

overview

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Dark-Fiber Charakterisierung für entanglement polarisierter QKD

Übersicht und Aufbereitung von Messverfahren und Messdaten

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Datensatz: Erfurt - Sundhausen

1 Einleitung: Kenngrößen, Messgerät und Datengrundlage

1.1 Ressourcen und Literatur

- [Messdaten](#)
- [Messgerät](#)
- [Abbildung: Poincaré-Kugel](#)

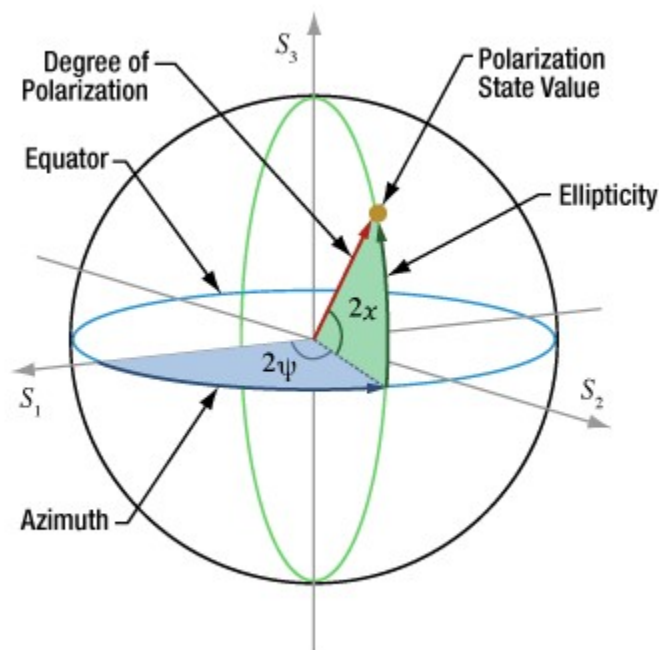
1.2 Kenngrößen: Bedeutung und Einheiten

Kenngröße	Einheit	Bedeutung	Wertebereiche
Time	date hh:mm:ss	absoluter Messzeitpunkt	-
Elapsed Time	hh:mm:ss:ms	relativer Messzeitpunkt	-
Normalized s1, s2, s3	-	Normierte Stokes-Komponenten	-1 x 1
S0	mW	Gesamtleistung	-
S1, S2, S3	mW	Stokes-Komponenten	-
Azimuth	°	Winkel in der Poincaré-Kugel (horizontaler Winkel)	-90° x 90°
Ellipticity	°	Maß für elliptische Polarisation (vertikaler Winkel)	-45° x 45°

KenngöÙe	Einheit	Bedeutung	Wertebereiche
DOP	%	Degree of Polarization	0% x 100%
DOCP	%	Degree of Circular Polarization	0% x 100%
DOLP	%	(Zirkularitätsmaß) Degree of Linear Polarization	0% x 100%
Power	mW / dBm	(Linearitätsmaß) Gesamtleistung	-60 dBm x +10 dBm
Pol Power	mW / dBm	Polarisierter Anteil (= DOP × Gesamtleistung)	-60 dBm x +10 dBm
Unpol Power	mW / dBm	Unpolarisierter Anteil (= (1-DOP) × Gesamtleistung)	-60 dBm x +10 dBm
Phase Difference	°	Phasenunterschied zw. Polarisationsmoden	-180° x 180°
Power-Split-Ratio	-	Leistungsverhältnis zweier Polarisationsrichtungen	0 x 1
Warning	-	Hinweis für fehlerhafte Messung	-

```
[1]: from IPython.display import Image
Image(filename="./img/poincare.jpg")
```

[1]:



1.3 Messgerät und Messverfahren

Kenngöße	Info
Device	PAX1000IR2/M
Serial Number	M00773008
Firmware Version	1.2.1
Software Version	1.4.2002.183
Wavelength [nm]	1560
Basic Sample Rate [Hz]	60
Operating Mode	1 revolution for one measurement, 1024 points for FFT

2 Aufbereitung der Messdaten

2.1 Verwendete Libraries / Softwaretools

```
[2]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import io
import datetime
import requests

from prettytable import PrettyTable
```

2.2 Eigene Hilfsfunktionen

```
[3]: def fast_parse_elapsed(s):
    try:
        d, h, m, s, ms = s.replace('.', ':').split(':')
        return pd.Timedelta(days=int(d), hours=int(h), minutes=int(m),
↪seconds=int(s), milliseconds=int(ms))
    except:
        return pd.NaT
```

2.3 Datenaufbereitungsschritte

```
[4]: filename = '20_02_to_26_02_Sundhausen to FZE port 2.csv'
skip = 8
sep = ";"
```

```
[5]:
```

```

columns = ['Time[date hh:mm:ss] ', ' Elapsed Time [hh:mm:ss:ms]', ' Normalized_
↪s 1 ', ' Normalized s 2 ', ' Normalized s 3 ', ' S 0 [mW]', ' S 1 [mW]', ' S_
↪2 [mW]', ' S 3 [mW]', ' Azimuth[°] ', ' Ellipticity[°] ', ' DOP[%] ', '_
↪DOCP[%] ', ' DOLP[%] ', ' Power[mW] ', ' Pol Power[mW] ', ' Unpol Power[mW]_
↪', ' Power[dBm] ', ' Pol Power[dBm] ', ' Unpol Power[dBm] ', '_
↪Power-Split-Ratio ', ' Phase Difference[°] ', ' Warning']
for c in range(len(columns)):
    print(c, ': ', columns[c])

```

```

0 : Time[date hh:mm:ss]
1 : Elapsed Time [hh:mm:ss:ms]
2 : Normalized s 1
3 : Normalized s 2
4 : Normalized s 3
5 : S 0 [mW]
6 : S 1 [mW]
7 : S 2 [mW]
8 : S 3 [mW]
9 : Azimuth[°]
10 : Ellipticity[°]
11 : DOP[%]
12 : DOCP[%]
13 : DOLP[%]
14 : Power[mW]
15 : Pol Power[mW]
16 : Unpol Power[mW]
17 : Power[dBm]
18 : Pol Power[dBm]
19 : Unpol Power[dBm]
20 : Power-Split-Ratio
21 : Phase Difference[°]
22 : Warning

```

3 Analyse der Messdaten

3.1 Fehler (Warning)

```
[6]: warning = pd.read_csv(filename, skiprows=skip, usecols=[columns[22]], sep=sep)
```

```

C:\Users\laura\AppData\Local\Temp\ipykernel_8048\1864455350.py:1: DtypeWarning:
Columns (22) have mixed types. Specify dtype option on import or set
low_memory=False.

```

```
    warning = pd.read_csv(filename, skiprows=skip, usecols=[columns[22]], sep=sep)
```

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

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2369654 entries, 0 to 2369653

```

Data columns (total 1 columns):

#	Column	Dtype
0	Warning	object

dtypes: object(1)
memory usage: 18.1+ MB

```
[8]: print(f'Null: {warning[columns[22]].isnull().sum()}\nNot-Null:{len(warning)-  
      ↳warning[columns[22]].isnull().sum()}')
```

Null: 2368270
Not-Null:1384

```
[9]: warning.dropna(subset = [columns[22]], inplace=True)
```

```
[10]: warning_groups = list(warning.groupby(columns[22]))  
  
print(f'Number of different Warnings: {len(warning_groups)}')  
print(f'Warning-Groups:')  
for i in range(len(warning_groups)):  
    print(f'\t- {warning_groups[i][0]} ({len(warning_groups[i][1])})')
```

Number of different Warnings: 1
Warning-Groups:
 - ADC overflow during last measurement. (1384)

3.2 Zeitwerte (Time, Elapsed Time)

```
[11]: time = pd.read_csv(filename,skiprows=skip, usecols=[columns[0], columns[1]],  
      ↳sep=sep)
```

```
[12]: time.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 2369654 entries, 0 to 2369653  
Data columns (total 2 columns):  
#   Column                                Dtype  
---  -----                                -  
0   Time[date hh:mm:ss]                  object  
1   Elapsed Time [hh:mm:ss:ms]           object  
dtypes: object(2)  
memory usage: 36.2+ MB
```

```
[13]: time.isnull().sum()
```

```
[13]: Time[date hh:mm:ss]                0  
      Elapsed Time [hh:mm:ss:ms]        0  
      dtype: int64
```

```
[14]: time[columns[0]] = pd.to_datetime(time[columns[0]])
time['time_difference'] = time[columns[0]].diff()
time['time_frequency'] = 1 / time['time_difference'].dt.total_seconds()
time['time_difference']
```

```
[14]: 0                               NaT
1      0 days 00:00:00.250000
2      0 days 00:00:00.250000
3      0 days 00:00:00.240000
4      0 days 00:00:00.250000
...
2369649 0 days 00:00:00.250000
2369650 0 days 00:00:00.250000
2369651 0 days 00:00:00.250000
2369652 0 days 00:00:00.240000
2369653 0 days 00:00:00.250000
Name: time_difference, Length: 2369654, dtype: timedelta64[ns]
```

```
[15]: time[columns[1]] = time[columns[1]].apply(fast_parse_elapsed)
time['elapsed_time_difference'] = time[columns[1]].diff()
time['elapsed_time_frequency'] = 1 / time['elapsed_time_difference'].dt.
    ↳total_seconds()
time['elapsed_time_difference']
```

```
[15]: 0                               NaT
1      0 days 00:00:00.253000
2      0 days 00:00:00.245000
3      0 days 00:00:00.249000
4      0 days 00:00:00.250000
...
2369649 0 days 00:00:00.246000
2369650 0 days 00:00:00.248000
2369651 0 days 00:00:00.249000
2369652 0 days 00:00:00.245000
2369653 0 days 00:00:00.248000
Name: elapsed_time_difference, Length: 2369654, dtype: timedelta64[ns]
```

```
[16]: t = PrettyTable([' ', 'Time', 'Elapsed Time'])
t.add_row(['Max', time['time_difference'].max(),
    ↳time['elapsed_time_difference'].max()])
t.add_row(['Min', time['time_difference'].min(),
    ↳time['elapsed_time_difference'].min()])
t.add_row(['Mean', time['time_difference'].mean(),
    ↳time['elapsed_time_difference'].mean()])
t.add_row(['Std', time['time_difference'].std(),
    ↳time['elapsed_time_difference'].std()])
```

```

t.add_row(['Most', time['time_difference'].value_counts(dropna=True).idxmax(),
↳time['elapsed_time_difference'].value_counts(dropna=True).idxmax()])
t.add_row(['Mean Frequency', time['time_frequency'].mean(),
↳time['elapsed_time_frequency'].mean()])

t

```

```

[16]: +-----+-----+-----+
|          |          Time          |          Elapsed Time          |
+-----+-----+-----+
|      Max      | 0 days 00:00:00.960000 | 0 days 00:00:00.958000 |
|      Min      | 0 days 00:00:00.240000 | 0 days 00:00:00.244000 |
|      Mean      | 0 days 00:00:00.255231293 | 0 days 00:00:00.255231293 |
|      Std       | 0 days 00:00:00.008582554 | 0 days 00:00:00.007129904 |
|      Most      | 0 days 00:00:00.260000 | 0 days 00:00:00.256000 |
| Mean Frequency |      3.9209356490794565 |      3.919556262116962 |
+-----+-----+-----+

```

```

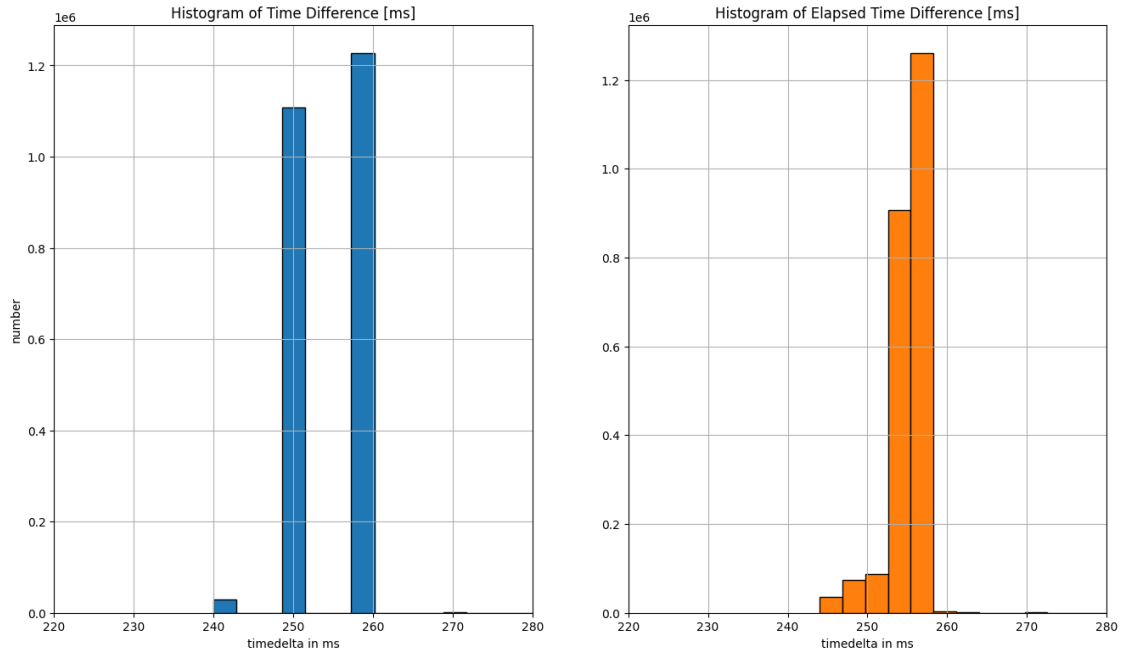
[17]: time_difference = time['time_difference'].dt.total_seconds() * 1000
elapsed_time_difference = time['elapsed_time_difference'].dt.total_seconds() *
↳1000

plt.figure(figsize = (16,9))
plt.subplot(1, 2, 1)
plt.hist(time_difference, bins=250, color='tab:blue', edgecolor='black')
plt.xlim([220, 280])
plt.grid()
plt.xlabel('timedelta in ms')
plt.ylabel('number')
plt.title('Histogram of Time Difference [ms]')

plt.subplot(1, 2, 2)
plt.hist(elapsed_time_difference, bins=250, color='tab:orange',
↳edgecolor='black')
plt.xlim([220, 280])
plt.grid()
plt.xlabel('timedelta in ms')
plt.title('Histogram of Elapsed Time Difference [ms]')

plt.show()

