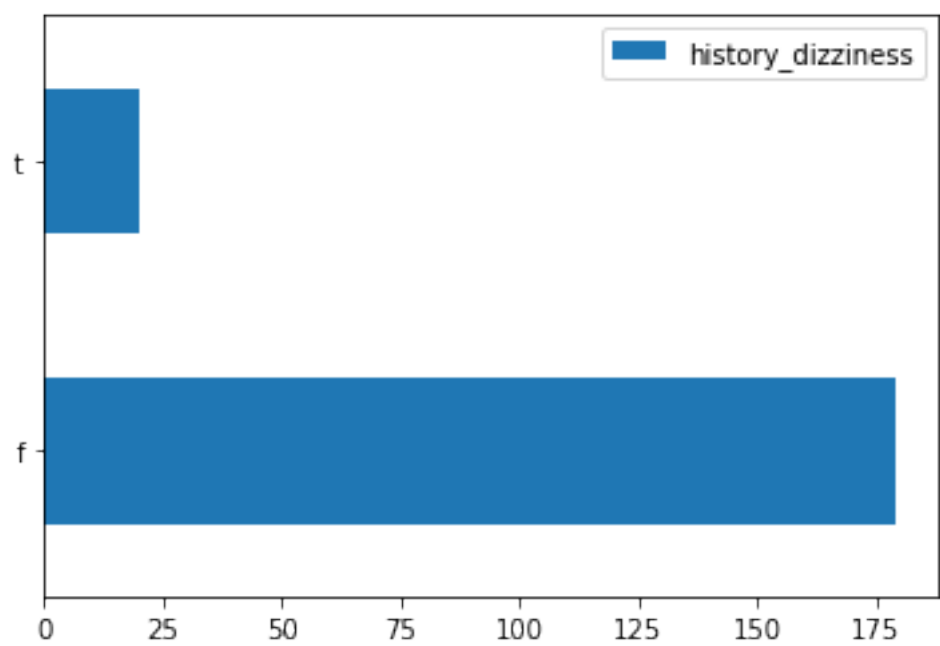
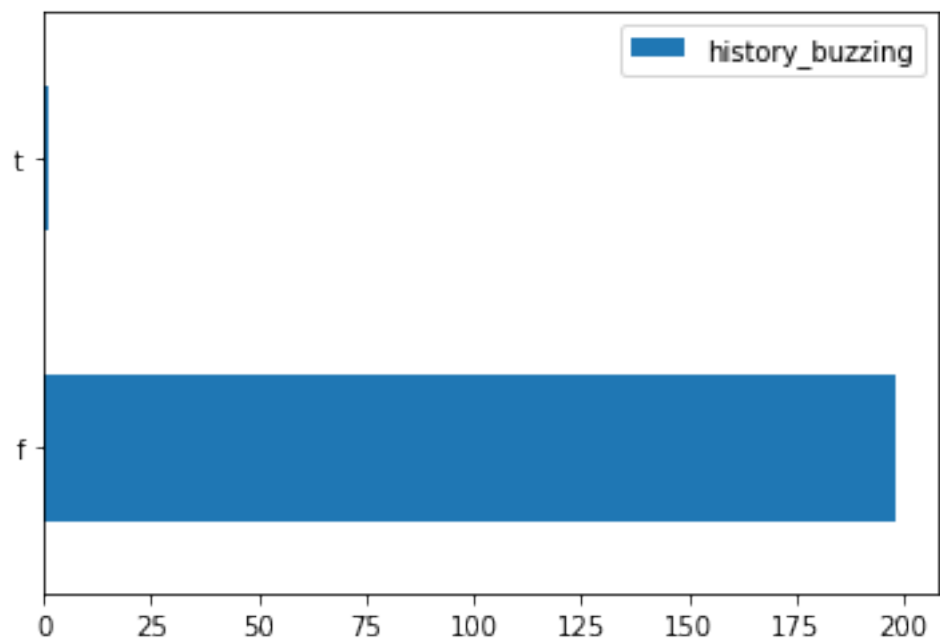
[199 rows x 64 columns]

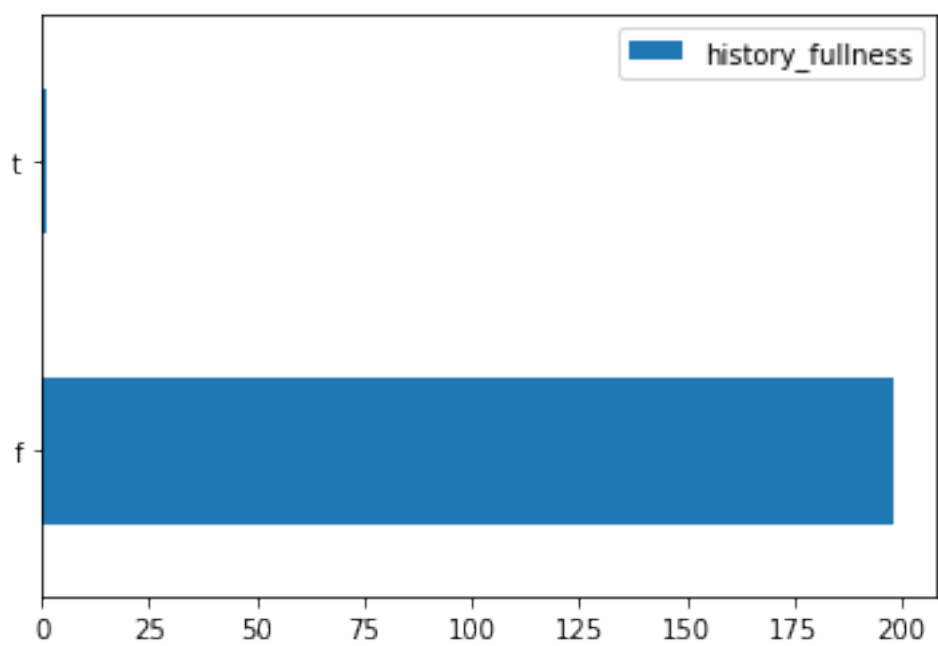
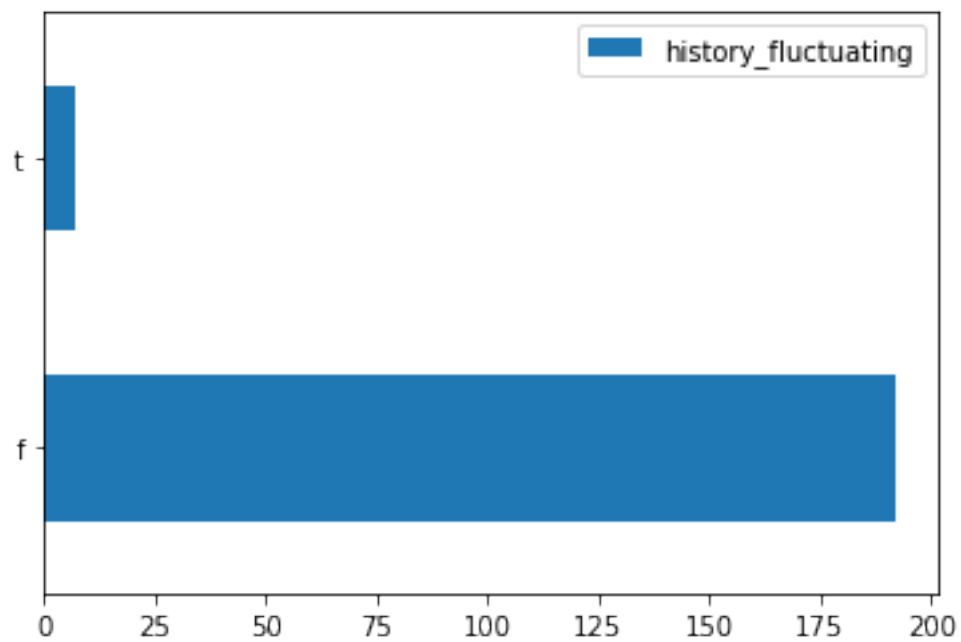
1.5 6. Graph of each values of each column

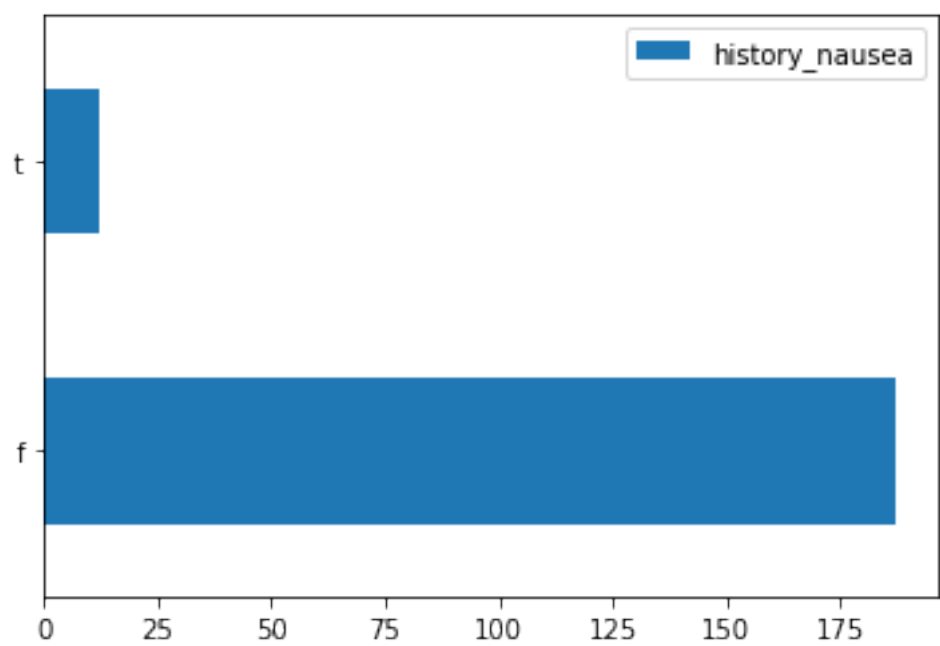
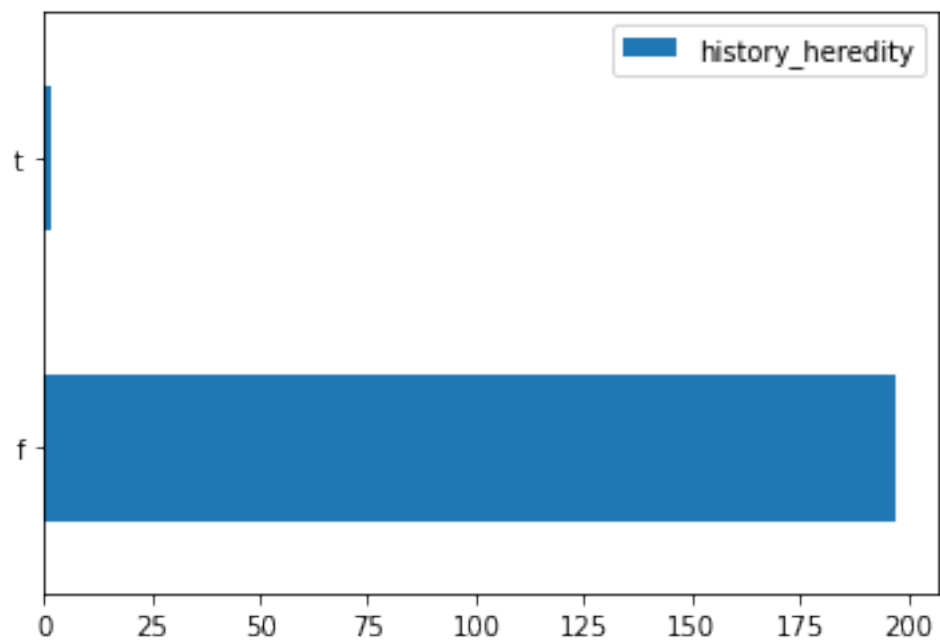
```
[12]: for col in data_df.columns:  
      each_value_count = pd.DataFrame(data_df[col].value_counts())  
      each_value_count.plot.barh()
```

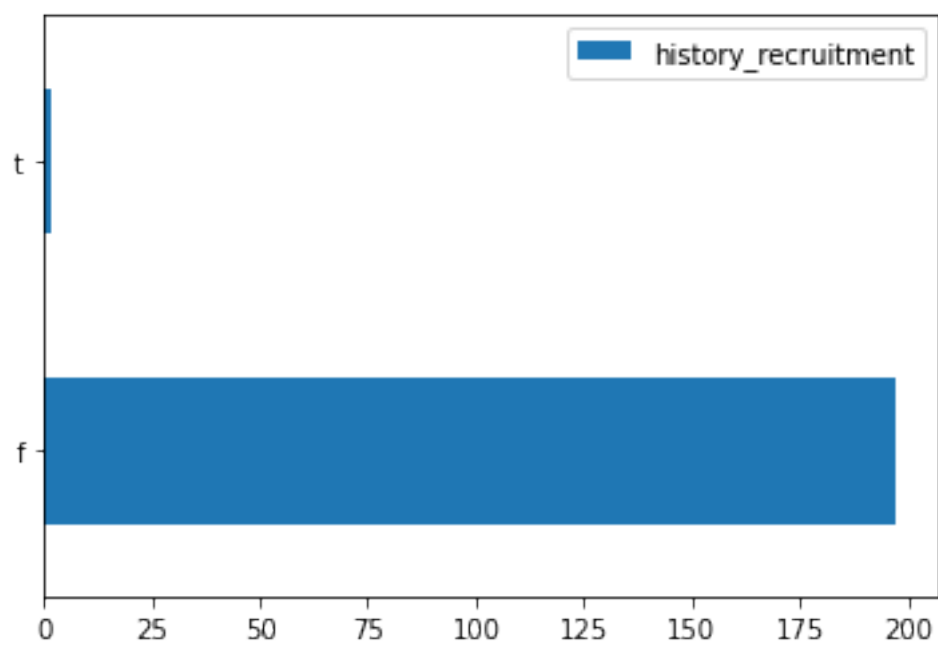
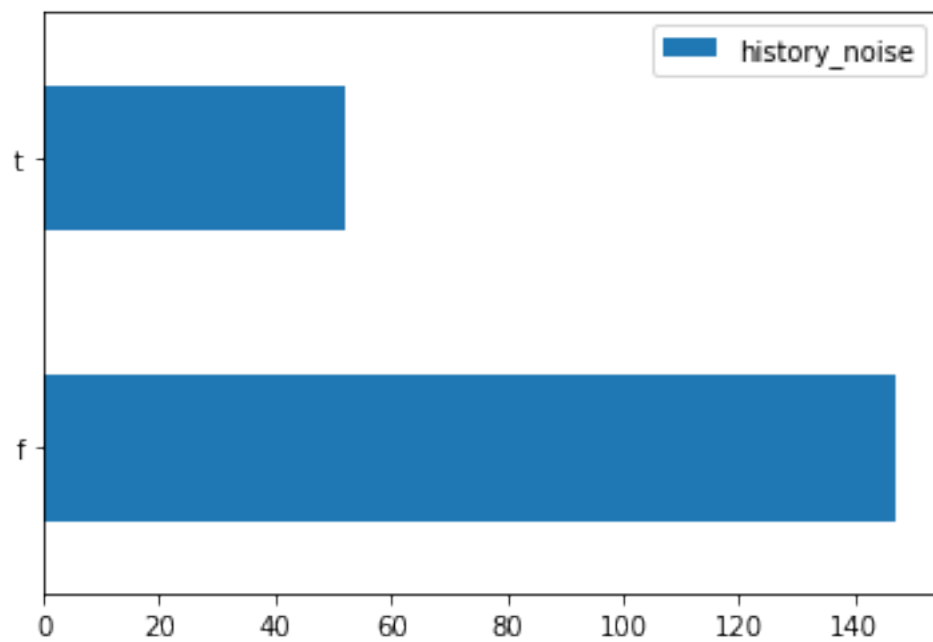