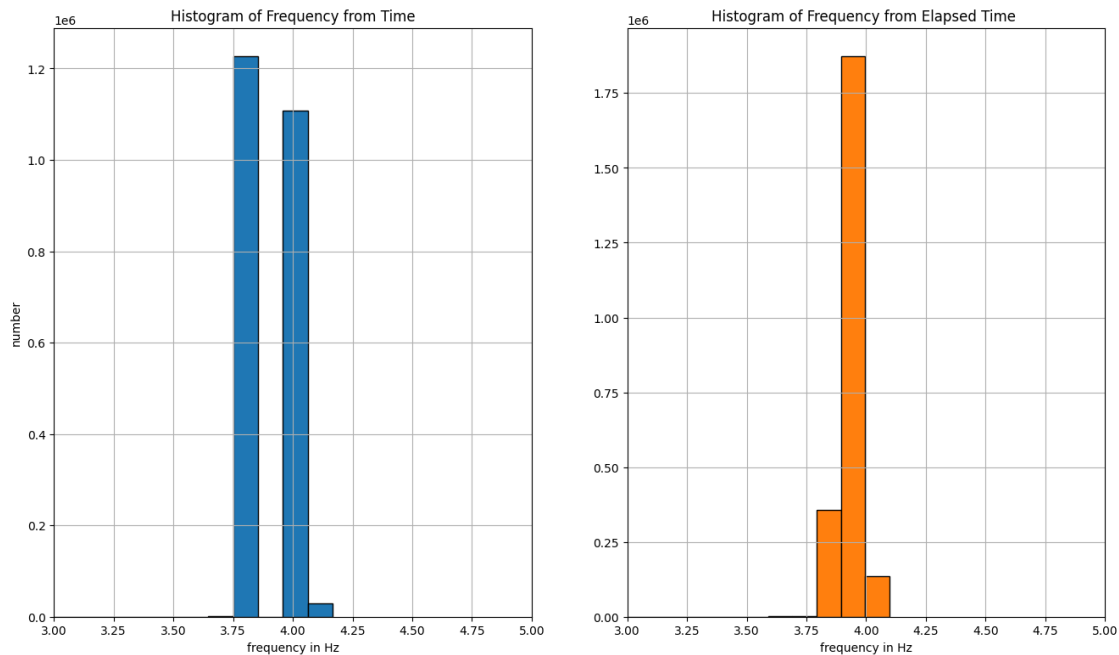
```



```
[18]: plt.figure(figsize = (16,9))
plt.subplot(1, 2, 1)
plt.hist(time['time_frequency'], bins=30, color='tab:blue', edgecolor='black')
plt.xlim([3, 5])
plt.grid()
plt.xlabel('frequency in Hz')
plt.ylabel('number')
plt.title('Histogram of Frequency from Time')

plt.subplot(1, 2, 2)
plt.hist(time['elapsed_time_frequency'], bins=30, color='tab:orange',
        edgecolor='black')
plt.xlim([3, 5])
plt.grid()
plt.xlabel('frequency in Hz')
plt.title('Histogram of Frequency from Elapsed Time ')

plt.show()
```

3.3 Polarisationsgrade (DOP, DOCP, DOLP)

```
[19]: degree = pd.read_csv(filename, skiprows=skip, usecols=[columns[0], columns[11],  
↪ columns[12], columns[13]], sep=sep)
```

```
[20]: degree[columns[0]] = pd.to_datetime(degree[columns[0]])  
degree.set_index(columns[0], inplace=True)
```

```
[21]: degree.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
DatetimeIndex: 2369654 entries, 2025-02-20 13:28:29.460000 to 2025-02-27  
13:28:39.060000  
Data columns (total 3 columns):  
#   Column      Dtype  
---  ---  
0    DOP[%]    float64  
1    DOCP[%]   float64  
2    DOLP[%]   float64  
dtypes: float64(3)  
memory usage: 72.3 MB
```

```
[22]: degree.isnull().sum()
```

```
[22]: DOP[%]      0  
      DOCP[%]   0
```

```
DOLP[%]      0
dtype: int64
```

```
[23]: degree.describe()
```

```
[23]:
```

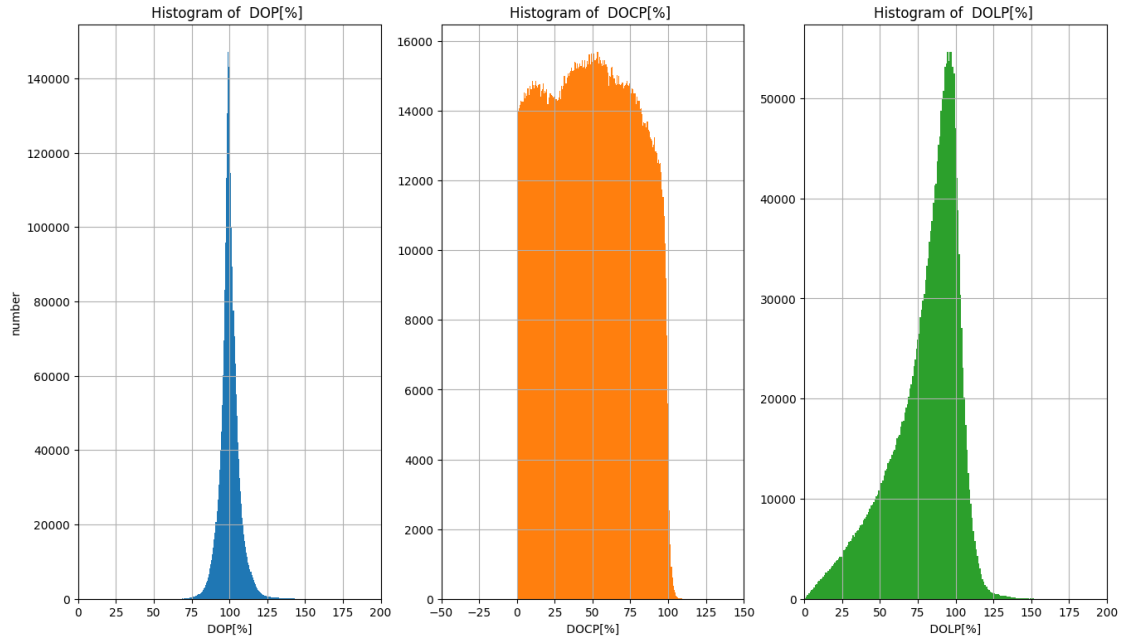
	DOP [%]	DOCP [%]	DOLP [%]
count	2.369654e+06	2.369654e+06	2.369654e+06
mean	1.000893e+02	4.901668e+01	7.972596e+01
std	6.368664e+00	2.815734e+01	2.251058e+01
min	5.856000e+01	0.000000e+00	5.000000e-02
25%	9.694000e+01	2.490000e+01	6.766000e+01
50%	9.974000e+01	4.894000e+01	8.564000e+01
75%	1.030100e+02	7.280000e+01	9.610000e+01
max	1.737000e+02	1.219400e+02	1.714200e+02

```
[24]: plt.figure(figsize = (16,9))
plt.subplot(1, 3, 1)
plt.hist(degree[columns[11]], bins=200, color='tab:blue')
plt.xlim([0, 200])
plt.grid()
plt.xlabel(columns[11])
plt.ylabel('number')
plt.title(f'Histogram of {columns[11]}')

plt.subplot(1, 3, 2)
plt.hist(degree[columns[12]], bins=200, color='tab:orange')
plt.xlim([-50, 150])
plt.grid()
plt.xlabel(columns[12])
plt.title(f'Histogram of {columns[12]}')

plt.subplot(1, 3, 3)
plt.hist(degree[columns[13]], bins=200, color='tab:green')
plt.xlim([0, 200])
plt.grid()
plt.xlabel(columns[13])
plt.title(f'Histogram of {columns[13]}')

plt.show()
```

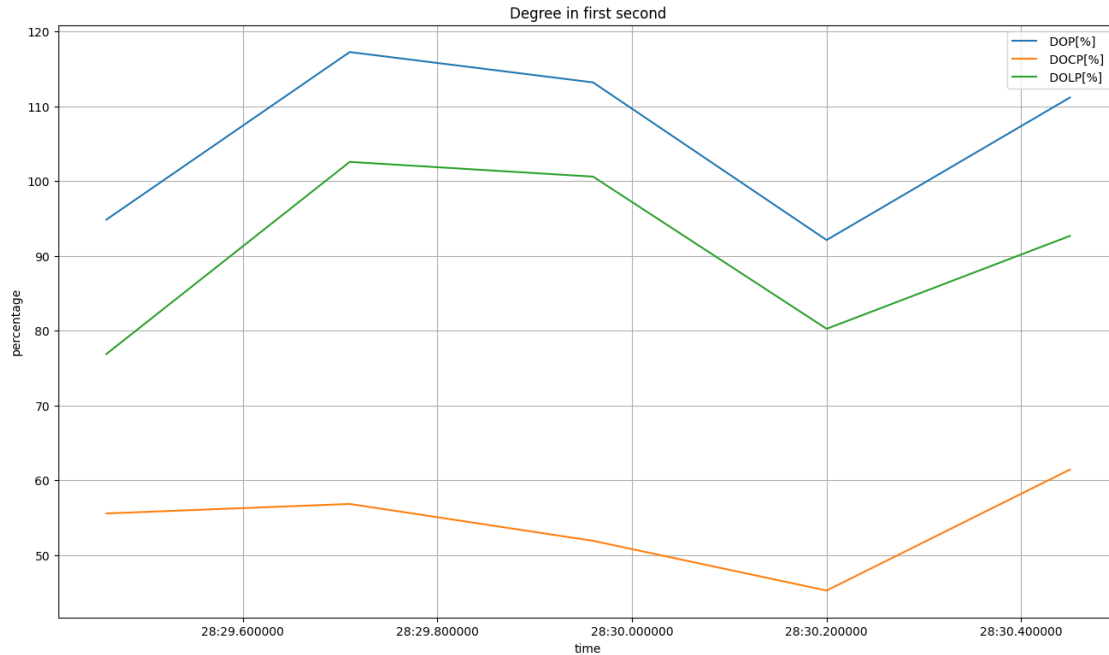


```
[25]: one_second = datetime.timedelta(seconds=1)
one_minute = datetime.timedelta(minutes=1)
one_hour = datetime.timedelta(hours=1)
one_day = datetime.timedelta(days=1)

start_time = datetime.datetime.strptime(str(degree.index[0]), '%Y-%m-%d %H:%M:
↪%S.%f')
end_time_second = start_time + one_second
end_time_minute = start_time + one_minute
end_time_hour = start_time + one_hour
end_time_day = start_time + one_day
```

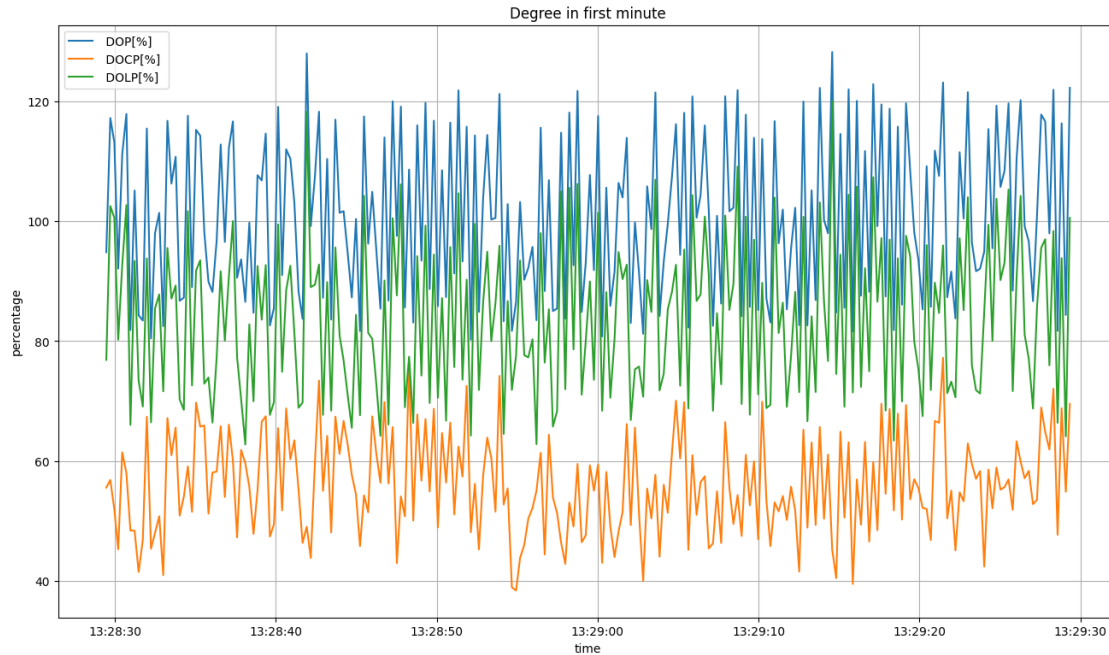
```
[26]: first_second = degree.loc[(degree.index >= start_time) & (degree.index <
↪end_time_second)]

plt.figure(figsize = (16,9))
plt.plot(first_second.index, first_second[columns[11]], label = columns[11])
plt.plot(first_second.index, first_second[columns[12]], label = columns[12])
plt.plot(first_second.index, first_second[columns[13]], label = columns[13])
plt.grid()
plt.legend(loc = 'best')
plt.title('Degree in first second')
plt.xlabel('time')
plt.ylabel('percentage')
plt.show()
```



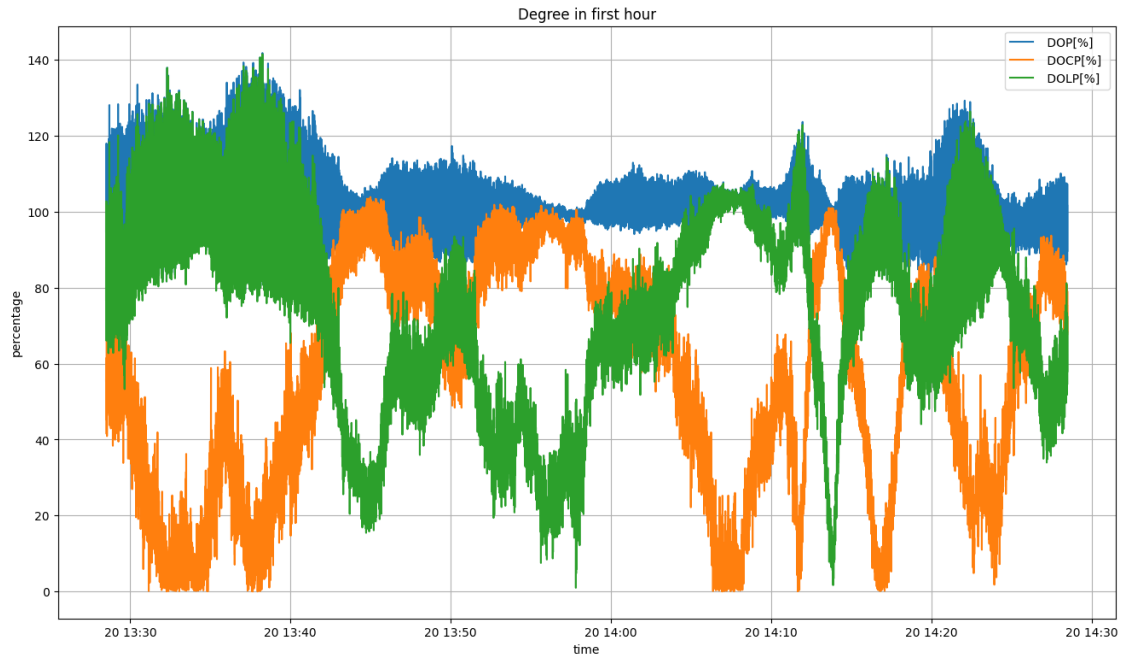
```
[27]: first_minute = degree.loc[(degree.index >= start_time) & (degree.index <
    ↪end_time_minute)]

plt.figure(figsize = (16,9))
plt.plot(first_minute.index, first_minute[columns[11]], label = columns[11])
plt.plot(first_minute.index, first_minute[columns[12]], label = columns[12])
plt.plot(first_minute.index, first_minute[columns[13]], label = columns[13])
plt.grid()
plt.legend(loc = 'best')
plt.title('Degree in first minute')
plt.xlabel('time')
plt.ylabel('percentage')
plt.show()
```



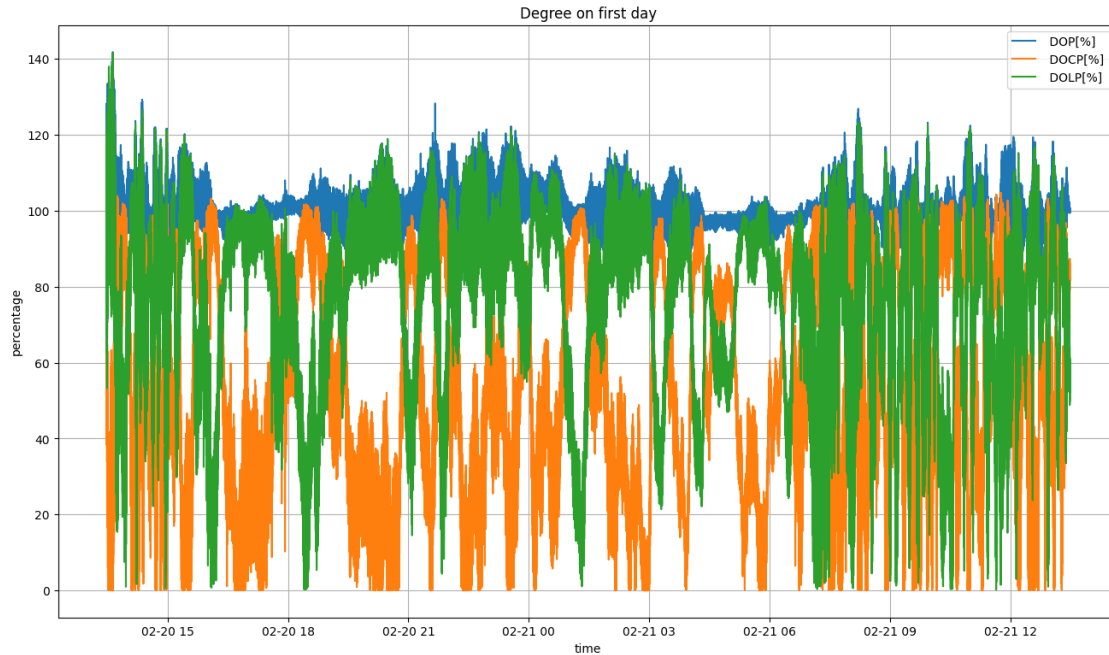
```
[28]: first_hour = degree.loc[(degree.index >= start_time) & (degree.index <
    ↪end_time_hour)]

plt.figure(figsize = (16,9))
plt.plot(first_hour.index, first_hour[columns[11]], label = columns[11])
plt.plot(first_hour.index, first_hour[columns[12]], label = columns[12])
plt.plot(first_hour.index, first_hour[columns[13]], label = columns[13])
plt.grid()
plt.legend(loc = 'best')
plt.title('Degree in first hour')
plt.xlabel('time')
plt.ylabel('percentage')
plt.show()
```



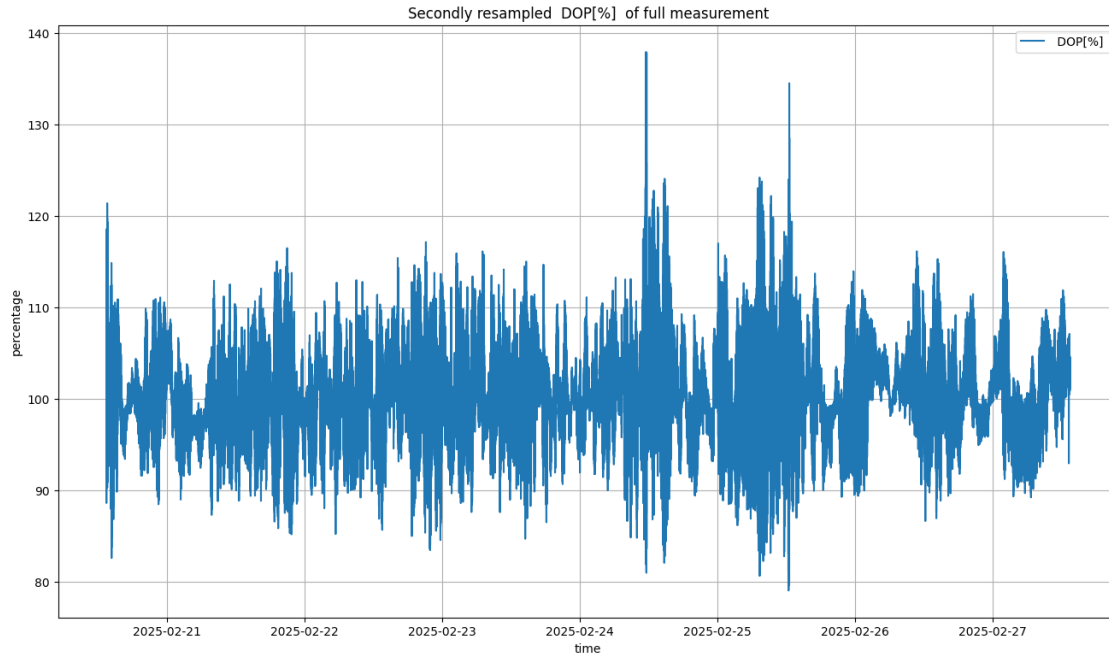
```
[29]: first_day = degree.loc[(degree.index >= start_time) & (degree.index <=
    ↪end_time_day)]

plt.figure(figsize = (16,9))
plt.plot(first_day.index, first_day[columns[11]], label = columns[11])
plt.plot(first_day.index, first_day[columns[12]], label = columns[12])
plt.plot(first_day.index, first_day[columns[13]], label = columns[13])
plt.grid()
plt.legend(loc = 'best')
plt.title('Degree on first day')
plt.xlabel('time')
plt.ylabel('percentage')
plt.show()
```

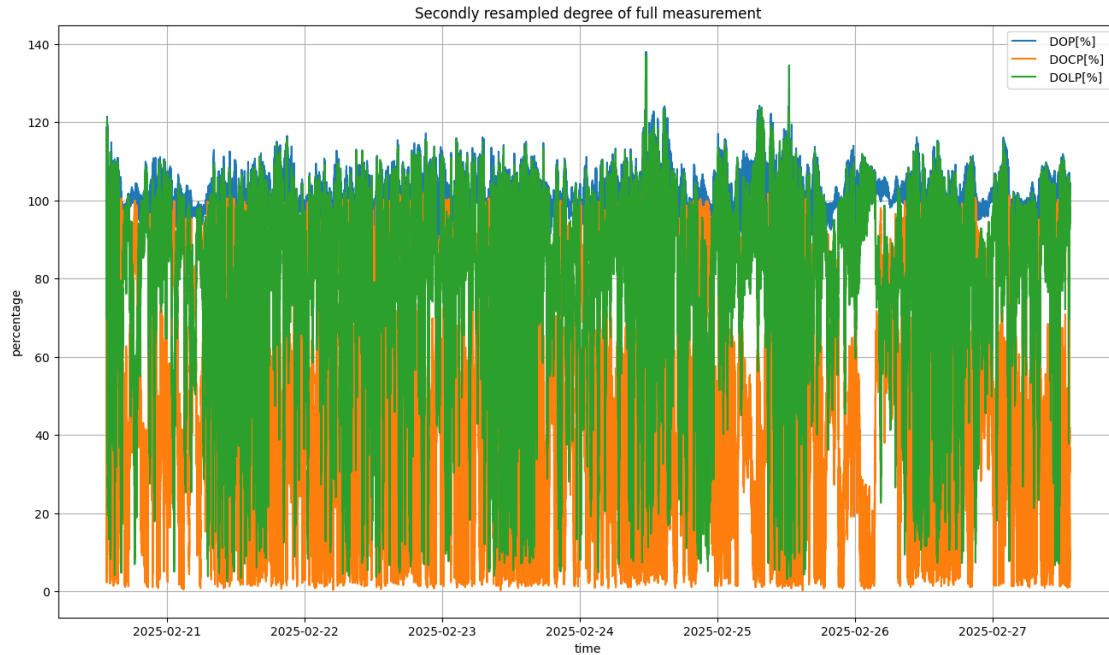