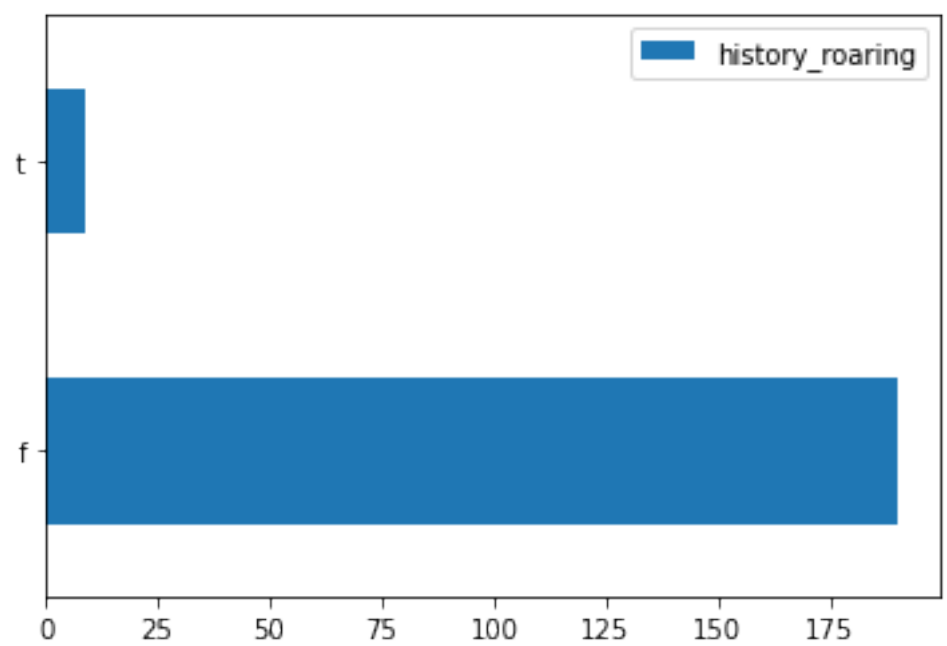
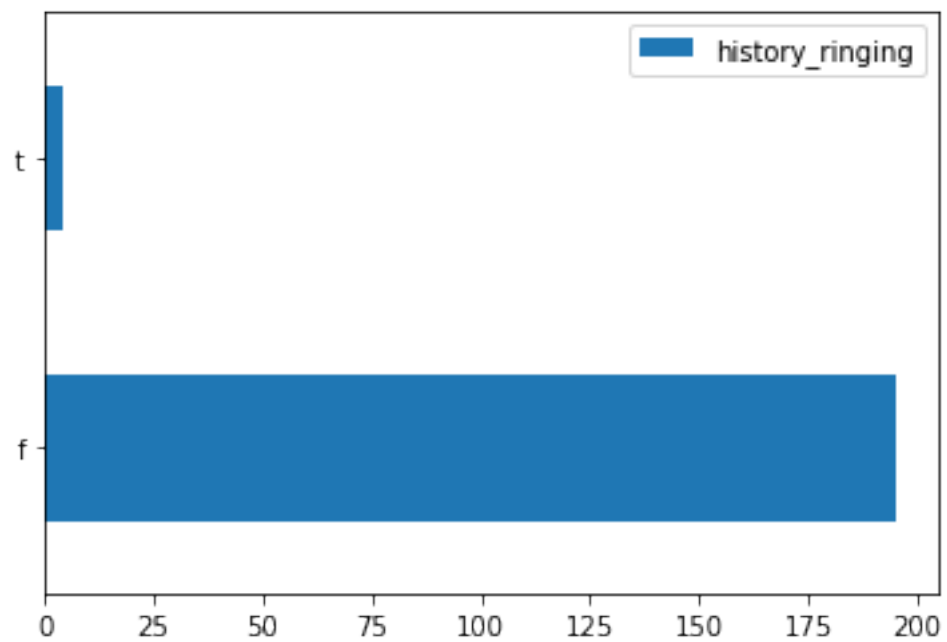


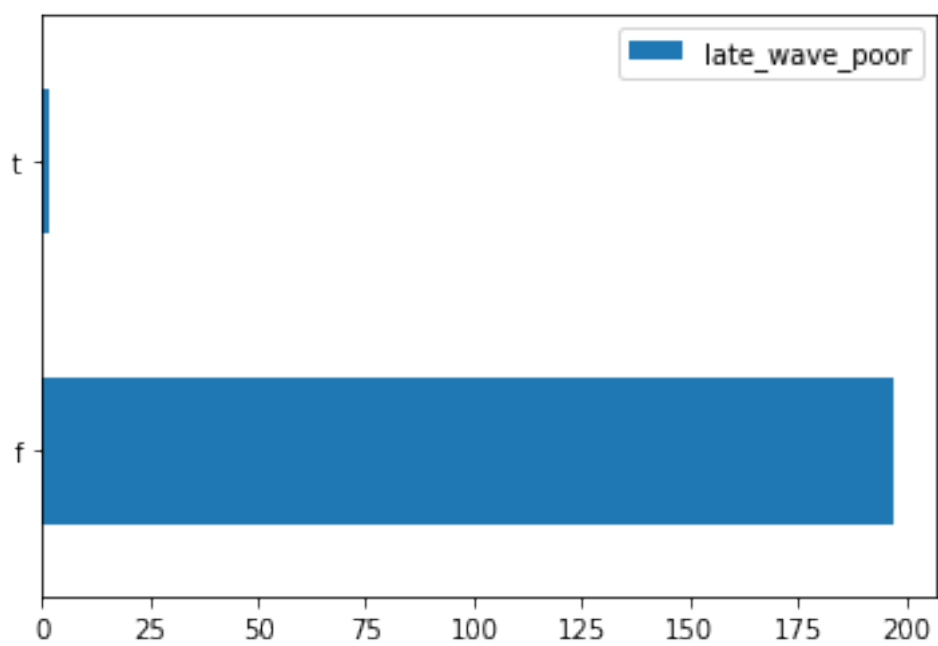
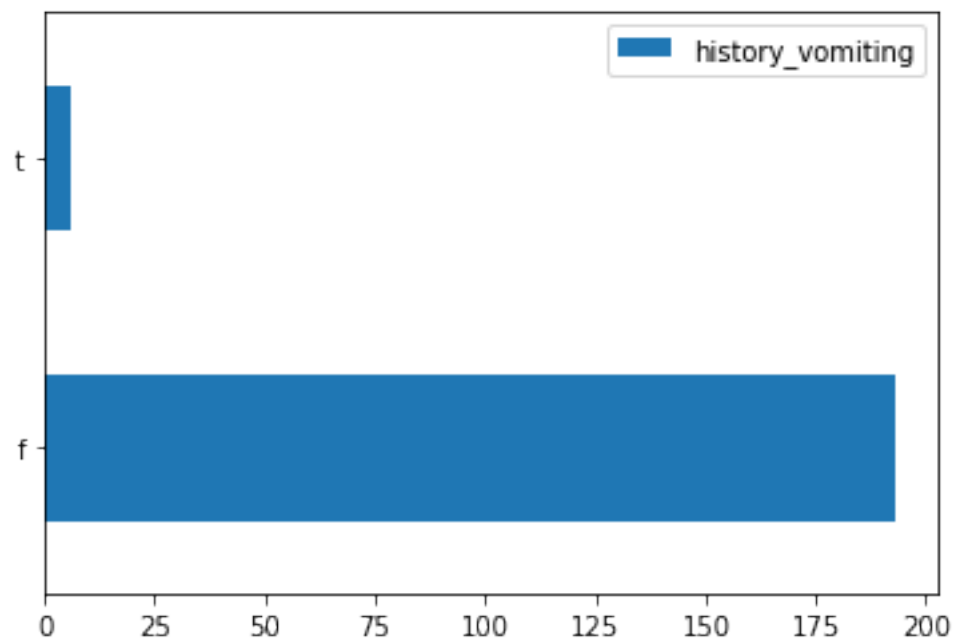


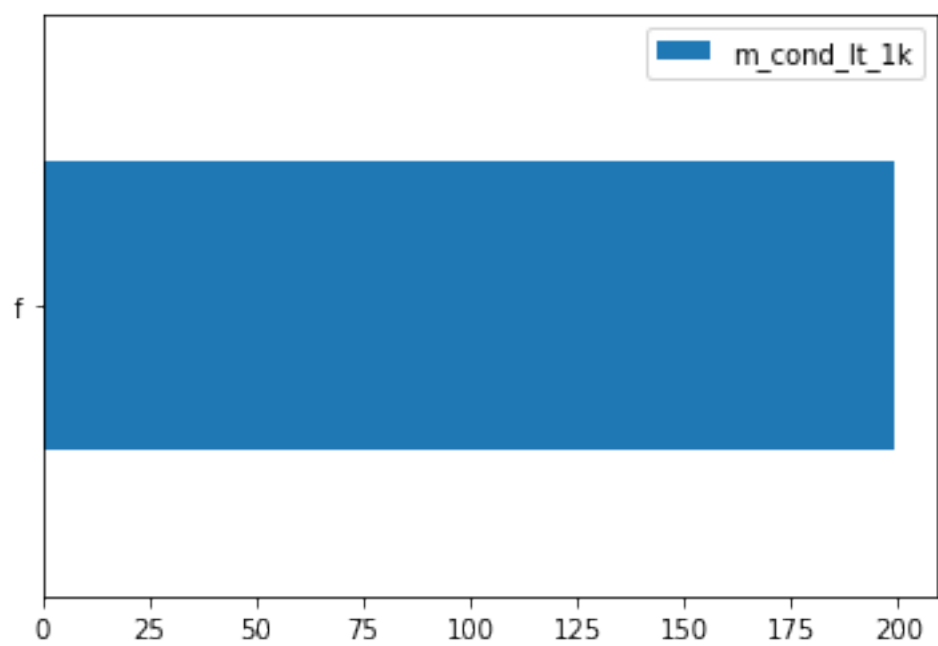
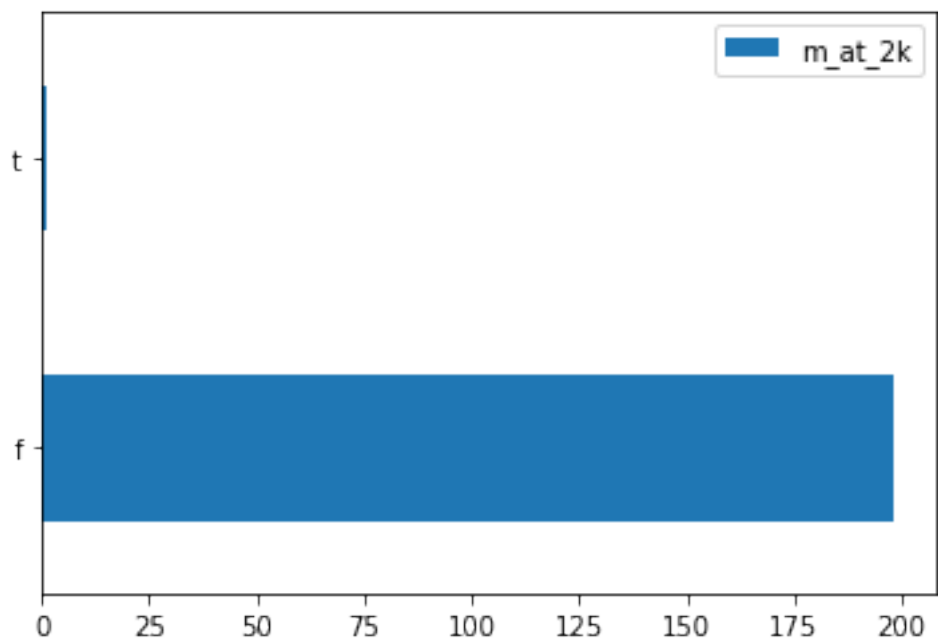