```
[30]: degree_seconds_resample = degree.resample('s').mean()
```

```
[31]: plt.figure(figsize = (16,9))
plt.plot(degree_seconds_resample.index, degree_seconds_resample[columns[11]],  
↪label = columns[11])
plt.grid()
plt.legend(loc = 'best')
plt.title(f'Secondly resampled {columns[11]} of full measurement')
plt.xlabel('time')
plt.ylabel('percentage')
plt.show()
```



```
[32]: plt.figure(figsize = (16,9))
plt.plot(degree_seconds_resample.index, degree_seconds_resample[columns[11]],
        ↪label = columns[11])
plt.plot(degree_seconds_resample.index, degree_seconds_resample[columns[12]],
        ↪label = columns[12])
plt.plot(degree_seconds_resample.index, degree_seconds_resample[columns[13]],
        ↪label = columns[13])
plt.grid()
plt.legend(loc = 'best')
plt.title('Secondly resampled degree of full measurement')
plt.xlabel('time')
plt.ylabel('percentage')
plt.show()
```

3.4 Energie (Power, Pol Power, Unpol Power [mW/dBm])

```
[33]: power_mw = pd.read_csv(filename, skiprows=skip, usecols=[columns[0], ↵
      ↪ columns[14], columns[15], columns[16]], sep=sep)
```

```
[34]: power_mw[columns[0]] = pd.to_datetime(power_mw[columns[0]])
      power_mw.set_index(columns[0], inplace=True)
```

```
[35]: power_mw.info()
```

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 2369654 entries, 2025-02-20 13:28:29.460000 to 2025-02-27
13:28:39.060000
Data columns (total 3 columns):
#   Column          Dtype
---  -
0   Power[mW]       float64
1   Pol Power[mW]   float64
2   Unpol Power[mW] float64
dtypes: float64(3)
memory usage: 72.3 MB
```

```
[36]: power_mw.isnull().sum()
```

```
[36]: Power[mW]          0
      Pol Power[mW]      0
```

```
Unpol Power[mW]      0
dtype: int64
```

```
[37]: power_mw.describe()
```

```
[37]:
```

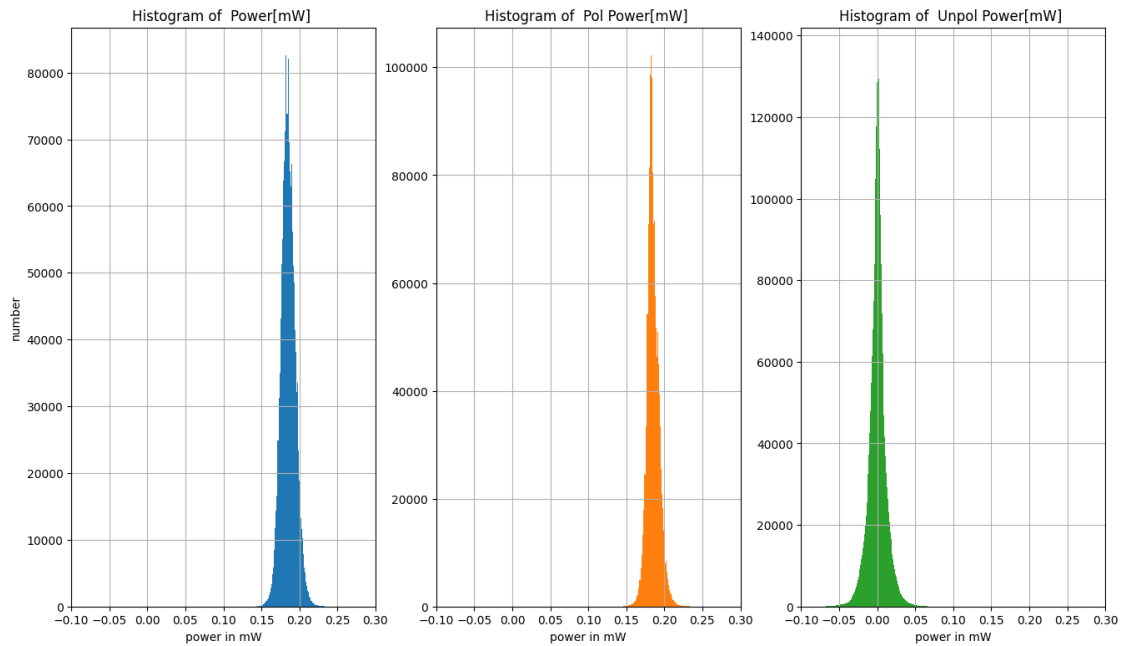
	Power[mW]	Pol Power[mW]	Unpol Power[mW]
count	2.369654e+06	2.369654e+06	2.369654e+06
mean	1.853396e-01	1.850686e-01	2.709176e-04
std	9.411575e-03	8.015791e-03	1.166701e-02
min	9.761000e-02	1.111000e-01	-1.006000e-01
25%	1.791000e-01	1.800000e-01	-5.437000e-03
50%	1.849000e-01	1.842000e-01	4.901000e-04
75%	1.914000e-01	1.898000e-01	5.802000e-03
max	2.415000e-01	2.417000e-01	9.792000e-02

```
[38]: plt.figure(figsize = (16,9))
plt.subplot(1, 3, 1)
plt.hist(power_mw[columns[14]], bins=200, color='tab:blue')
plt.xlim([-0.1, 0.3])
plt.grid()
plt.xlabel('power in mW')
plt.ylabel('number')
plt.title(f'Histogram of {columns[14]}')

plt.subplot(1, 3, 2)
plt.hist(power_mw[columns[15]], bins=200, color='tab:orange')
plt.xlim([-0.1, 0.3])
plt.grid()
plt.xlabel('power in mW')
plt.title(f'Histogram of {columns[15]}')

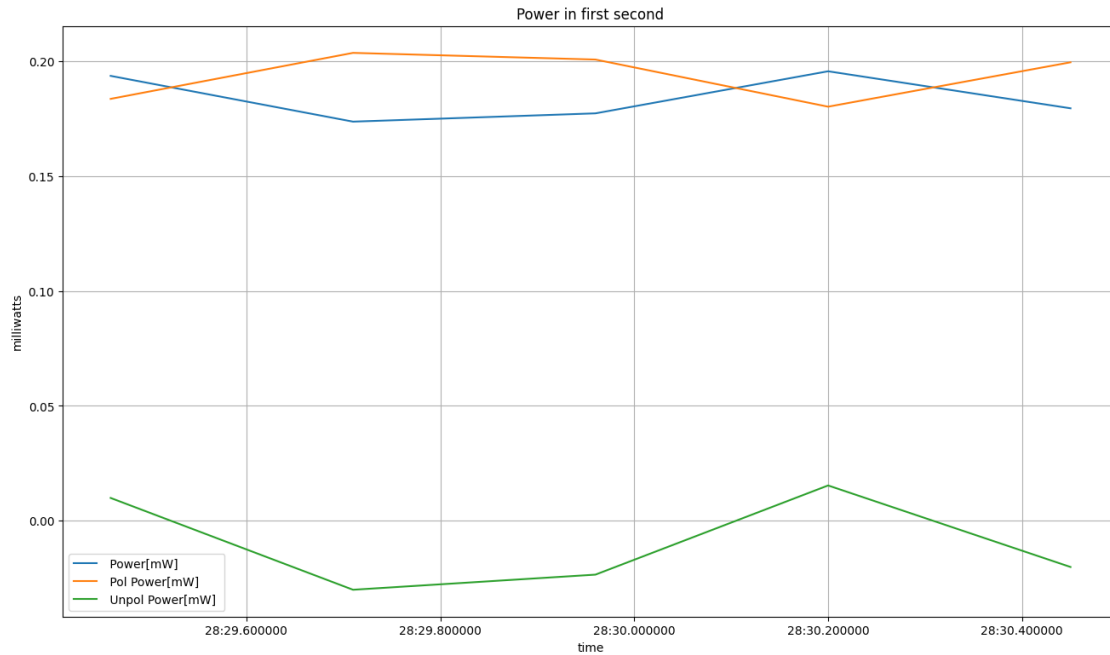
plt.subplot(1, 3, 3)
plt.hist(power_mw[columns[16]], bins=200, color='tab:green')
plt.xlim([-0.1, 0.3])
plt.grid()
plt.xlabel('power in mW')
plt.title(f'Histogram of {columns[16]}')

plt.show()
```



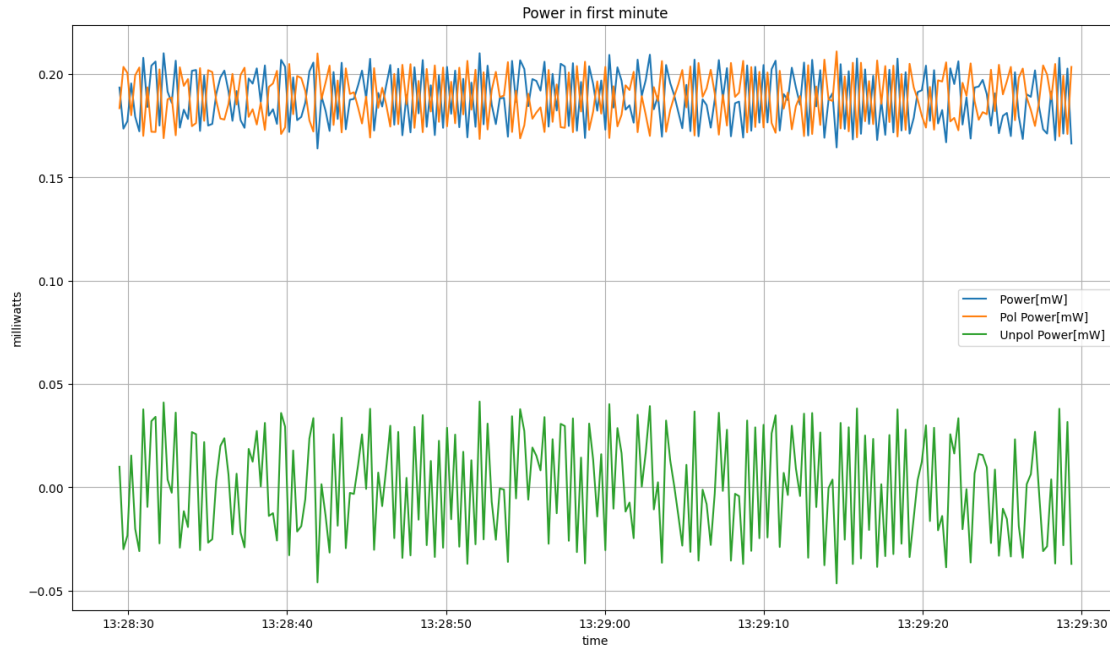
```
[39]: first_second = power_mw.loc[(power_mw.index >= start_time) & (power_mw.index <=
    ↪end_time_second)]

plt.figure(figsize = (16,9))
plt.plot(first_second.index, first_second[columns[14]], label = columns[14])
plt.plot(first_second.index, first_second[columns[15]], label = columns[15])
plt.plot(first_second.index, first_second[columns[16]], label = columns[16])
plt.grid()
plt.legend(loc = 'best')
plt.title('Power in first second')
plt.xlabel('time')
plt.ylabel('milliwatts')
plt.show()
```



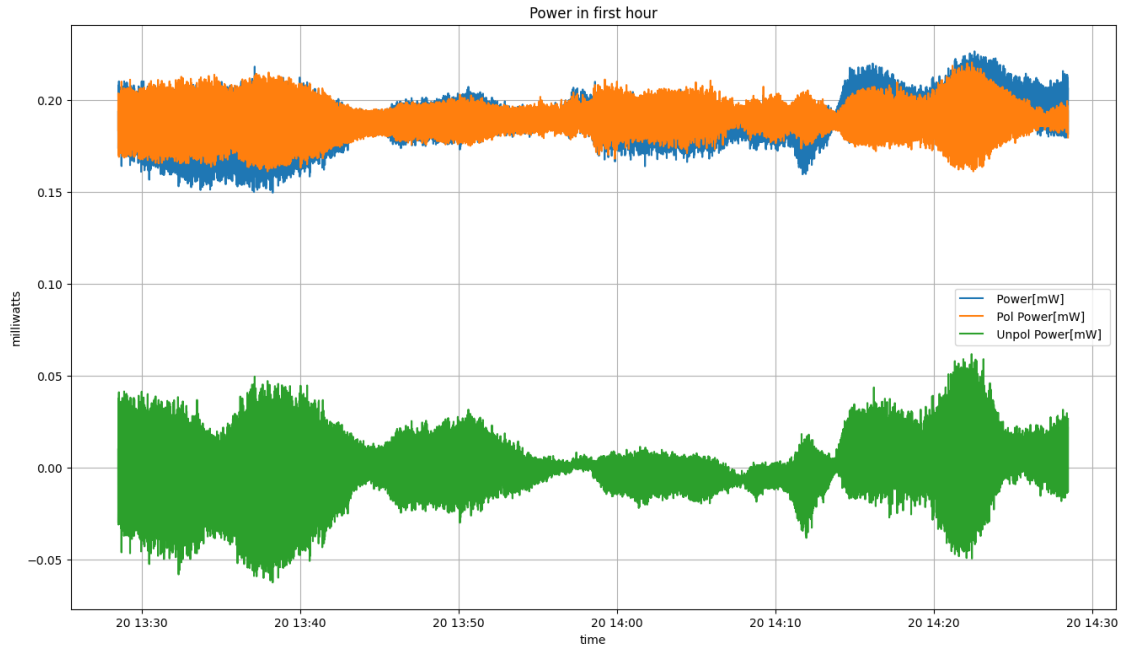
```
[40]: first_minute = power_mw.loc[(power_mw.index >= start_time) & (power_mw.index <
    ↪end_time_minute)]

plt.figure(figsize = (16,9))
plt.plot(first_minute.index, first_minute[columns[14]], label = columns[14])
plt.plot(first_minute.index, first_minute[columns[15]], label = columns[15])
plt.plot(first_minute.index, first_minute[columns[16]], label = columns[16])
plt.grid()
plt.legend(loc = 'best')
plt.title('Power in first minute')
plt.xlabel('time')
plt.ylabel('milliwatts')
plt.show()
```