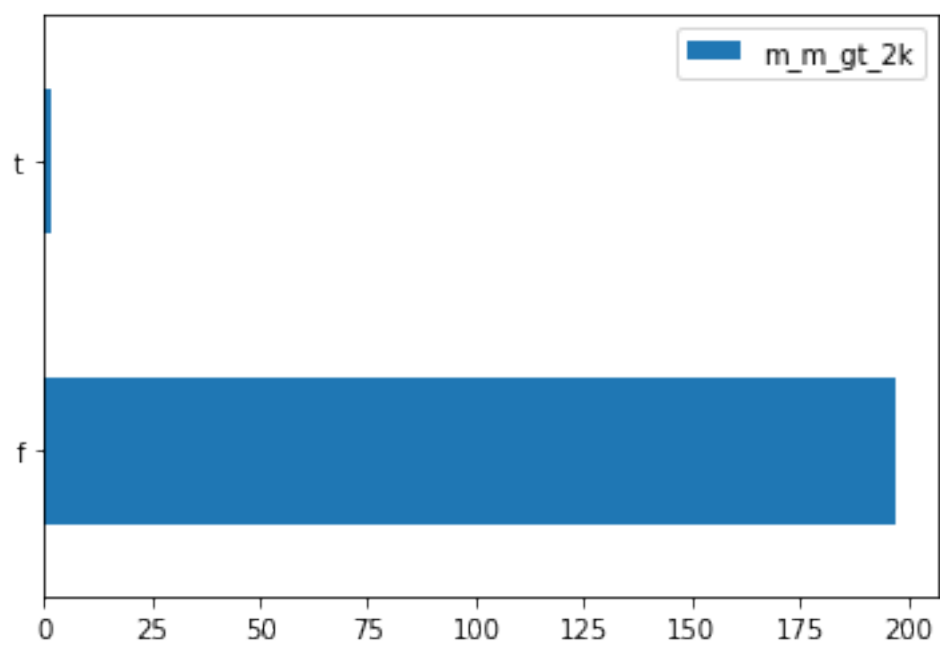
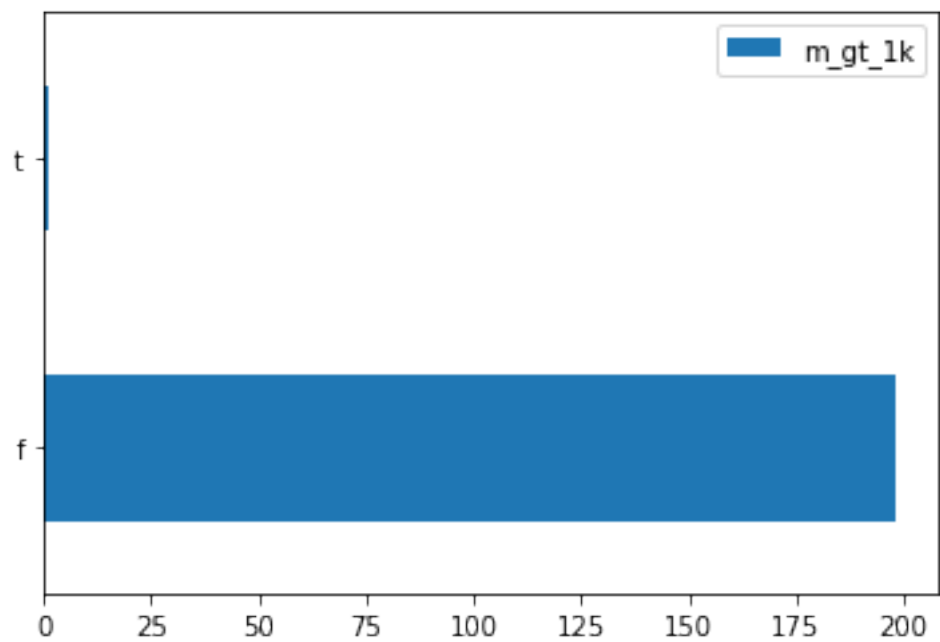


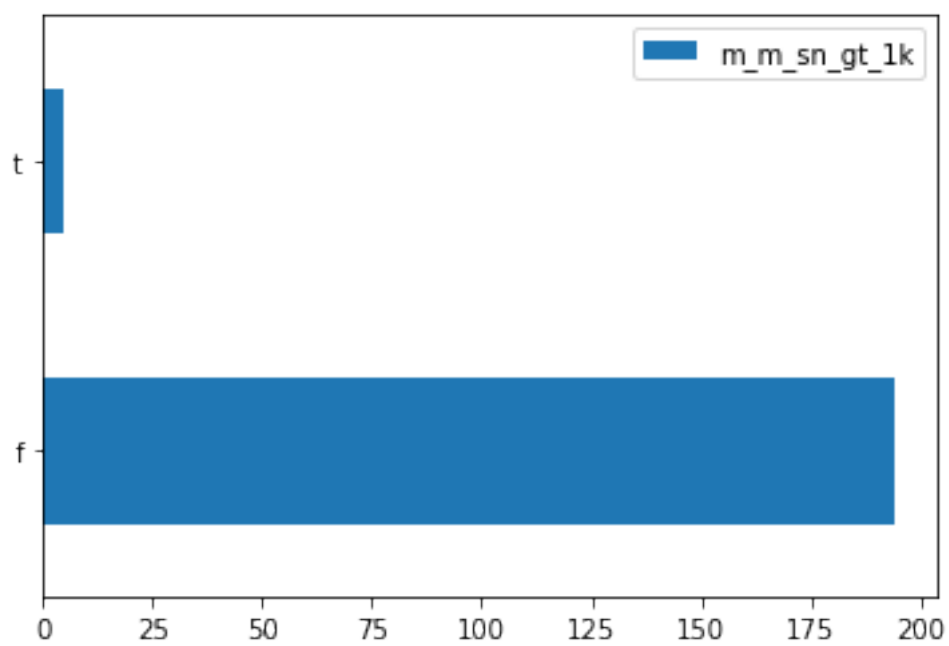
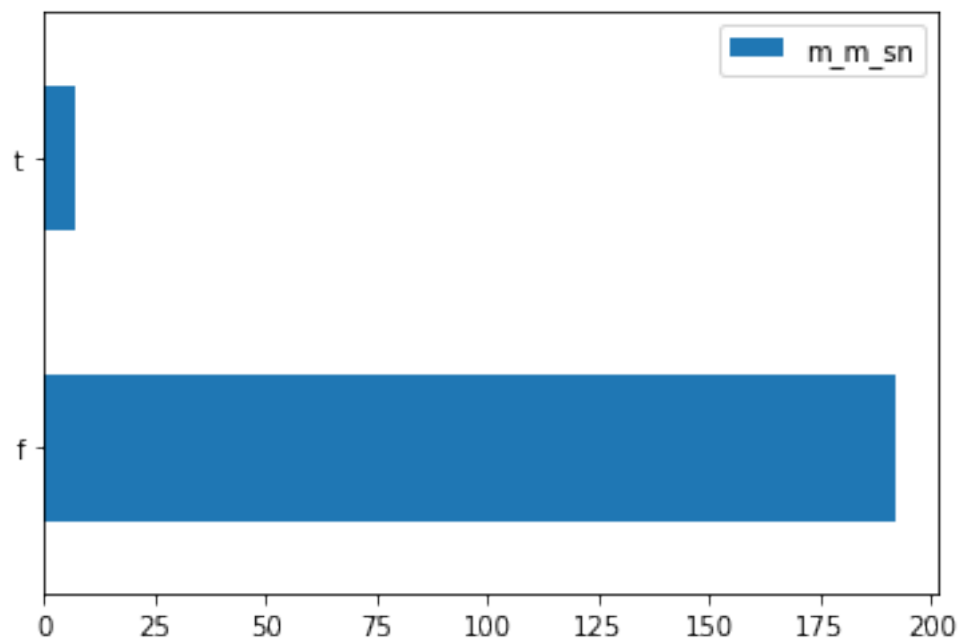


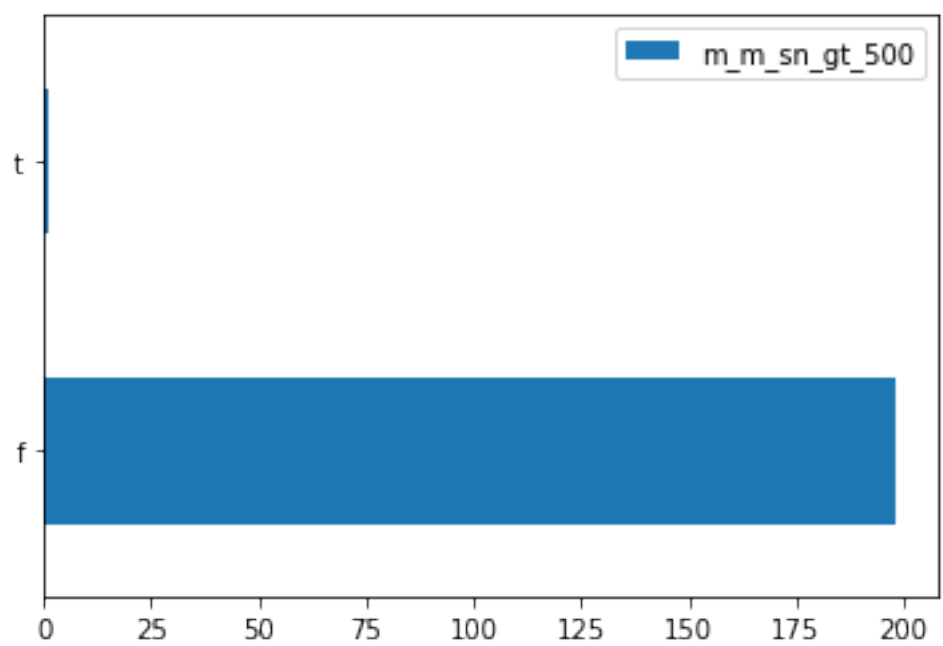
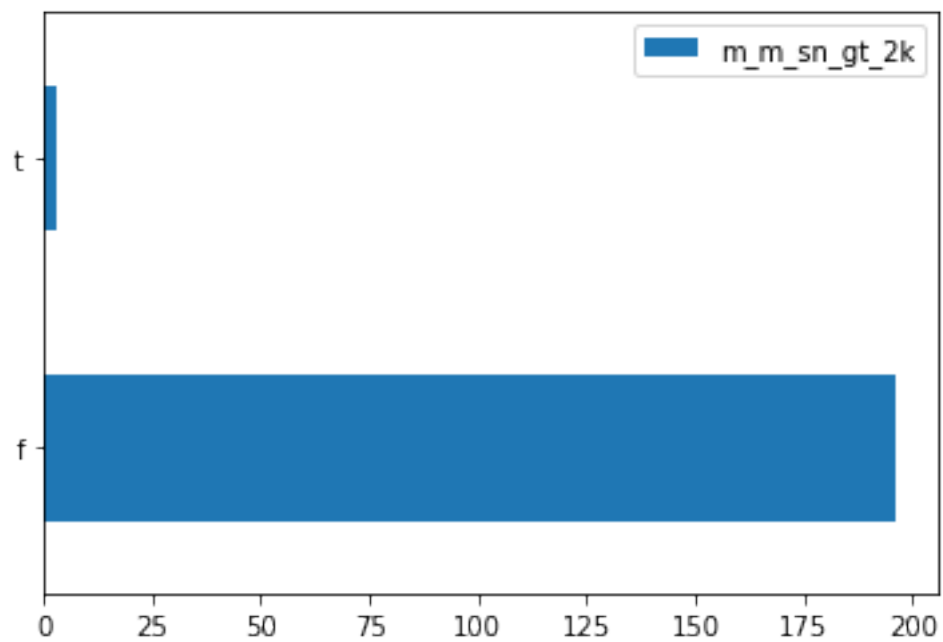


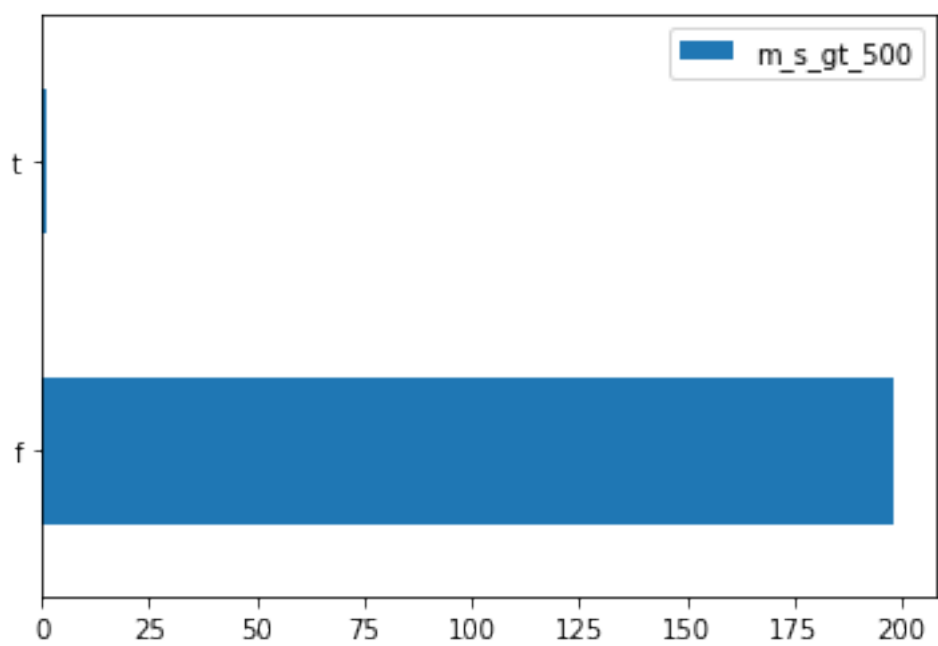
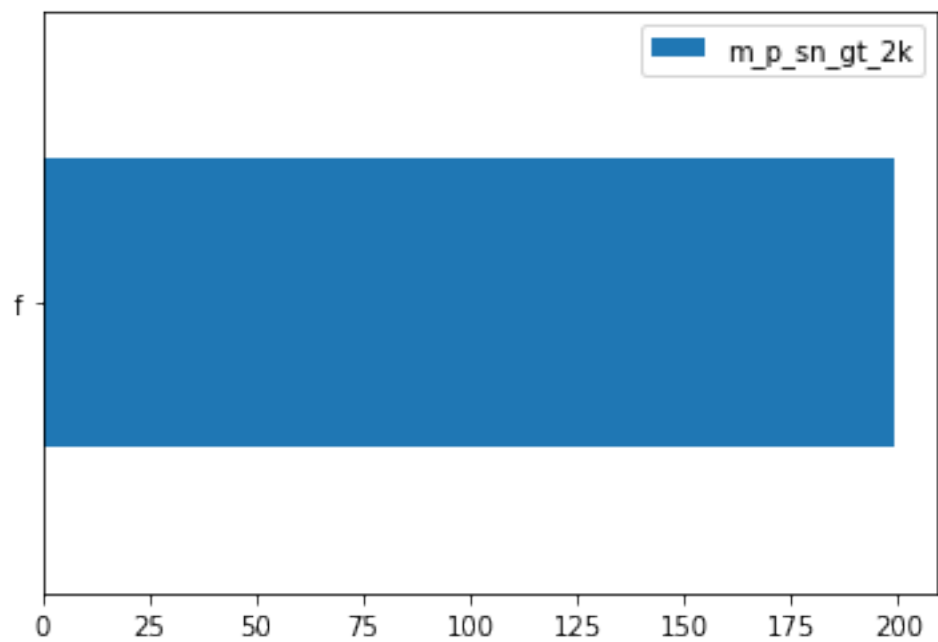