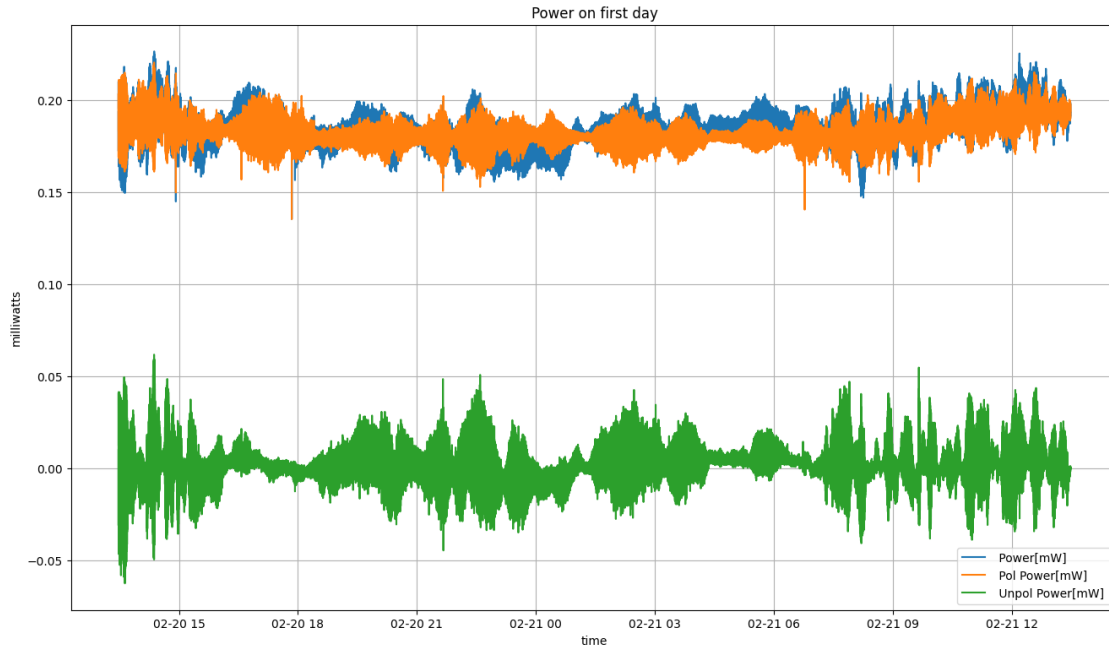
```
[41]: first_hour = power_mw.loc[(power_mw.index >= start_time) & (power_mw.index <
    ↪end_time_hour)]

plt.figure(figsize = (16,9))
plt.plot(first_hour.index, first_hour[columns[14]], label = columns[14])
plt.plot(first_hour.index, first_hour[columns[15]], label = columns[15])
plt.plot(first_hour.index, first_hour[columns[16]], label = columns[16])
plt.grid()
plt.legend(loc = 'best')
plt.title('Power in first hour')
plt.xlabel('time')
plt.ylabel('milliwatts')
plt.show()
```



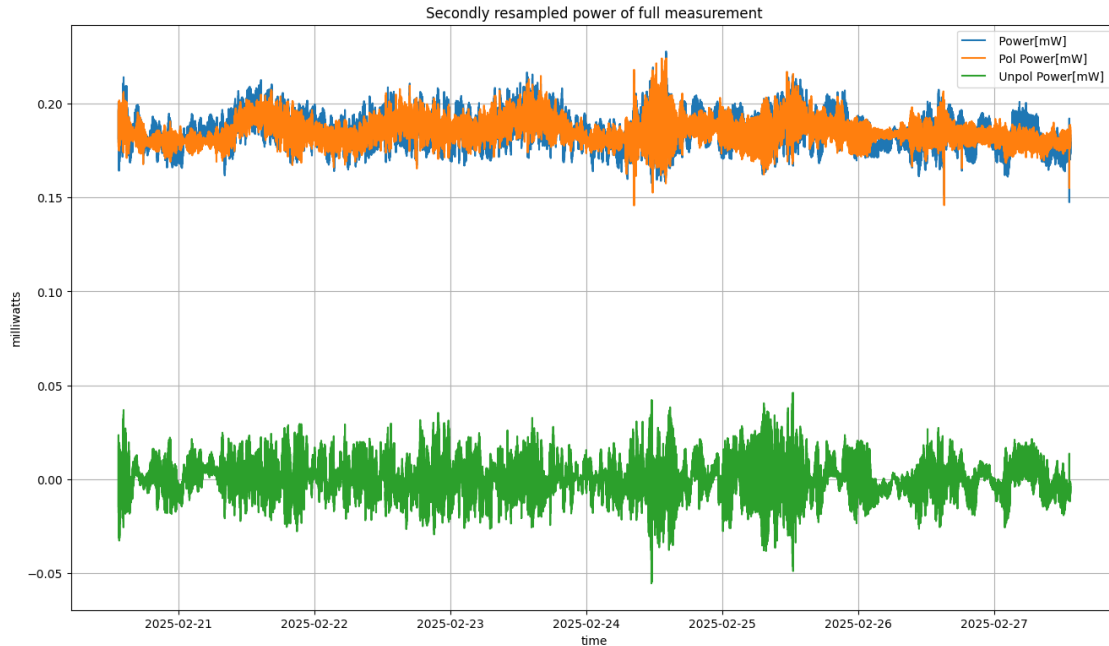
```
[42]: first_day = power_mw.loc[(power_mw.index >= start_time) & (power_mw.index <
    ↪end_time_day)]

plt.figure(figsize = (16,9))
plt.plot(first_day.index, first_day[columns[14]], label = columns[14])
plt.plot(first_day.index, first_day[columns[15]], label = columns[15])
plt.plot(first_day.index, first_day[columns[16]], label = columns[16])
plt.grid()
plt.legend(loc = 'best')
plt.title('Power on first day')
plt.xlabel('time')
plt.ylabel('milliwatts')
plt.show()
```



```
[43]: power_mw_seconds_resample = power_mw.resample('s').mean()
```

```
[44]: plt.figure(figsize = (16,9))
plt.plot(power_mw_seconds_resample.index,
         ↪power_mw_seconds_resample[columns[14]], label = columns[14])
plt.plot(power_mw_seconds_resample.index,
         ↪power_mw_seconds_resample[columns[15]], label = columns[15])
plt.plot(power_mw_seconds_resample.index,
         ↪power_mw_seconds_resample[columns[16]], label = columns[16])
plt.grid()
plt.legend(loc = 'best')
plt.title('Secondly resampled power of full measurement')
plt.xlabel('time')
plt.ylabel('milliwatts')
plt.show()
```



```
[45]: power_dbm = pd.read_csv(filename, skiprows=skip, usecols=[columns[0],  
↪ columns[17], columns[18], columns[19]], sep=sep)
```

```
[46]: power_dbm[columns[0]] = pd.to_datetime(power_dbm[columns[0]])  
power_dbm.set_index(columns[0], inplace=True)
```

```
[47]: power_dbm.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
DatetimeIndex: 2369654 entries, 2025-02-20 13:28:29.460000 to 2025-02-27  
13:28:39.060000  
Data columns (total 3 columns):  
#   Column          Dtype  
---  ---  
0   Power[dBm]      float64  
1   Pol Power[dBm]  float64  
2   Unpol Power[dBm] float64  
dtypes: float64(3)  
memory usage: 72.3 MB
```

```
[48]: power_dbm.isnull().sum()
```

```
[48]: Power[dBm]          0  
Pol Power[dBm]          0  
Unpol Power[dBm]        0  
dtype: int64
```



```
[49]: power_dbm.describe()
```

```
[49]:
```

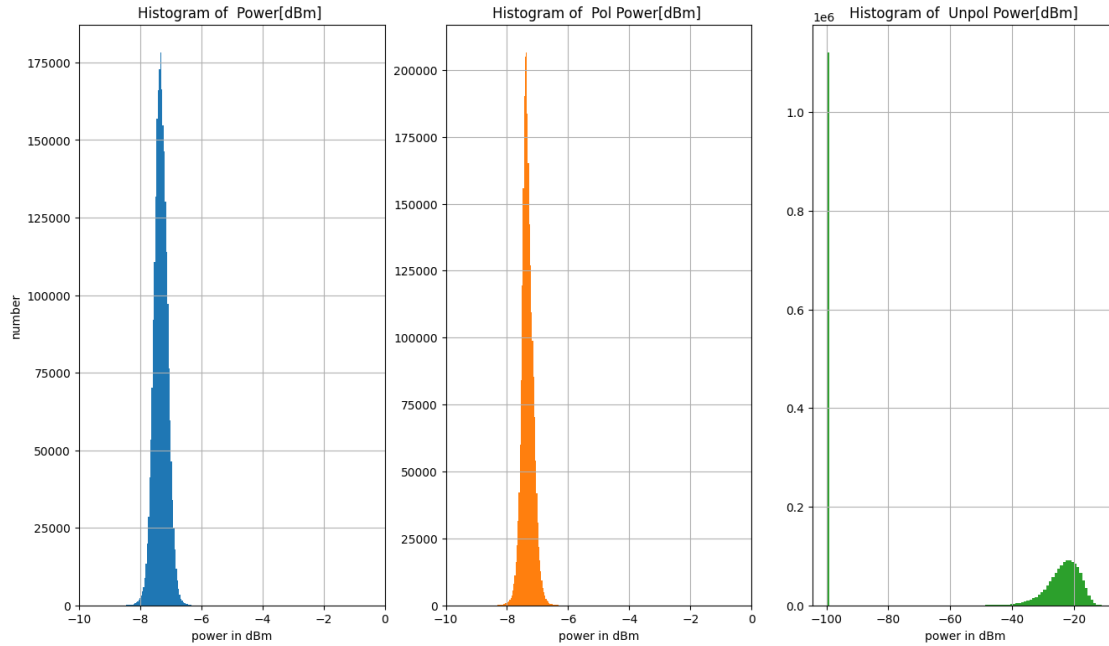
	Power[dBm]	Pol Power[dBm]	Unpol Power[dBm]
count	2.369654e+06	2.369654e+06	2.369654e+06
mean	-7.325910e+00	-7.330714e+00	-5.963221e+01
std	2.203316e-01	1.870446e-01	3.844336e+01
min	-1.010500e+01	-9.542000e+00	-9.999000e+01
25%	-7.469000e+00	-7.447000e+00	-9.999000e+01
50%	-7.330000e+00	-7.347000e+00	-3.309700e+01
75%	-7.181000e+00	-7.218000e+00	-2.236400e+01
max	-6.171000e+00	-6.168000e+00	-1.009200e+01

```
[50]: plt.figure(figsize = (16,9))
plt.subplot(1, 3, 1)
plt.hist(power_dbm[columns[17]], bins=100, color='tab:blue')
plt.xlim([-10, 0])
plt.grid()
plt.xlabel('power in dBm')
plt.ylabel('number')
plt.title(f'Histogram of {columns[17]}')

plt.subplot(1, 3, 2)
plt.hist(power_dbm[columns[18]], bins=100, color='tab:orange')
plt.xlim([-10, 0])
plt.grid()
plt.xlabel('power in dBm')
plt.title(f'Histogram of {columns[18]}')

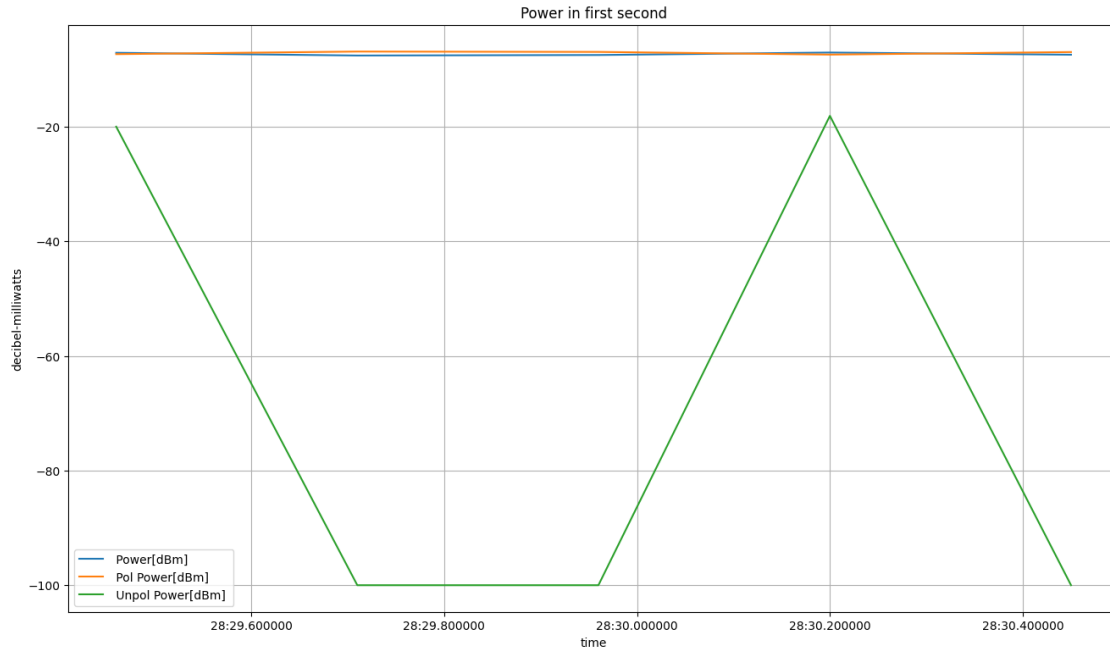
plt.subplot(1, 3, 3)
plt.hist(power_dbm[columns[19]], bins=100, color='tab:green')
plt.grid()
plt.xlabel('power in dBm')
plt.title(f'Histogram of {columns[19]}')

plt.show()
```