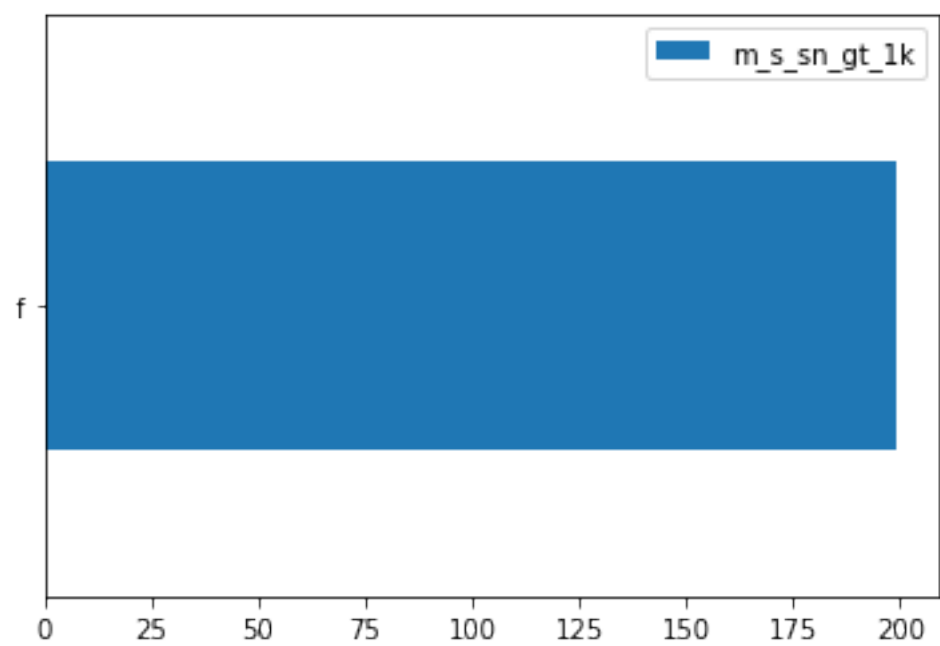
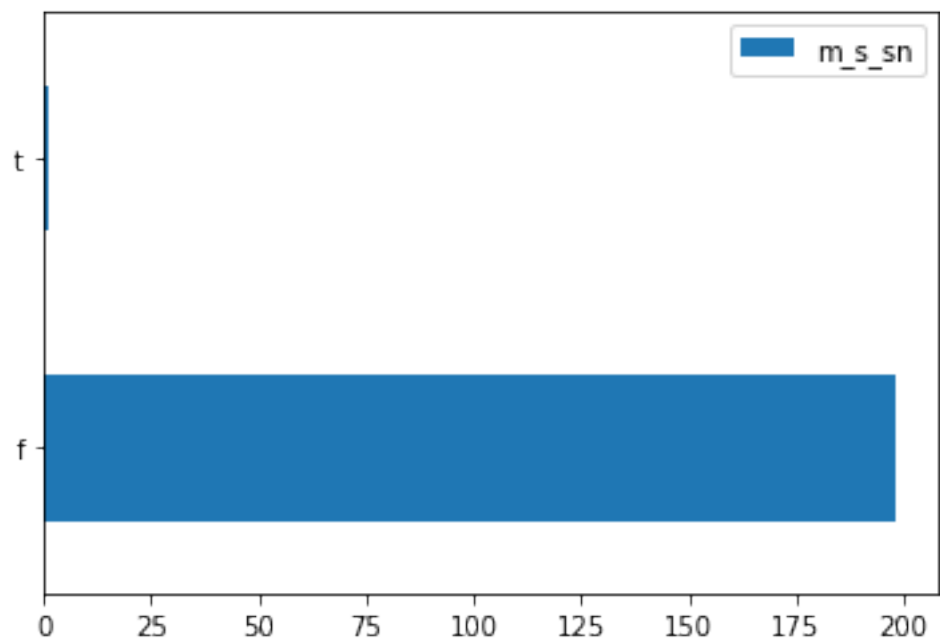


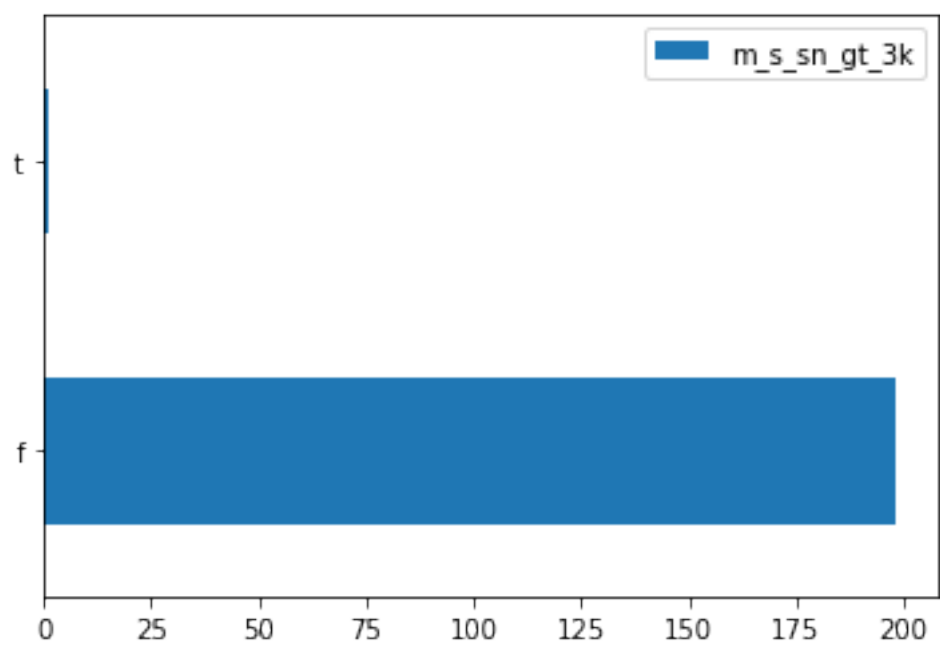
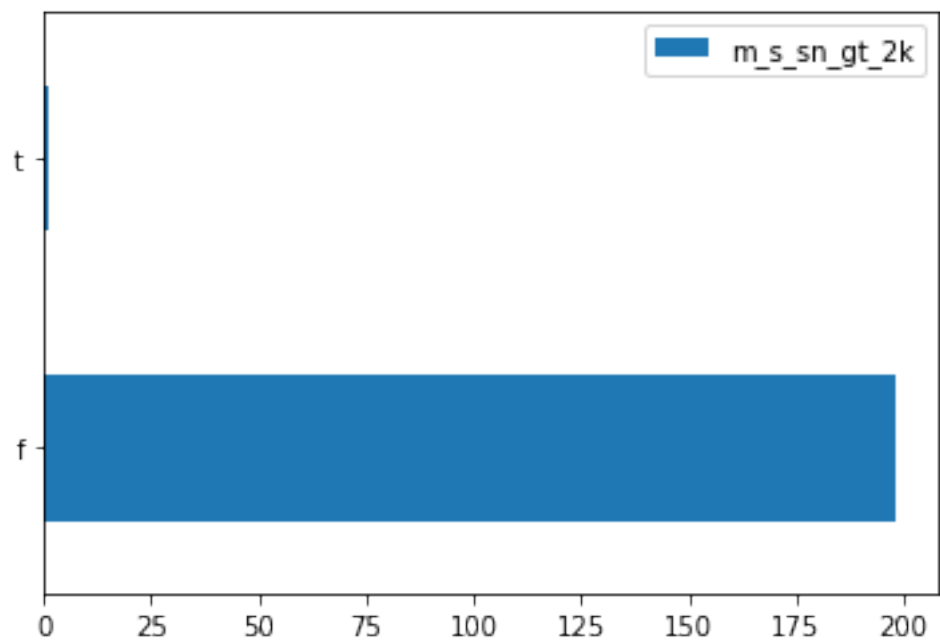


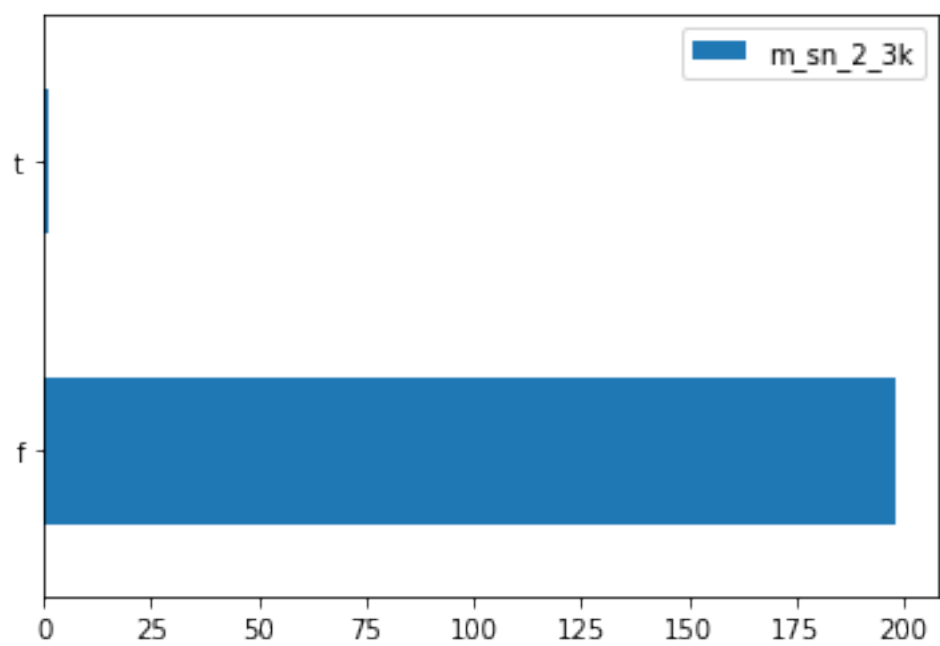
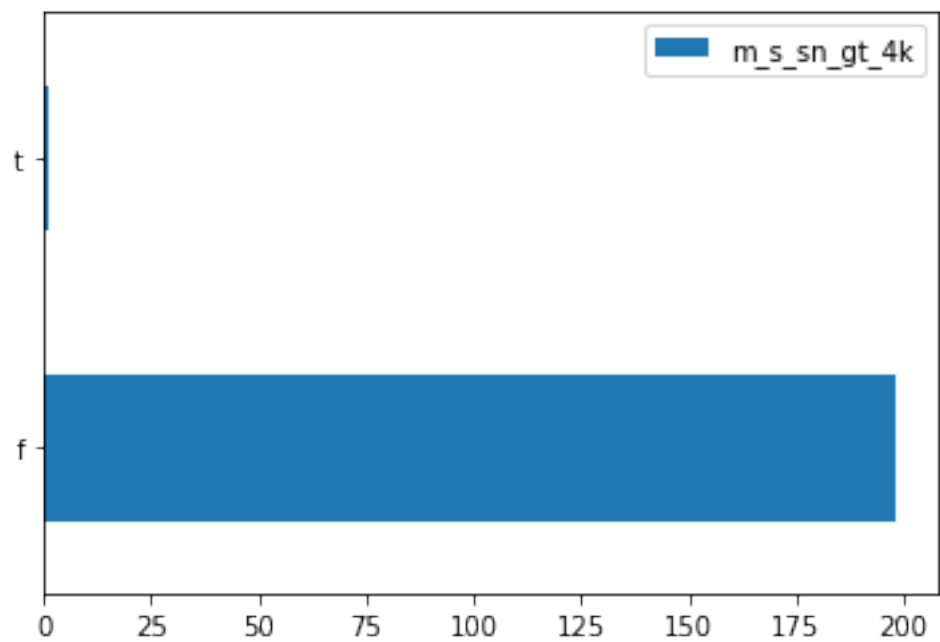


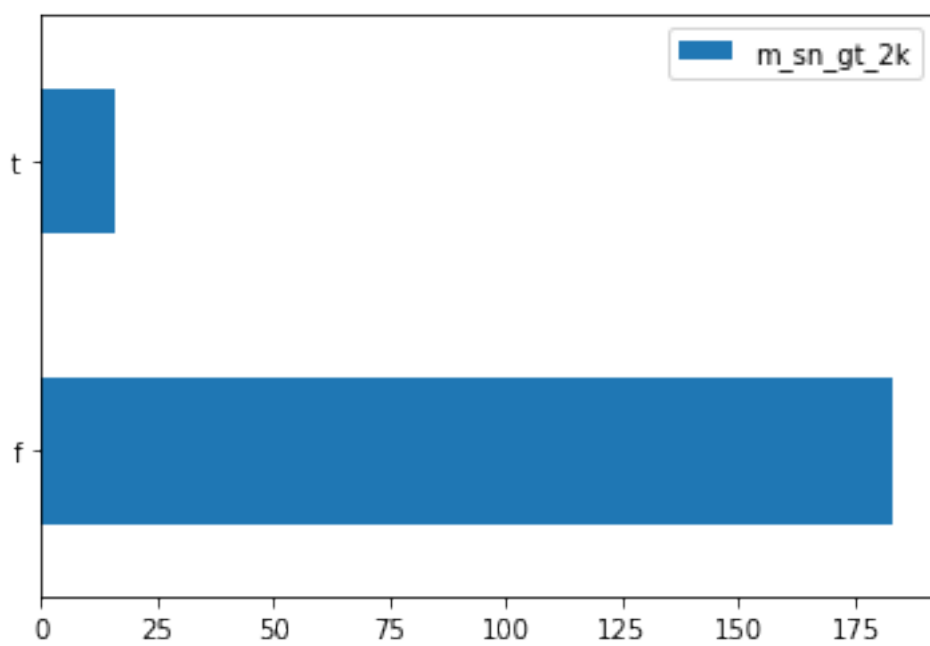
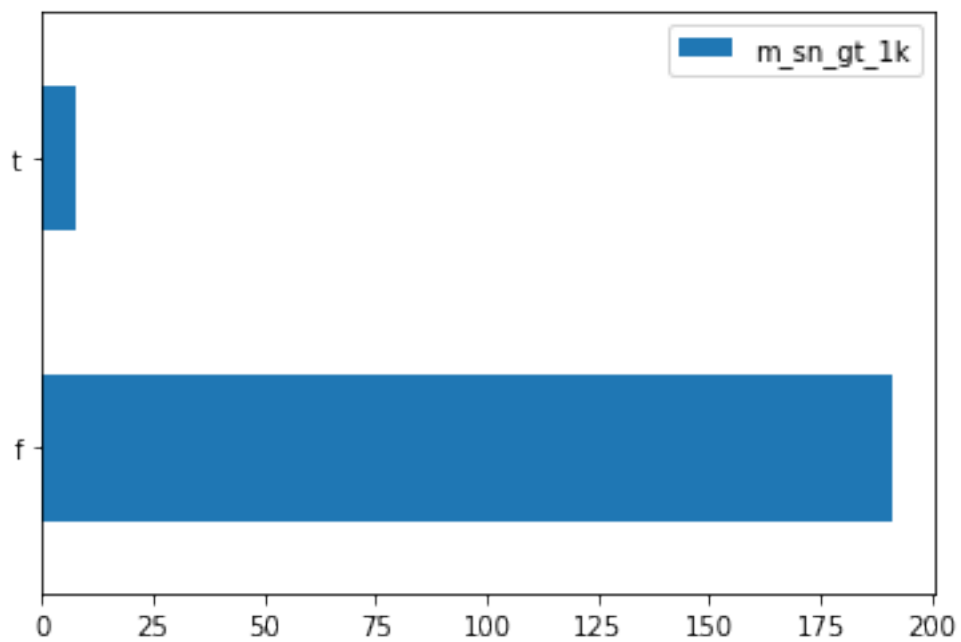