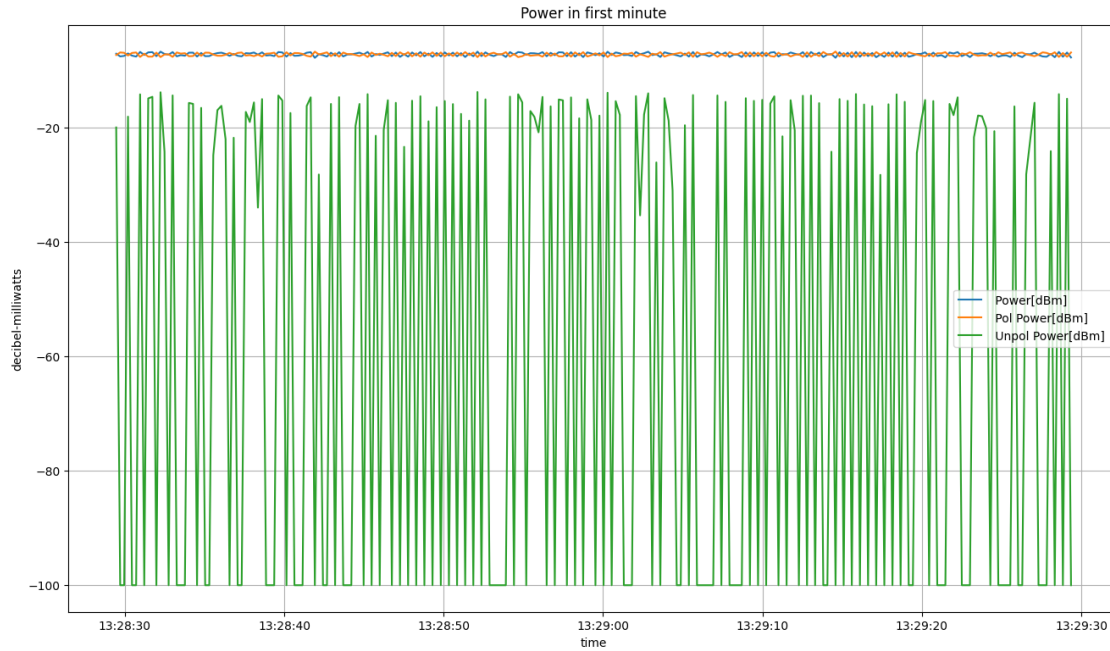
```
[51]: first_second = power_dbm.loc[(power_dbm.index >= start_time) & (power_dbm.index
    ↪ < end_time_second)]

plt.figure(figsize = (16,9))
plt.plot(first_second.index, first_second[columns[17]], label = columns[17])
plt.plot(first_second.index, first_second[columns[18]], label = columns[18])
plt.plot(first_second.index, first_second[columns[19]], label = columns[19])
plt.grid()
plt.legend(loc = 'best')
plt.title('Power in first second')
plt.xlabel('time')
plt.ylabel('decibel-milliwatts')
plt.show()
```



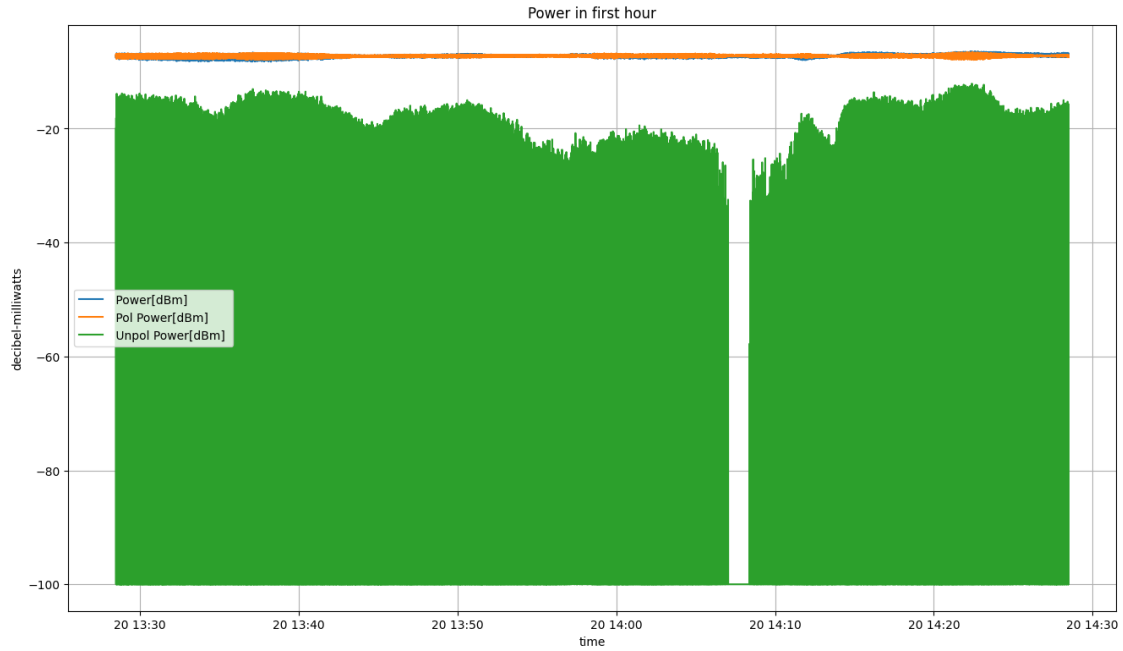
```
[52]: first_minute = power_dbm.loc[(power_dbm.index >= start_time) & (power_dbm.index
    ↪ < end_time_minute)]

plt.figure(figsize = (16,9))
plt.plot(first_minute.index, first_minute[columns[17]], label = columns[17])
plt.plot(first_minute.index, first_minute[columns[18]], label = columns[18])
plt.plot(first_minute.index, first_minute[columns[19]], label = columns[19])
plt.grid()
plt.legend(loc = 'best')
plt.title('Power in first minute')
plt.xlabel('time')
plt.ylabel('decibel-milliwatts')
plt.show()
```



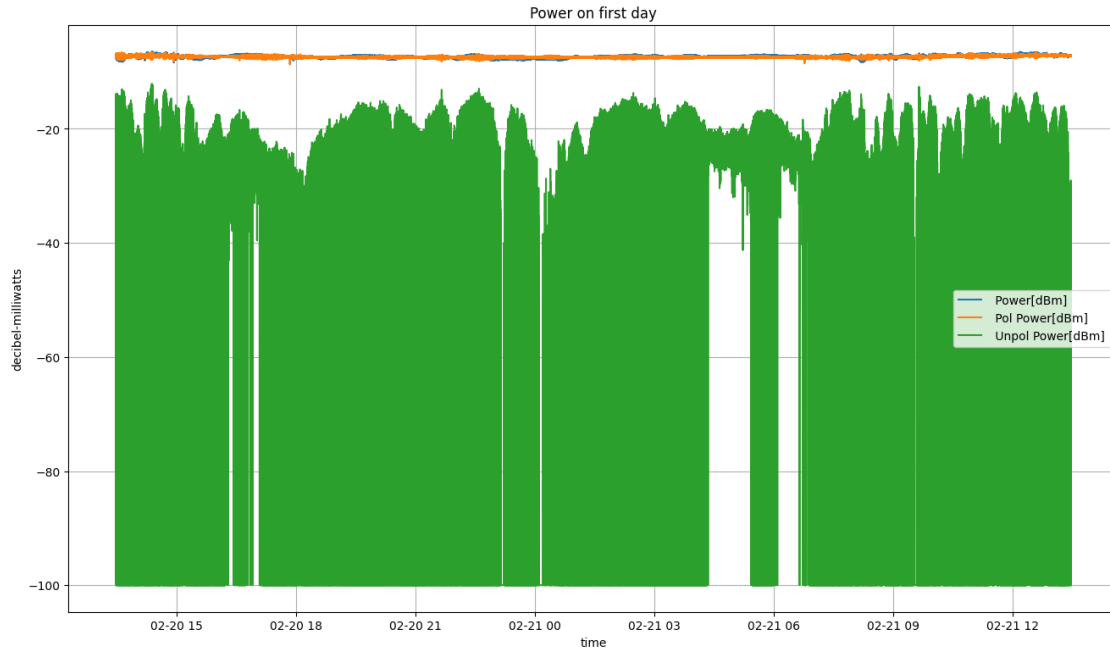
```
[53]: first_hour = power_dbm.loc[(power_dbm.index >= start_time) & (power_dbm.index <
    ↪end_time_hour)]

plt.figure(figsize = (16,9))
plt.plot(first_hour.index, first_hour[columns[17]], label = columns[17])
plt.plot(first_hour.index, first_hour[columns[18]], label = columns[18])
plt.plot(first_hour.index, first_hour[columns[19]], label = columns[19])
plt.grid()
plt.legend(loc = 'best')
plt.title('Power in first hour')
plt.xlabel('time')
plt.ylabel('decibel-milliwatts')
plt.show()
```



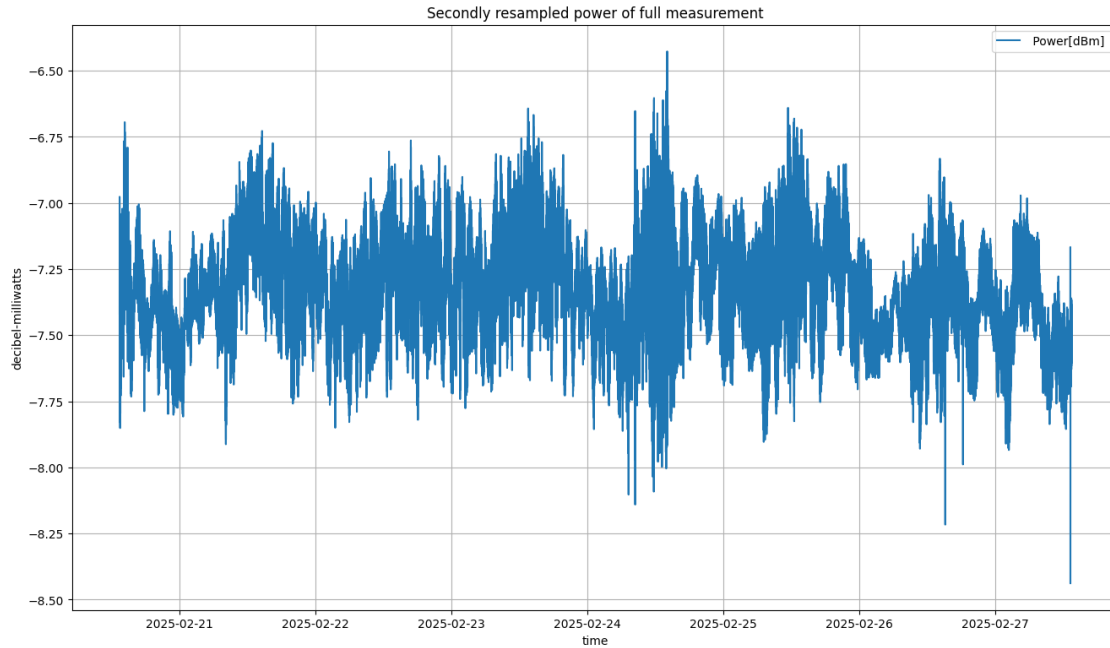
```
[54]: first_day = power_dbm.loc[(power_dbm.index >= start_time) & (power_dbm.index <
    ↪ end_time_day)]

plt.figure(figsize = (16,9))
plt.plot(first_day.index, first_day[columns[17]], label = columns[17])
plt.plot(first_day.index, first_day[columns[18]], label = columns[18])
plt.plot(first_day.index, first_day[columns[19]], label = columns[19])
plt.grid()
plt.legend(loc = 'best')
plt.title('Power on first day')
plt.xlabel('time')
plt.ylabel('decibel-milliwatts')
plt.show()
```