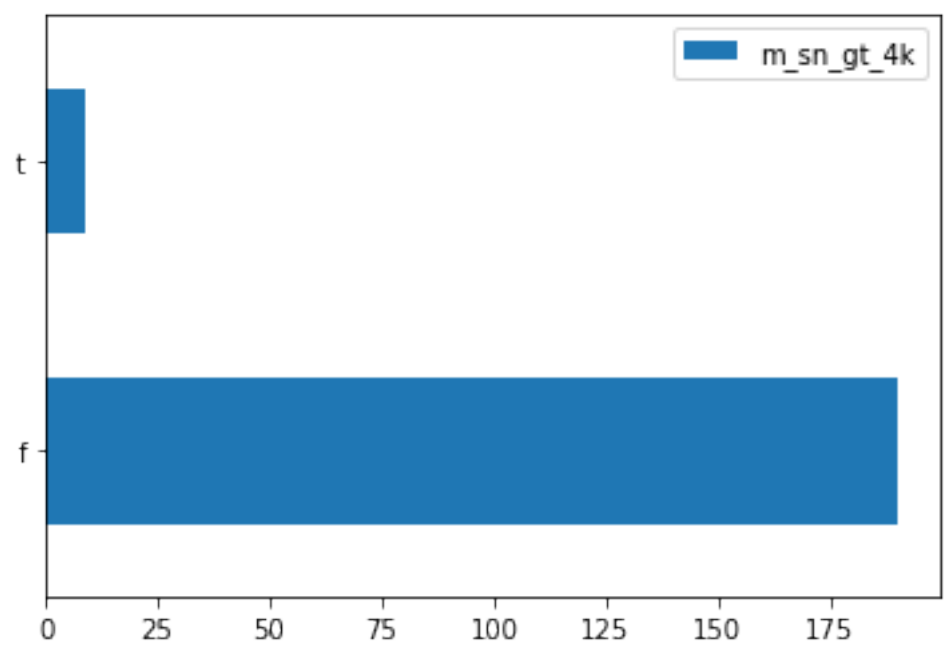
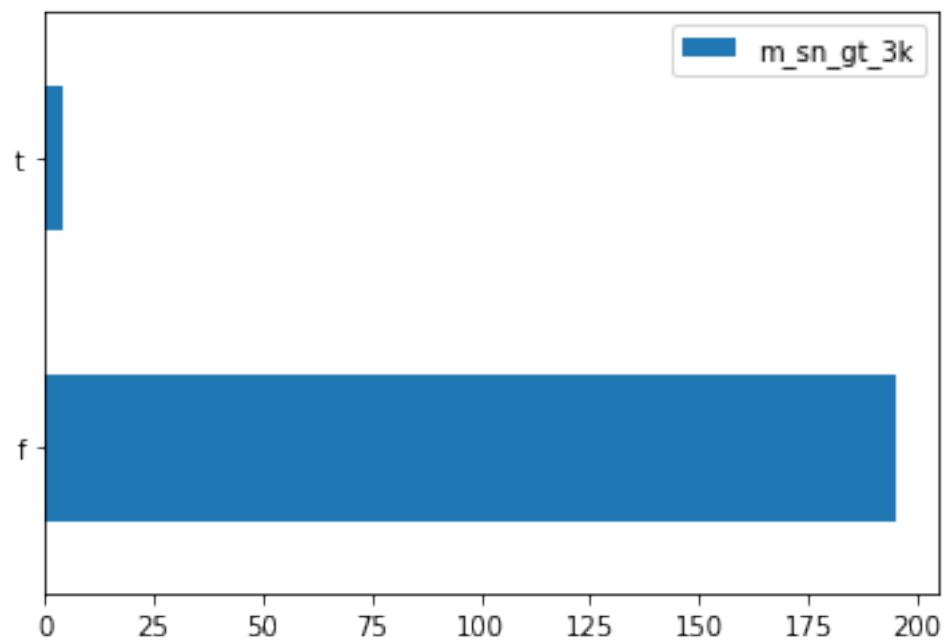


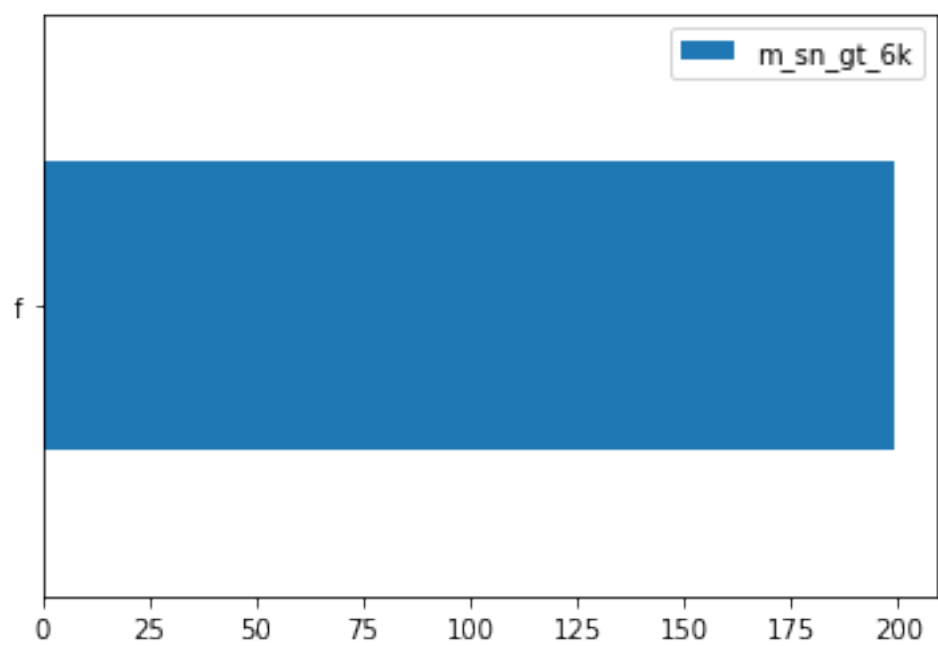
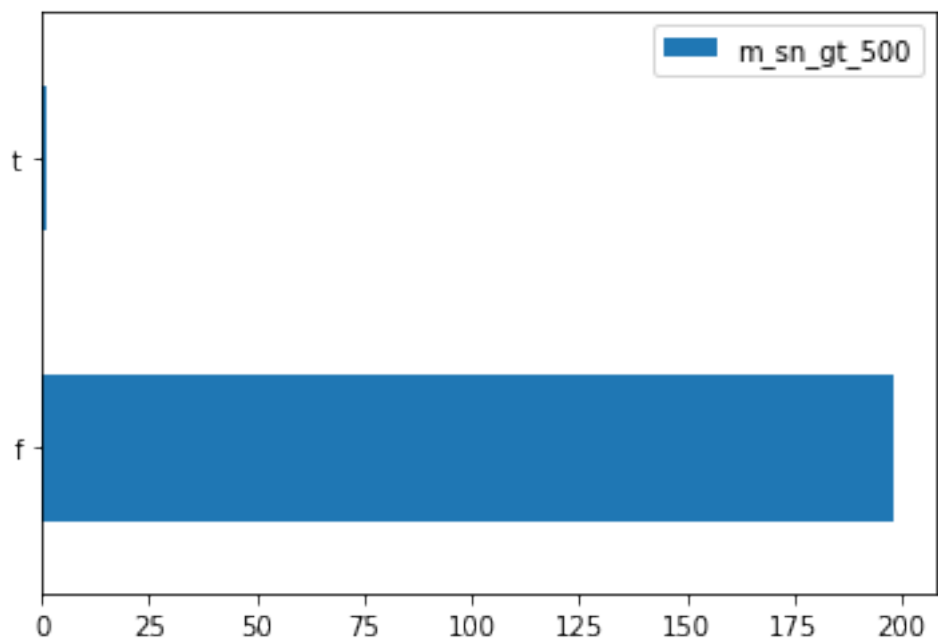


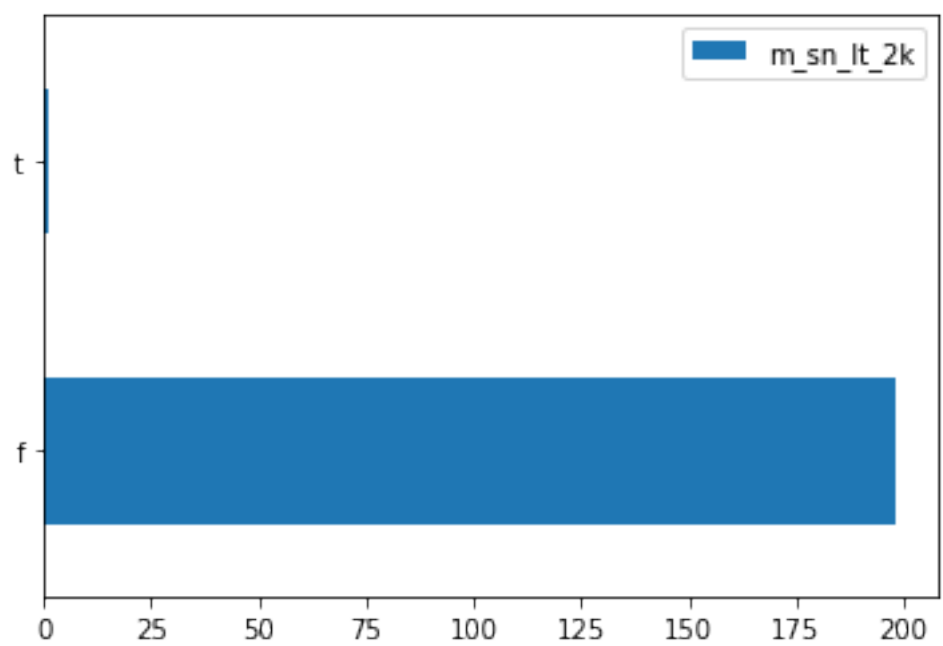
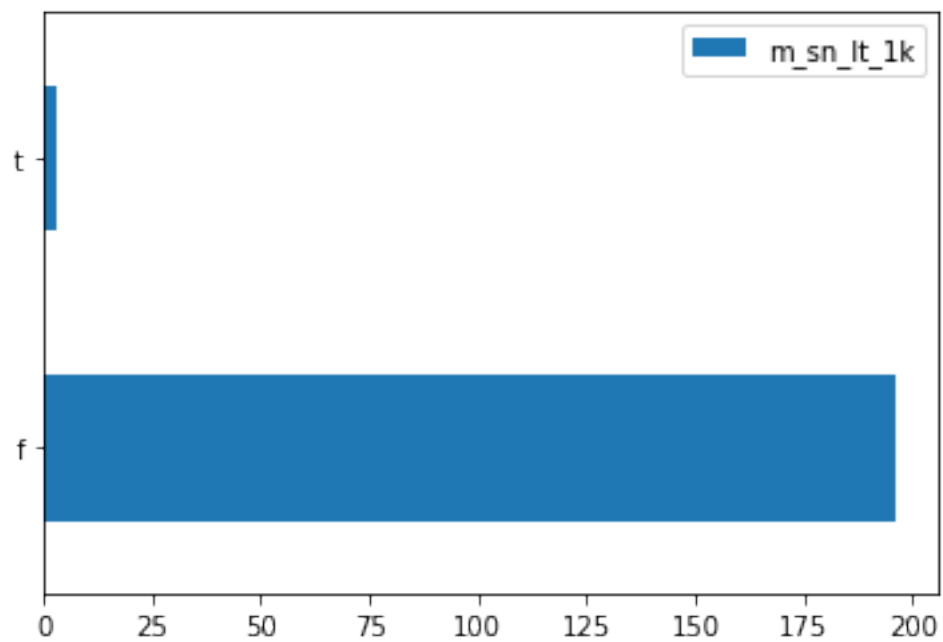


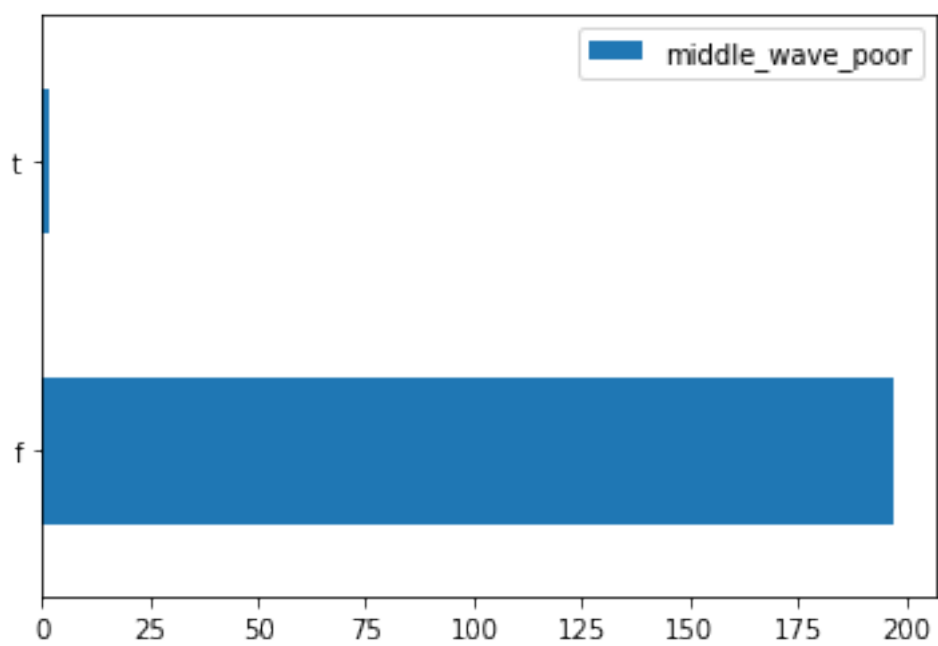
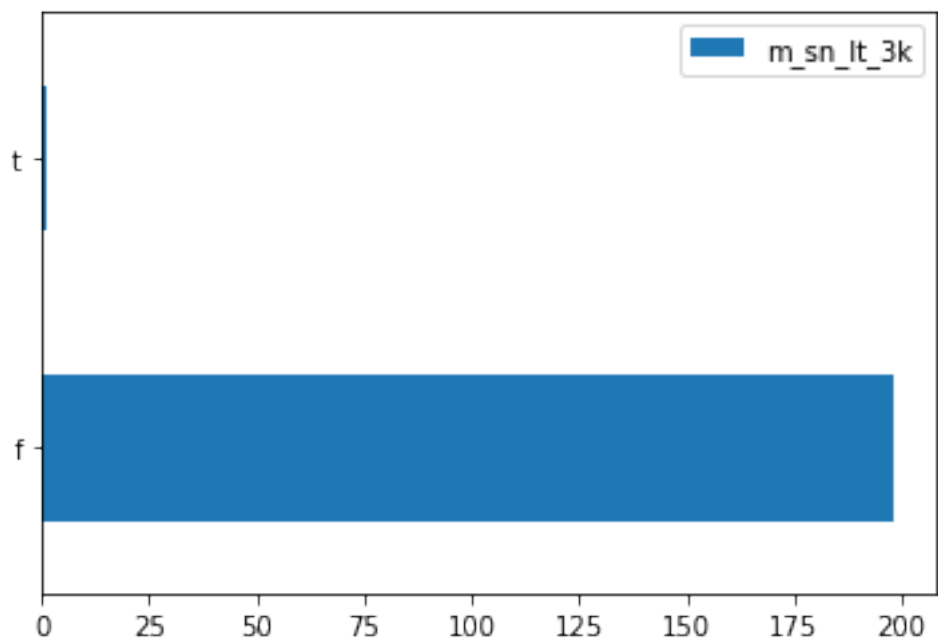


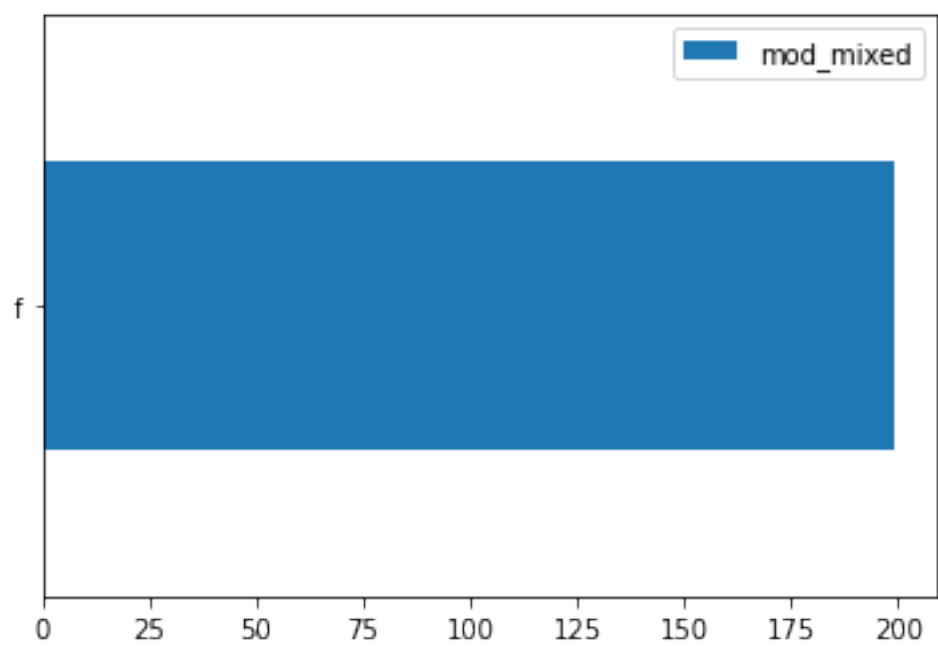
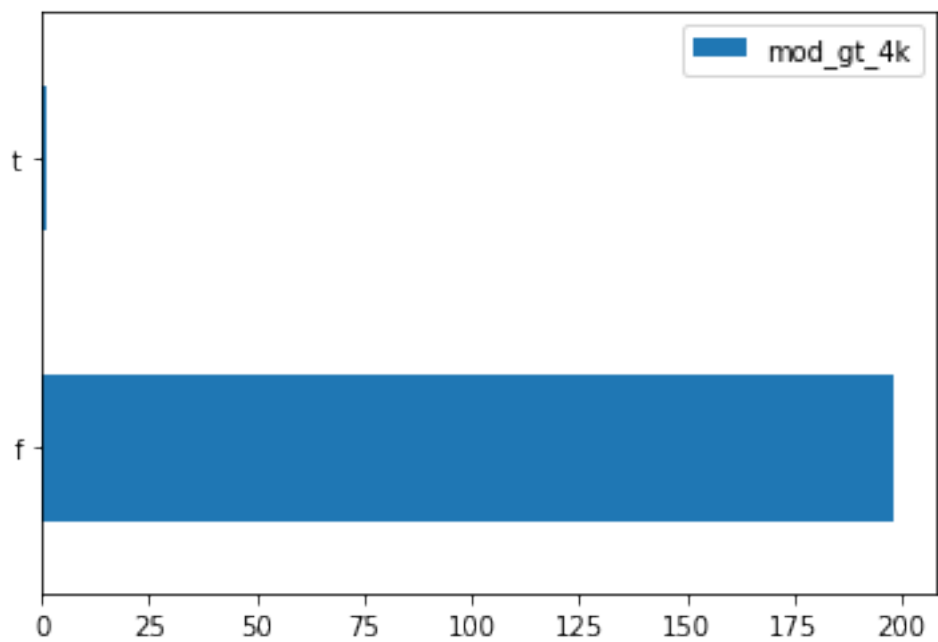


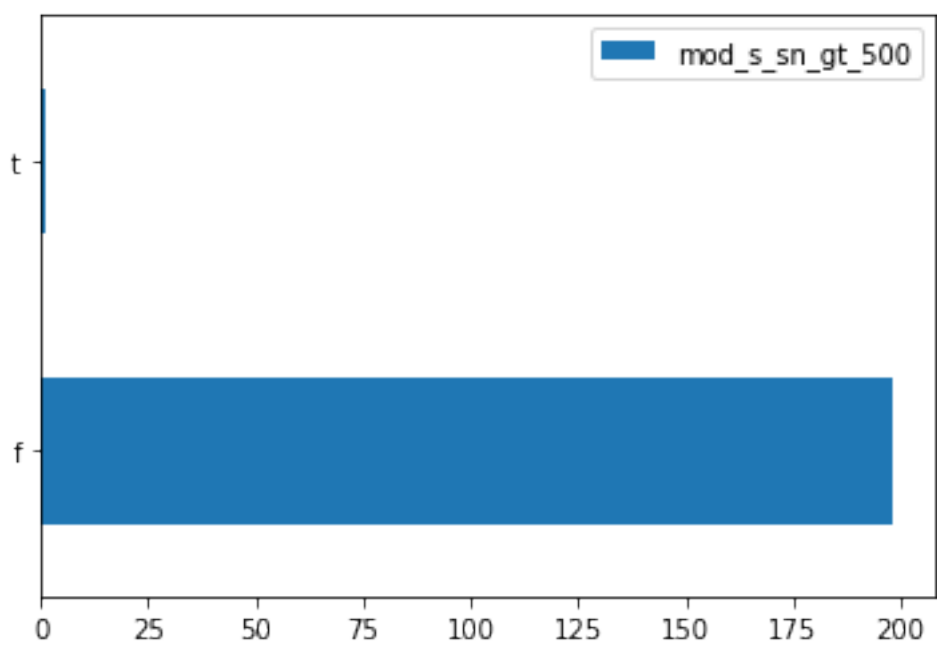
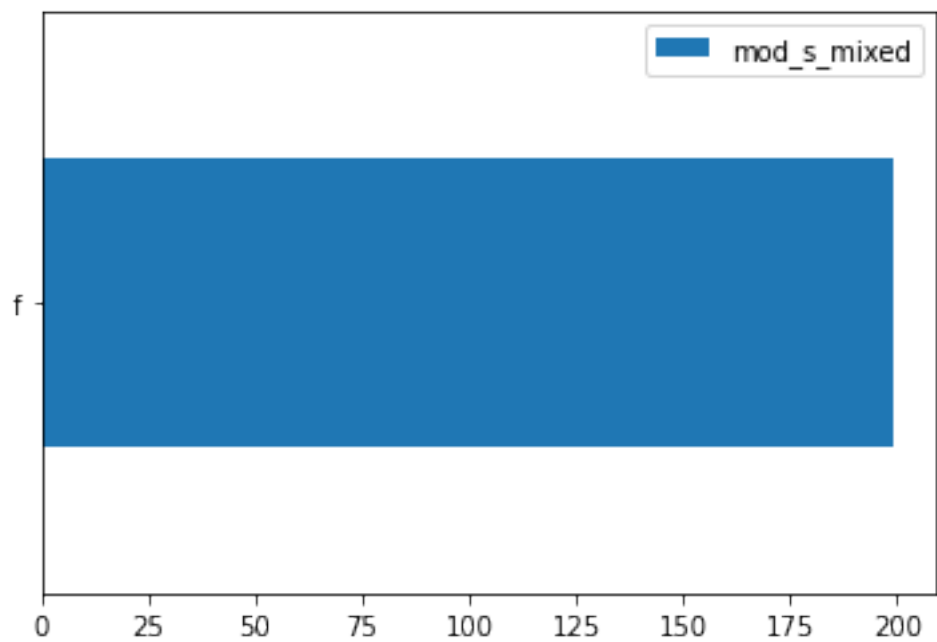


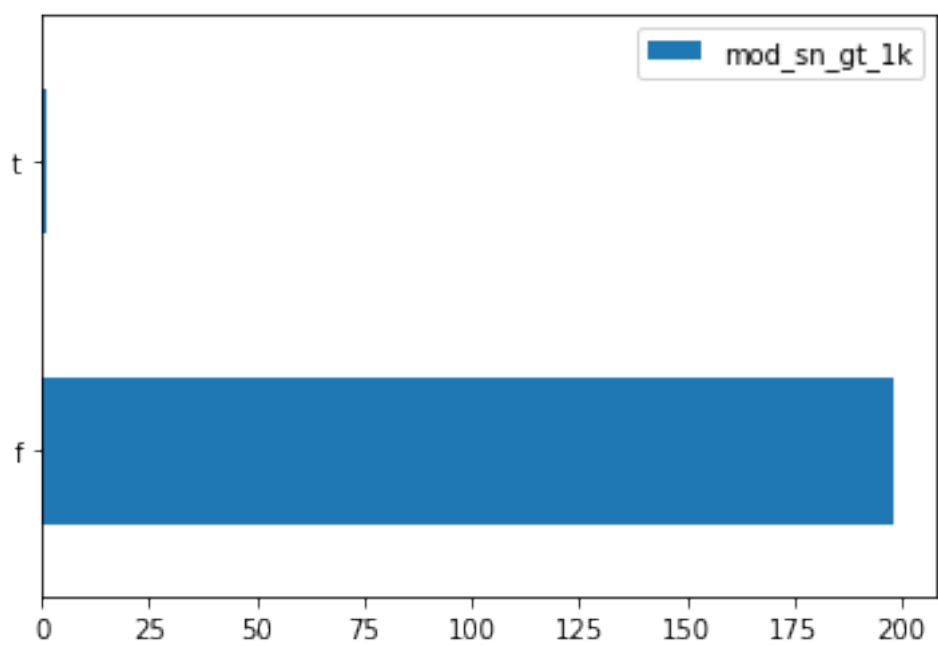
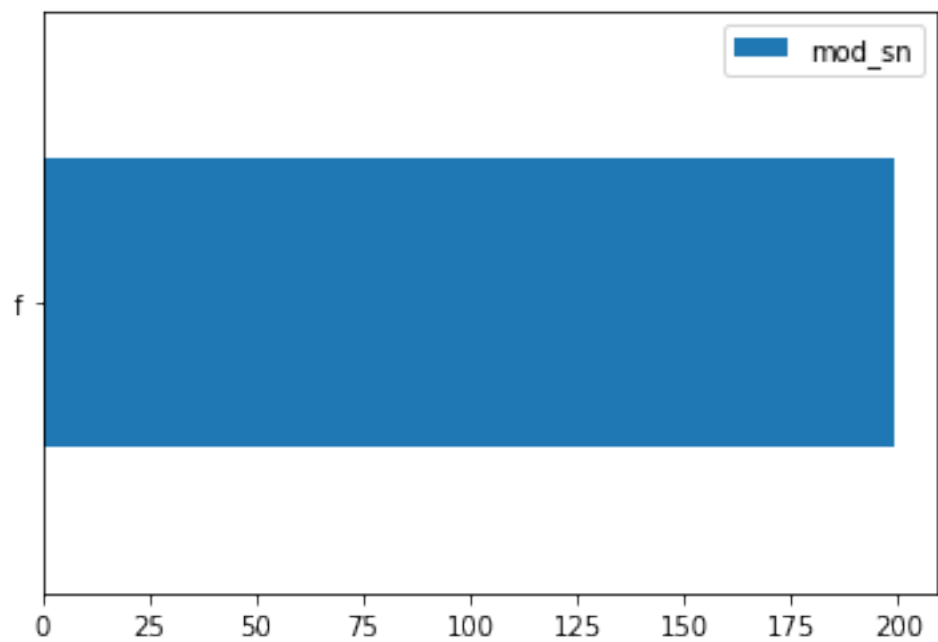