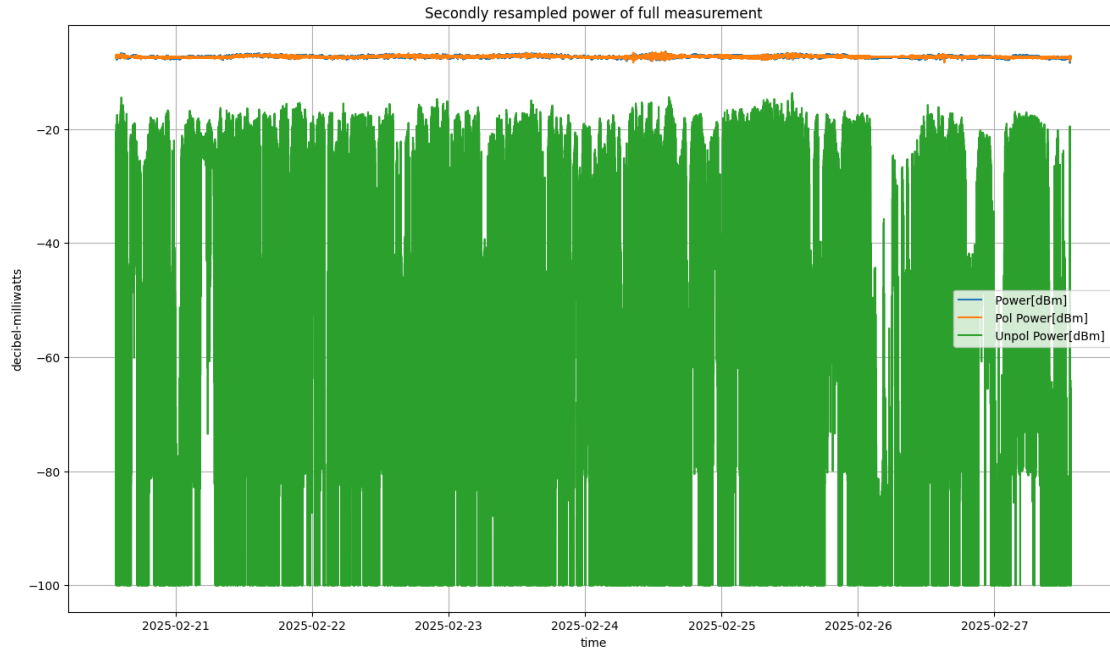


```
[55]: power_dbm_seconds_resample = power_dbm.resample('s').mean()
```

```
[56]: plt.figure(figsize = (16,9))
plt.plot(power_dbm_seconds_resample.index,
         ↪power_dbm_seconds_resample[columns[17]], label = columns[17])
plt.grid()
plt.legend(loc = 'best')
plt.title('Secondly resampled power of full measurement')
plt.xlabel('time')
plt.ylabel('decibel-milliwatts')
plt.show()
```



```
[57]: plt.figure(figsize = (16,9))
plt.plot(power_dbm_seconds_resample.index,
         ↪power_dbm_seconds_resample[columns[17]], label = columns[17])
plt.plot(power_dbm_seconds_resample.index,
         ↪power_dbm_seconds_resample[columns[18]], label = columns[18])
plt.plot(power_dbm_seconds_resample.index,
         ↪power_dbm_seconds_resample[columns[19]], label = columns[19])
plt.grid()
plt.legend(loc = 'best')
plt.title('Secondly resampled power of full measurement')
plt.xlabel('time')
plt.ylabel('decibel-milliwatts')
plt.show()
```



3.5 Winkel (Azimuth, Ellipticity)

```
[58]: angle = pd.read_csv(filename, skiprows=skip, usecols=[columns[0], columns[9], ↵
      ↪ columns[10]], sep=sep)
```

```
[59]: angle[columns[0]] = pd.to_datetime(angle[columns[0]])
      angle.set_index(columns[0], inplace=True)
```

```
[60]: angle.info()
```

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 2369654 entries, 2025-02-20 13:28:29.460000 to 2025-02-27
13:28:39.060000
Data columns (total 2 columns):
#   Column          Dtype
---  -
0   Azimuth[°]      float64
1   Ellipticity[°]  float64
dtypes: float64(2)
memory usage: 54.2 MB
```

```
[61]: angle.isnull().sum()
```

```
[61]: Azimuth[°]      0
      Ellipticity[°]  0
      dtype: int64
```



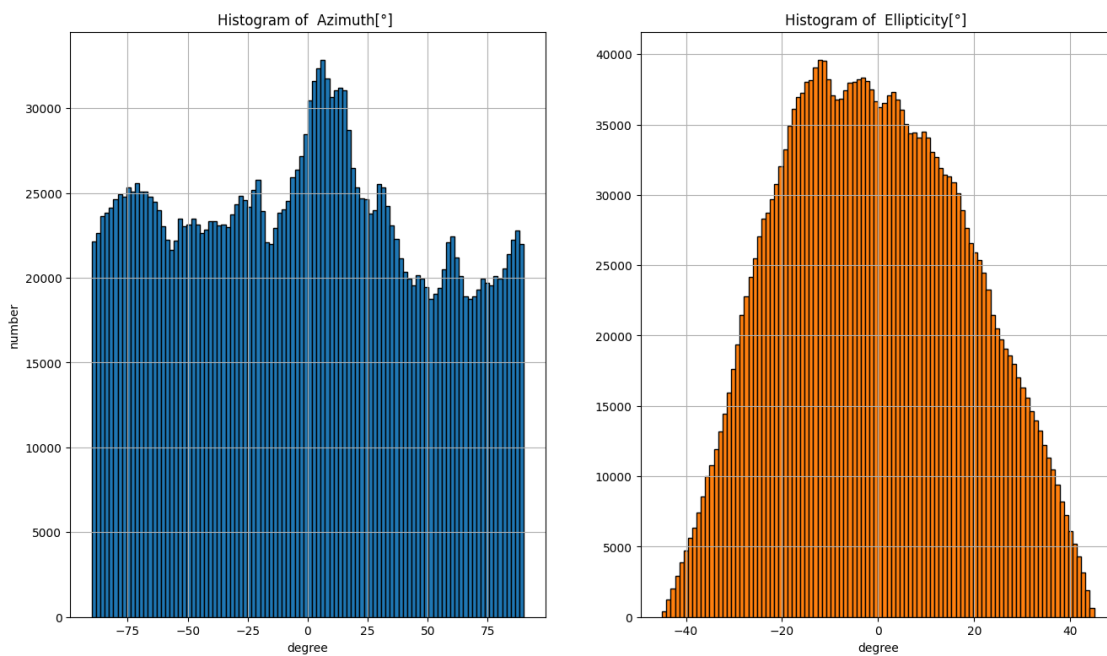
```
[62]: angle.describe()
```

```
[62]:      Azimuth[°]  Ellipticity[°]
count  2.369654e+06  2.369654e+06
mean   -2.375038e+00 -8.207911e-01
std     5.038627e+01  1.906634e+01
min     -9.000000e+01 -4.498000e+01
25%     -4.525000e+01 -1.561000e+01
50%     -9.900000e-01 -1.620000e+00
75%      3.744000e+01  1.354000e+01
max      9.000000e+01  4.497000e+01
```

```
[63]: plt.figure(figsize = (16,9))
plt.subplot(1, 2, 1)
plt.hist(angle[columns[9]], bins=100, color='tab:blue', edgecolor='black')
plt.grid()
plt.xlabel('degree')
plt.ylabel('number')
plt.title(f'Histogram of {columns[9]}')

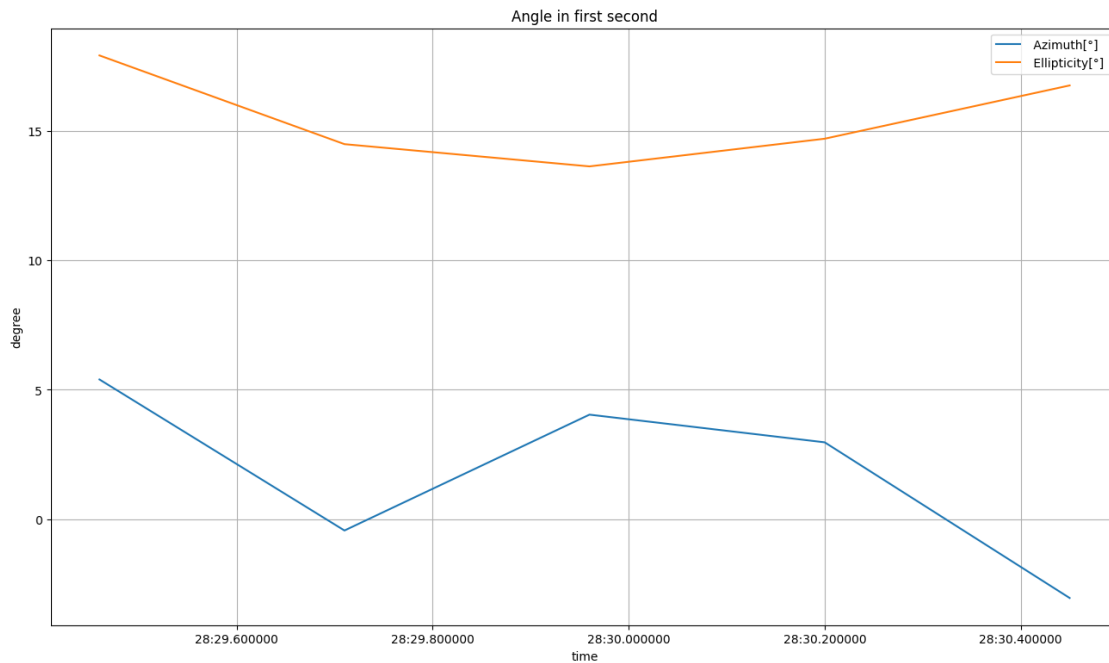
plt.subplot(1, 2, 2)
plt.hist(angle[columns[10]], bins=100, color='tab:orange', edgecolor='black')
plt.grid()
plt.xlabel('degree')
plt.title(f'Histogram of {columns[10]}')

plt.show()
```



```
[64]: first_second = angle.loc[(angle.index >= start_time) & (angle.index <
    ↪end_time_second)]

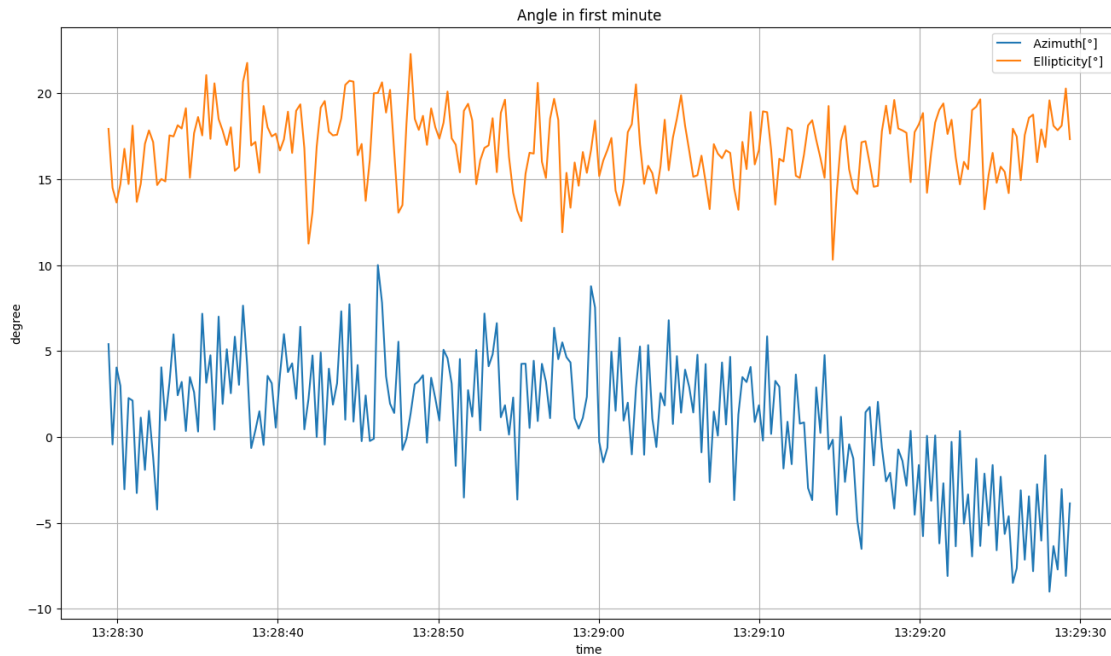
plt.figure(figsize = (16,9))
plt.plot(first_second.index, first_second[columns[9]], label = columns[9])
plt.plot(first_second.index, first_second[columns[10]], label = columns[10])
plt.grid()
plt.legend(loc = 'best')
plt.title('Angle in first second')
plt.xlabel('time')
plt.ylabel('degree')
plt.show()
```