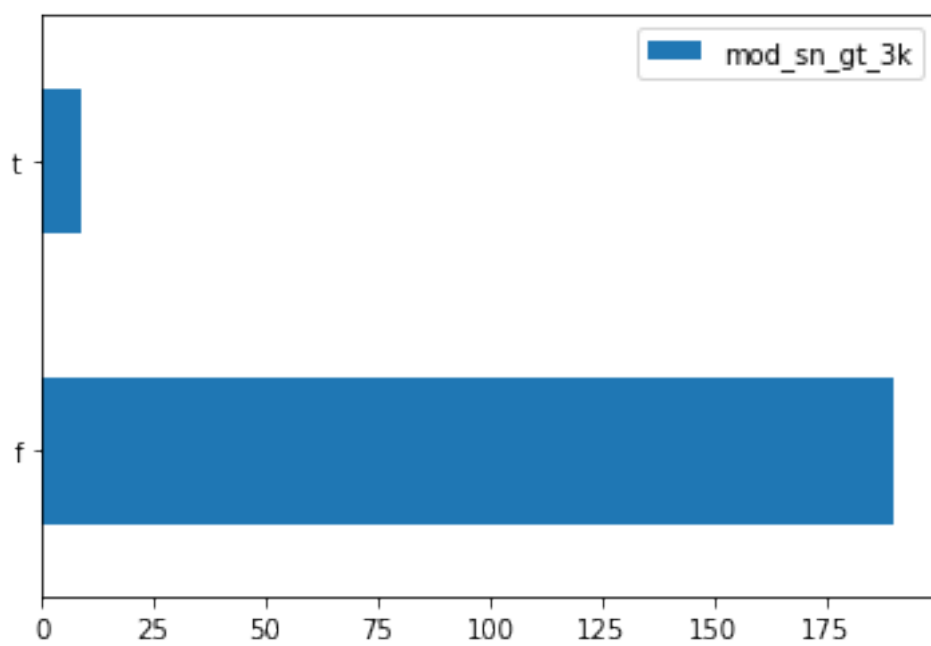
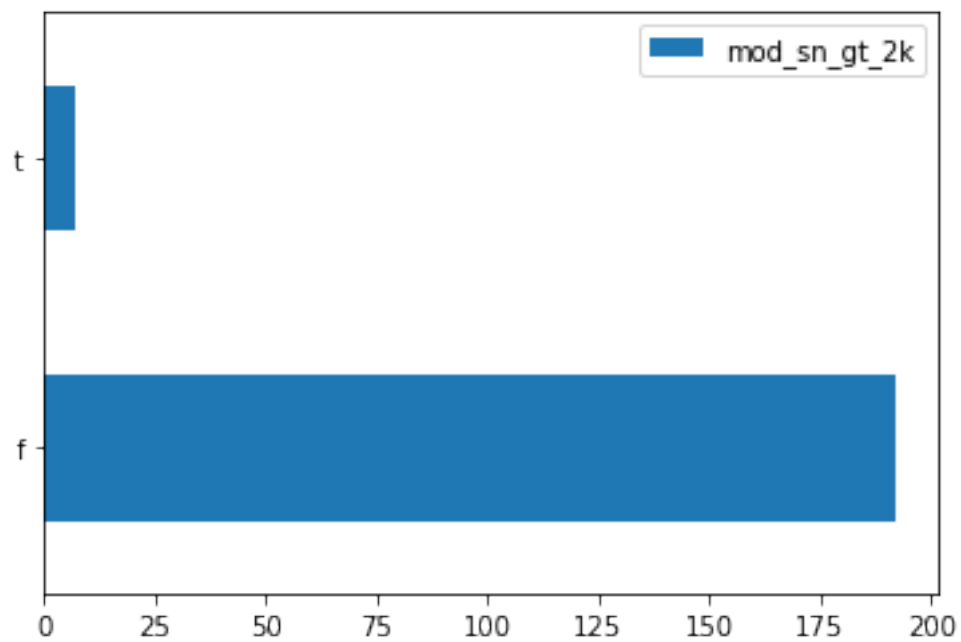


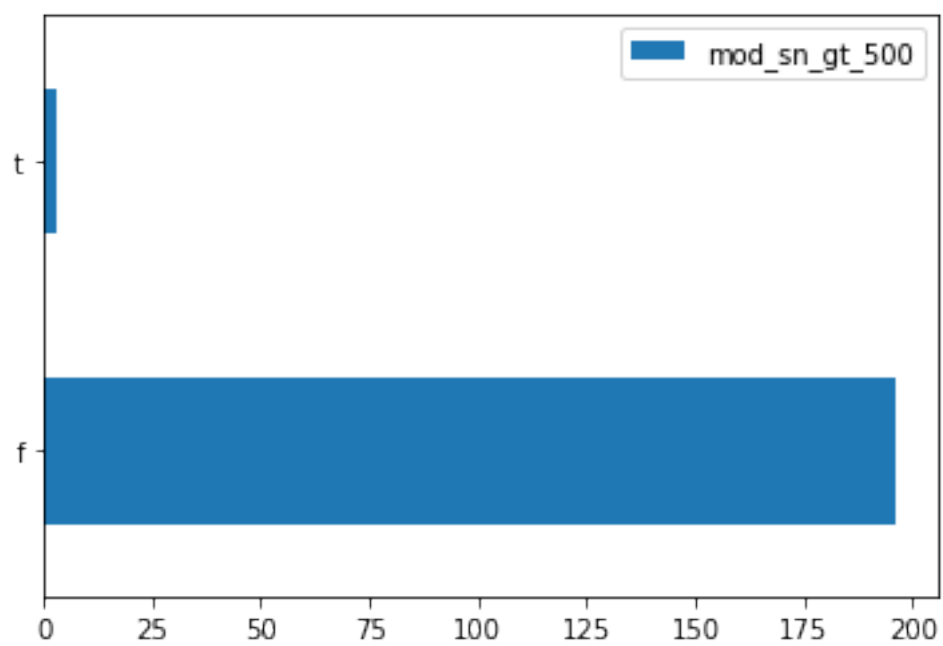
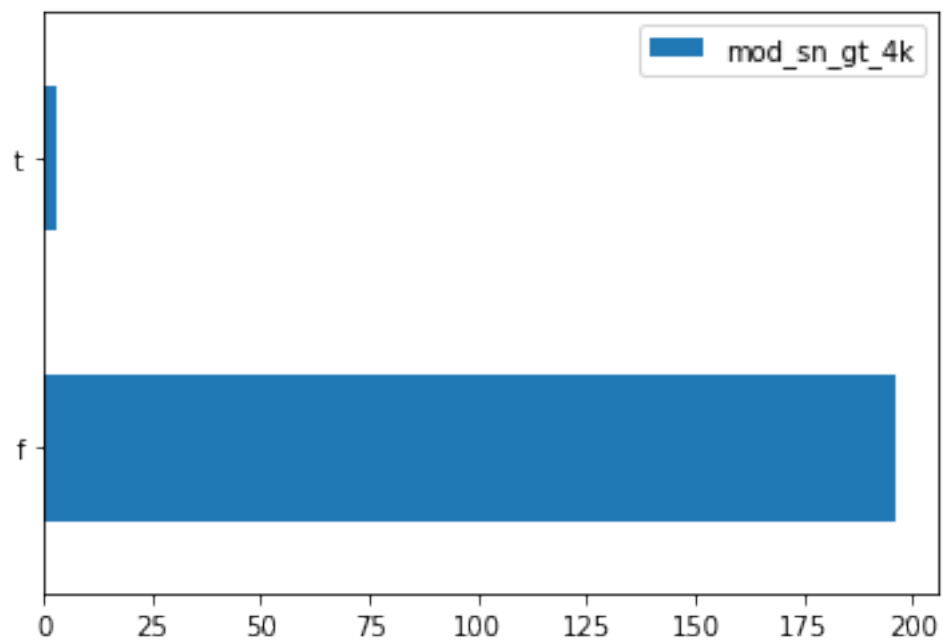


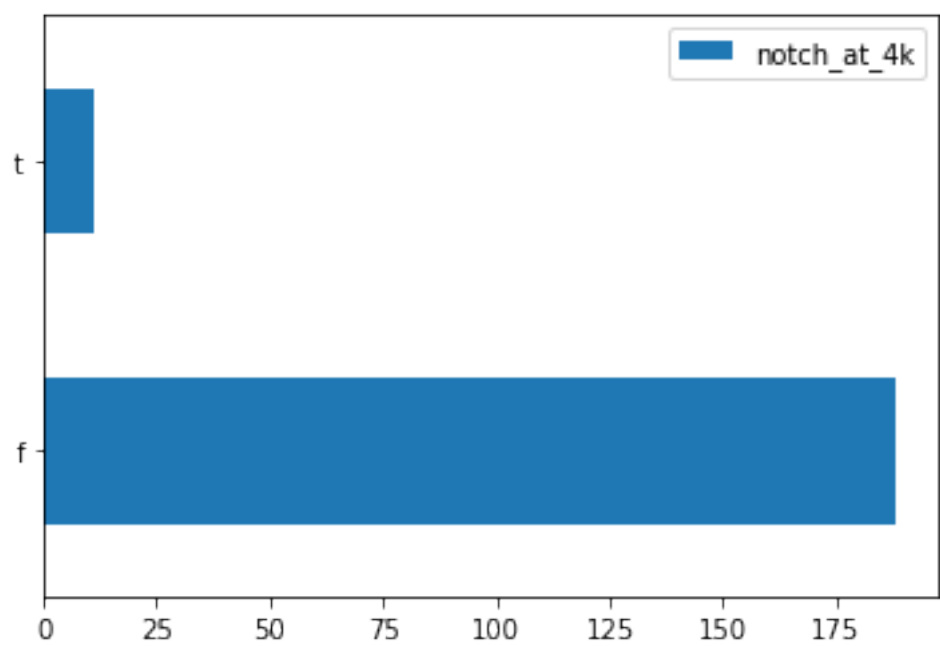
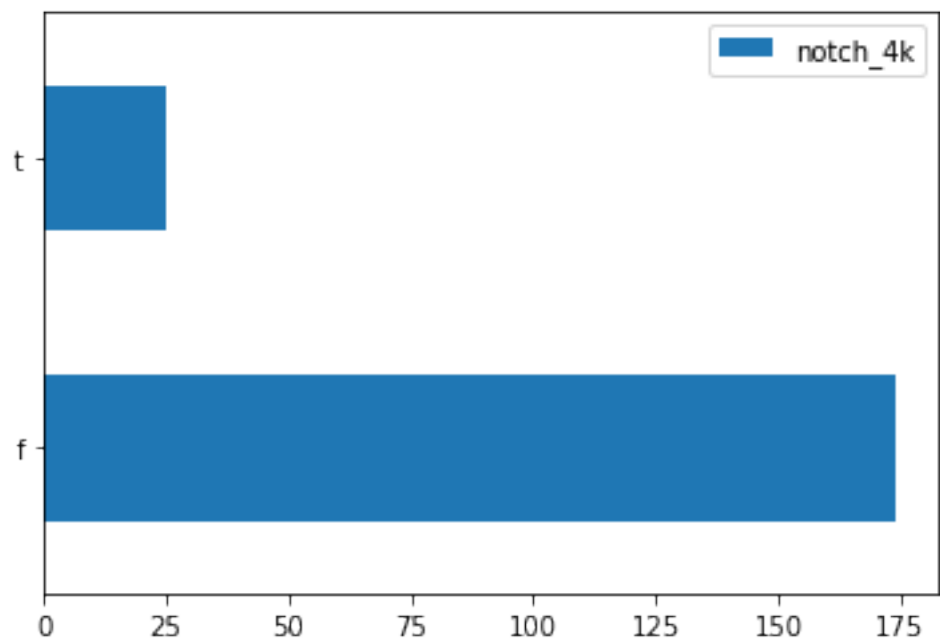


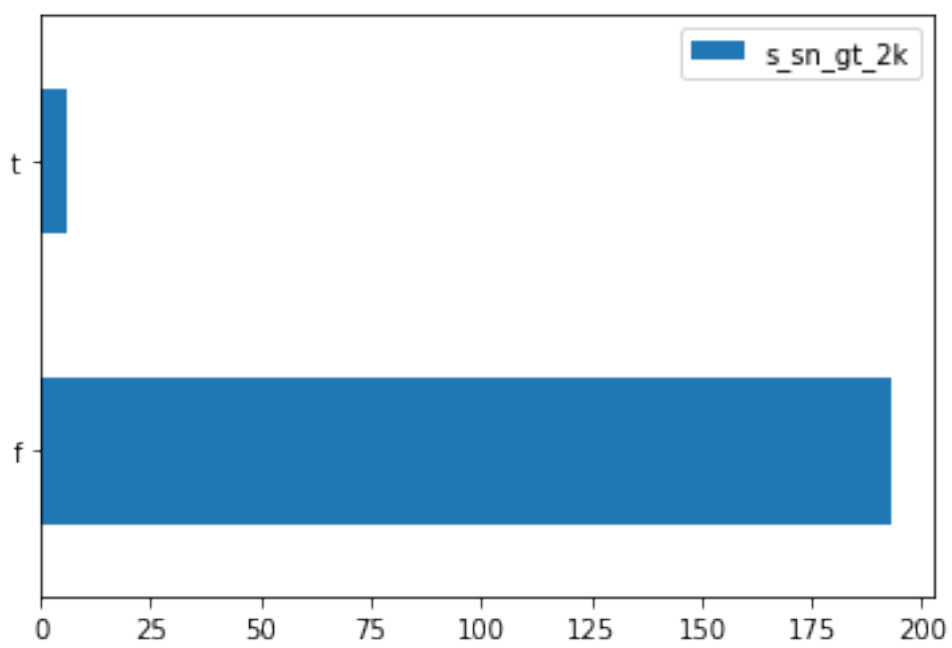
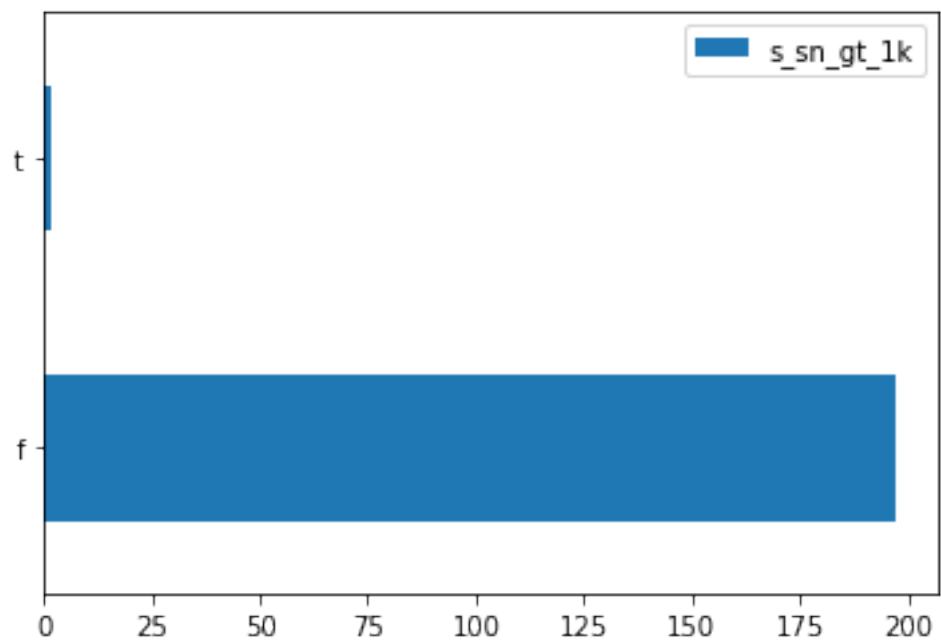


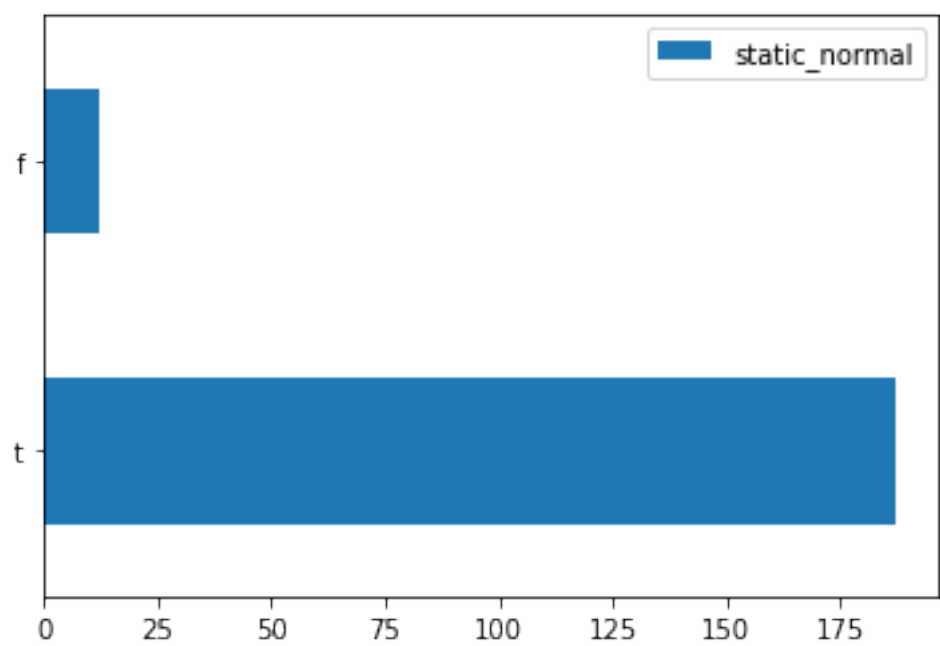
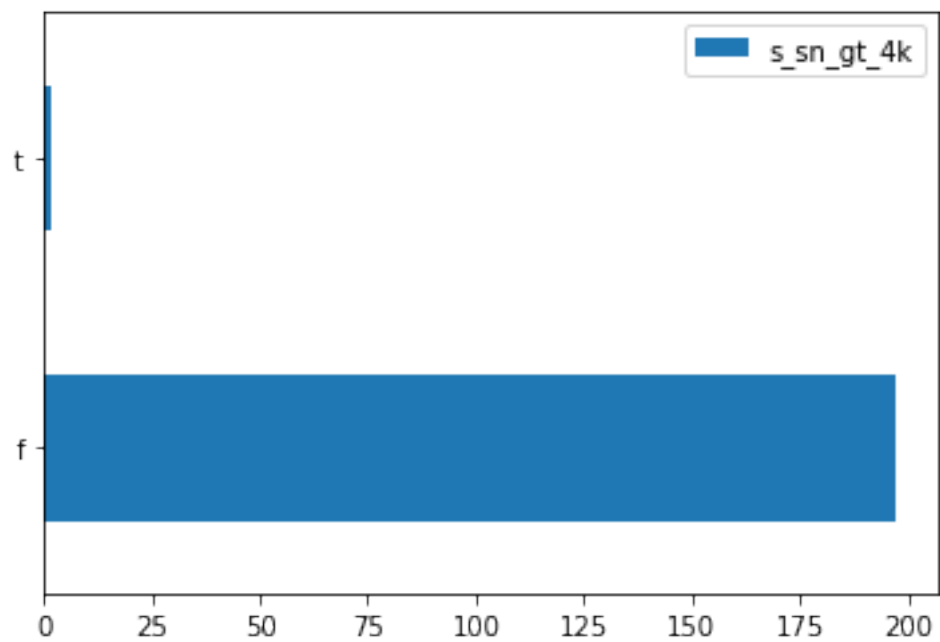


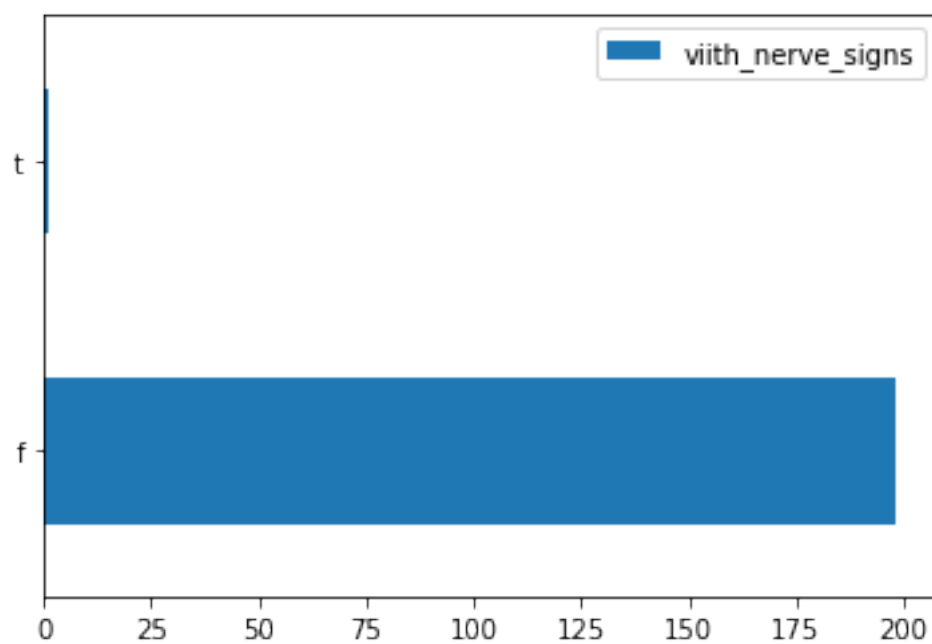
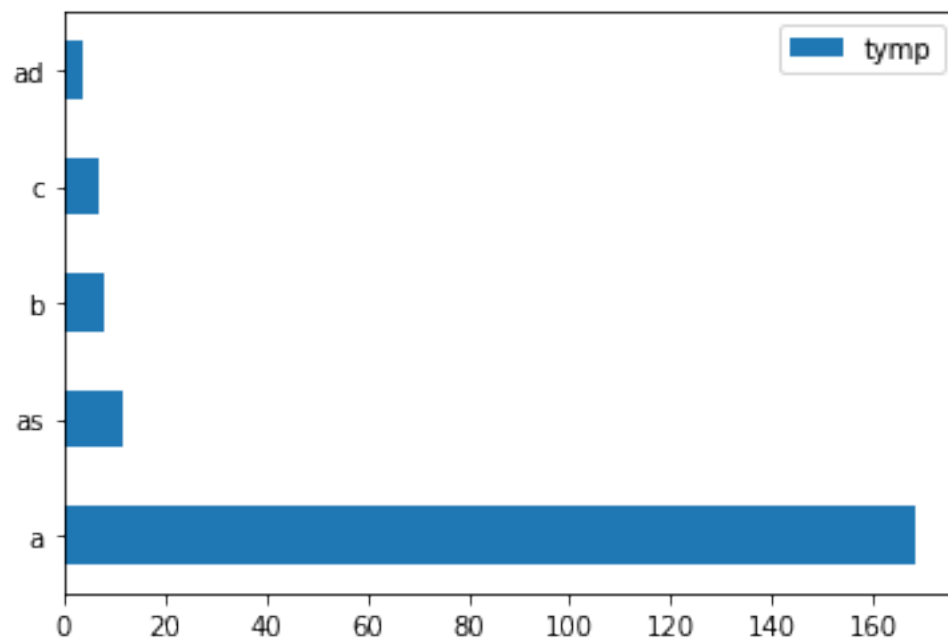


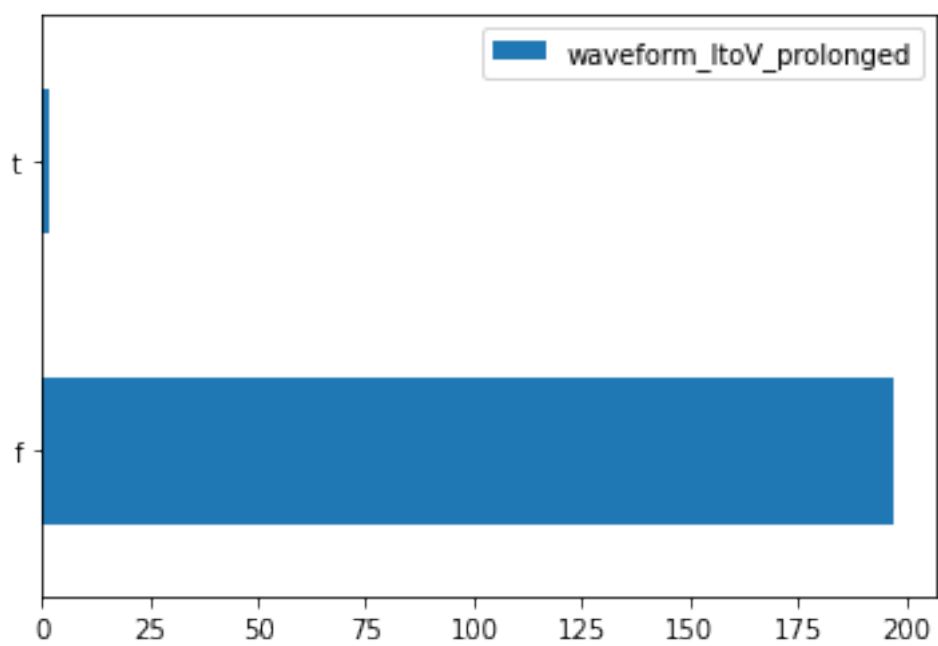
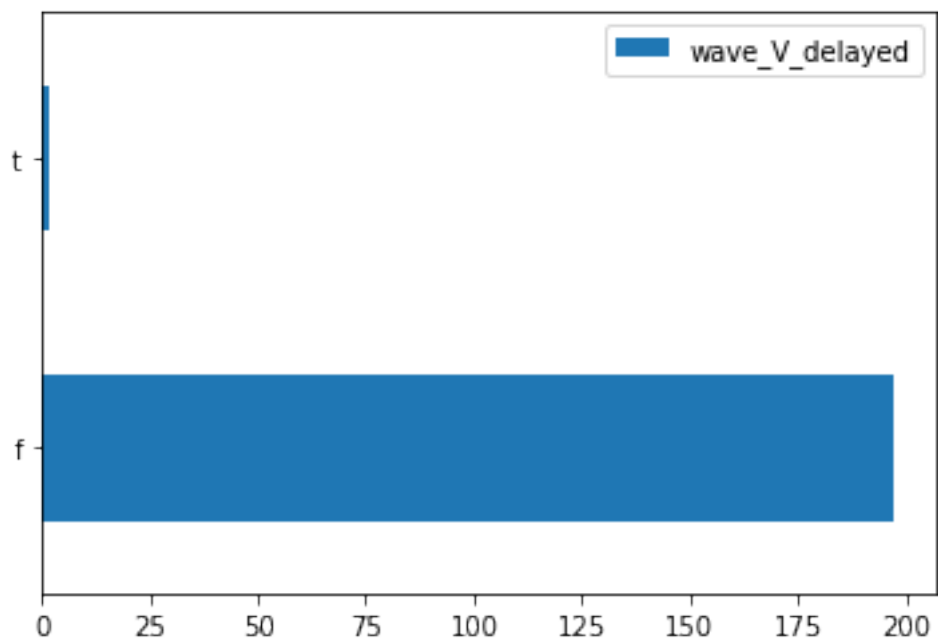


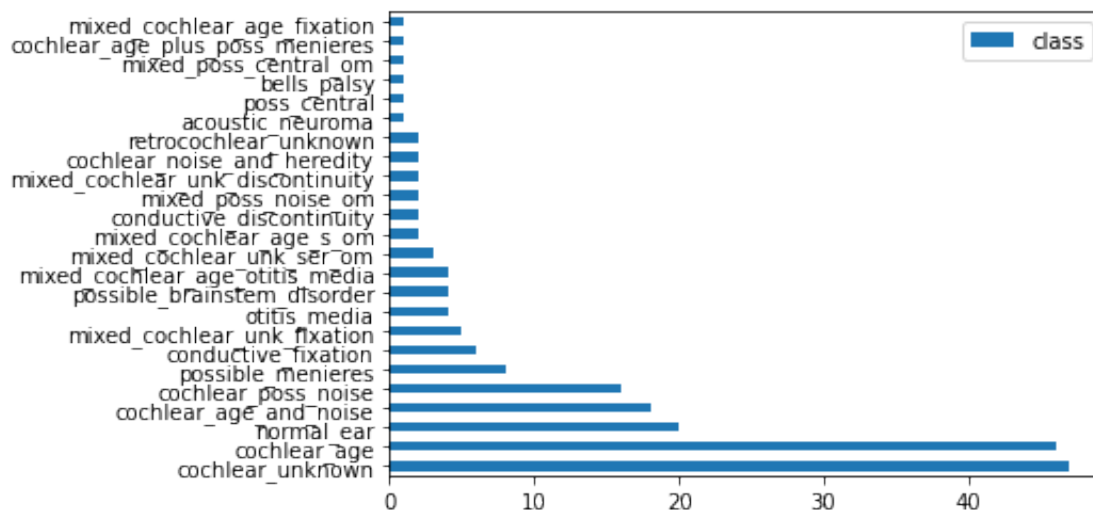
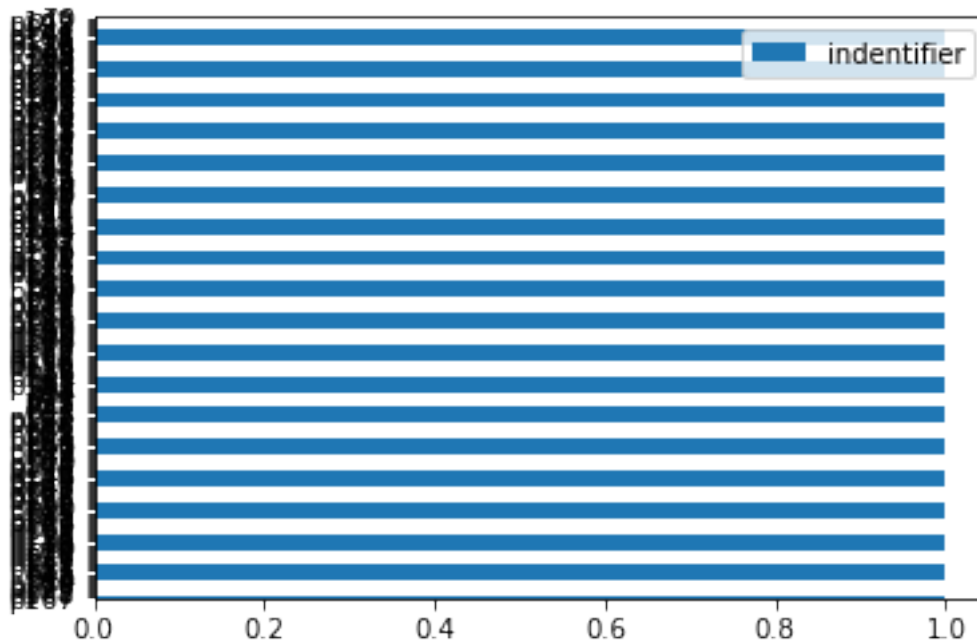












1.6 7. Draw class value counts

```
[11]: value_count = pd.DataFrame(data_df['class'].value_counts())
      value_count.plot.barh(figsize=(11,11))
```

```
[11]: <matplotlib.axes._subplots.AxesSubplot at 0x7fd2f30bafd0>
```