```
[65]: first_minute = angle.loc[(angle.index >= start_time) & (angle.index <
    ↪end_time_minute)]

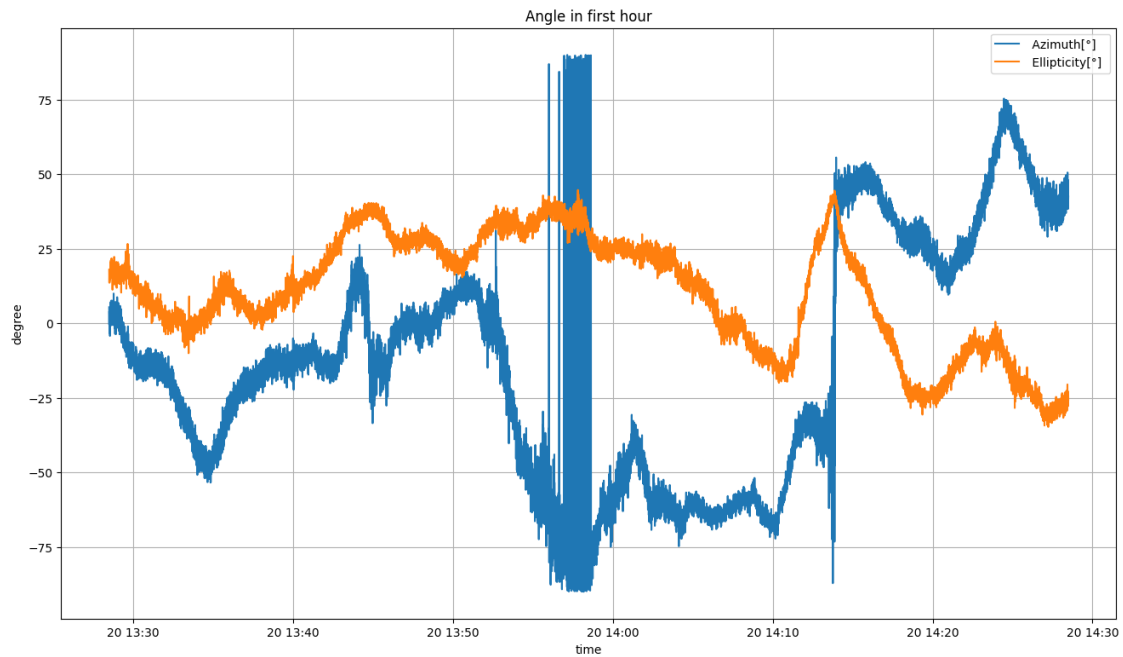
plt.figure(figsize = (16,9))
plt.plot(first_minute.index, first_minute[columns[9]], label = columns[9])
plt.plot(first_minute.index, first_minute[columns[10]], label = columns[10])
plt.grid()
plt.legend(loc = 'best')
plt.title('Angle in first minute')
plt.xlabel('time')
```

```
plt.ylabel('degree')
plt.show()
```



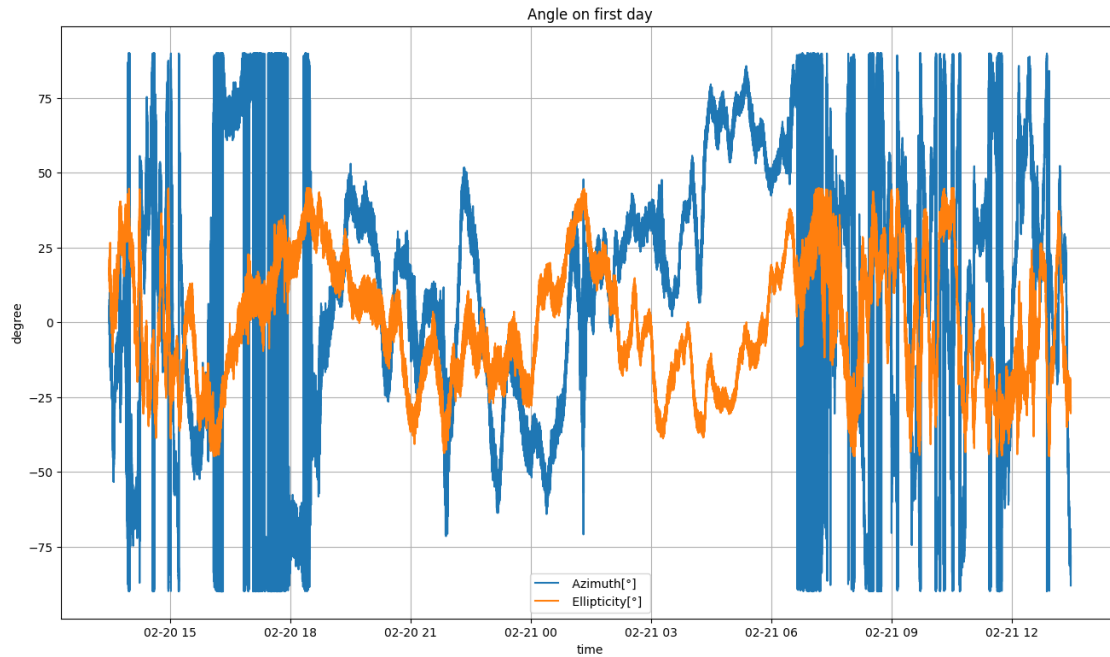
```
[66]: first_hour = angle.loc[(angle.index >= start_time) & (angle.index <
    ↪end_time_hour)]

plt.figure(figsize = (16,9))
plt.plot(first_hour.index, first_hour[columns[9]], label = columns[9])
plt.plot(first_hour.index, first_hour[columns[10]], label = columns[10])
plt.grid()
plt.legend(loc = 'best')
plt.title('Angle in first hour')
plt.xlabel('time')
plt.ylabel('degree')
plt.show()
```



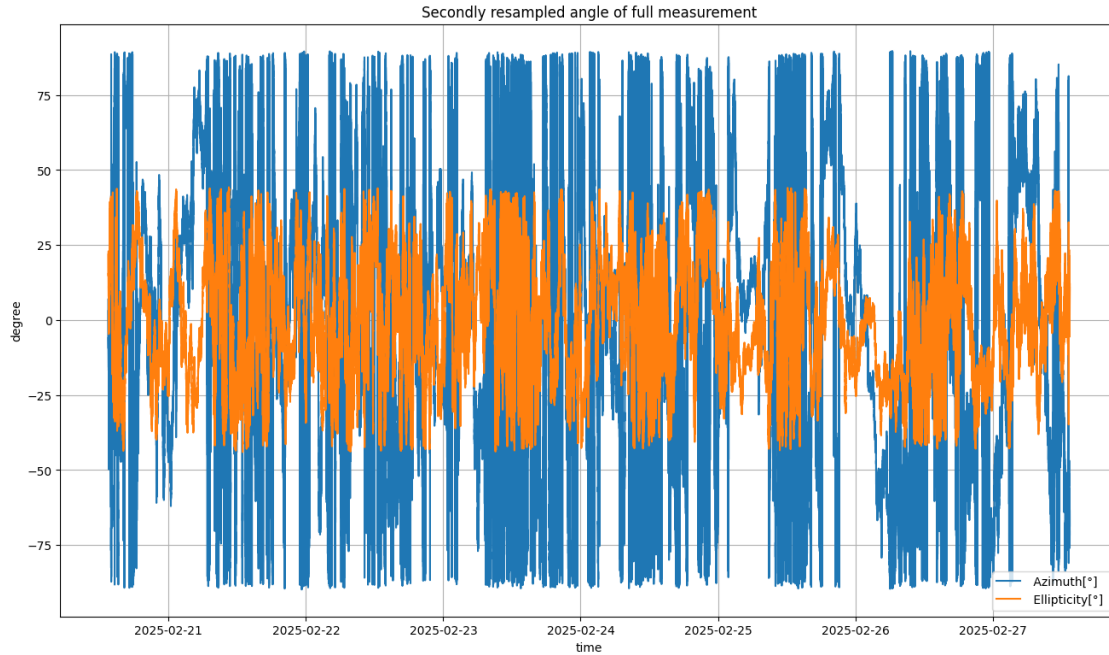
```
[67]: first_day = angle.loc[(angle.index >= start_time) & (angle.index <
    ↪end_time_day)]

plt.figure(figsize = (16,9))
plt.plot(first_day.index, first_day[columns[9]], label = columns[9])
plt.plot(first_day.index, first_day[columns[10]], label = columns[10])
plt.grid()
plt.legend(loc = 'best')
plt.title('Angle on first day')
plt.xlabel('time')
plt.ylabel('degree')
plt.show()
```



```
[68]: angle_seconds_resample = angle.resample('s').mean()
```

```
[69]: plt.figure(figsize = (16,9))
plt.plot(angle_seconds_resample.index, angle_seconds_resample[columns[9]],
         ↪label = columns[9])
plt.plot(angle_seconds_resample.index, angle_seconds_resample[columns[10]],
         ↪label = columns[10])
plt.grid()
plt.legend(loc = 'best')
plt.title('Secondly resampled angle of full measurement')
plt.xlabel('time')
plt.ylabel('degree')
plt.show()
```



3.6 Stokes-Parameter (S0, S1, S2, S3 / Normalized S1, S2, S3)

```
[70]: stokes = pd.read_csv(filename, skiprows=skip, usecols=[columns[0], columns[5],
↪ columns[6], columns[7], columns[8]], sep=sep)
```

C:\Users\laura\AppData\Local\Temp\ipykernel_8048\3101761779.py:1: DtypeWarning: Columns (5) have mixed types. Specify dtype option on import or set low_memory=False.

```
stokes = pd.read_csv(filename, skiprows=skip, usecols=[columns[0], columns[5],
columns[6], columns[7], columns[8]], sep=sep)
```

```
[71]: stokes[columns[0]] = pd.to_datetime(stokes[columns[0]])
stokes.set_index(columns[0], inplace=True)
```

```
[72]: stokes.info()
```

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 2369654 entries, 2025-02-20 13:28:29.460000 to 2025-02-27
13:28:39.060000
Data columns (total 4 columns):
#   Column      Dtype
---  -
0   S 0 [mW]    object
1   S 1 [mW]    float64
2   S 2 [mW]    float64
3   S 3 [mW]    float64
```

```
dtypes: float64(3), object(1)
memory usage: 90.4+ MB
```

```
[73]: stokes.isnull().sum()
```

```
[73]: S 0 [mW]    0
      S 1 [mW]    0
      S 2 [mW]    0
      S 3 [mW]    0
      dtype: int64
```

```
[74]: stokes.describe()
```

```
[74]:
```

	S 1 [mW]	S 2 [mW]	S 3 [mW]
count	2.369654e+06	2.369654e+06	2.369654e+06
mean	1.119711e-02	-3.962398e-03	-5.020136e-03
std	1.096478e-01	1.056081e-01	1.047504e-01
min	-2.349000e-01	-2.198000e-01	-2.017000e-01
25%	-8.511000e-02	-9.668000e-02	-9.554000e-02
50%	1.477000e-02	-2.372500e-03	-1.044000e-02
75%	1.104000e-01	8.352000e-02	8.420000e-02
max	2.405000e-01	2.174000e-01	2.031000e-01

```
[75]: normalized_stokes = pd.read_csv(filename, skiprows=skip, usecols=[columns[0],  
    ↪ columns[2], columns[3], columns[4]], sep=sep)
```

```
[76]: normalized_stokes[columns[0]] = pd.to_datetime(normalized_stokes[columns[0]])
      normalized_stokes.set_index(columns[0], inplace=True)
```

```
[77]: normalized_stokes.info()
```

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 2369654 entries, 2025-02-20 13:28:29.460000 to 2025-02-27
13:28:39.060000
Data columns (total 3 columns):
#   Column          Dtype
---  -
0   Normalized s 1   float64
1   Normalized s 2   float64
2   Normalized s 3   float64
dtypes: float64(3)
memory usage: 72.3 MB
```

```
[78]: normalized_stokes.isnull().sum()
```

```
[78]: Normalized s 1    0
      Normalized s 2    0
      Normalized s 3    0
```

dtype: int64

```
[79]: normalized_stokes.describe()
```

```
[79]:
```

	Normalized s 1	Normalized s 2	Normalized s 3
count	2.369654e+06	2.369654e+06	2.369654e+06
mean	6.296412e-02	-1.947966e-02	-2.810160e-02
std	5.914234e-01	5.698981e-01	5.659980e-01
min	-1.000000e+00	-1.000000e+00	-1.000000e+00
25%	-4.600000e-01	-5.200000e-01	-5.200000e-01
50%	8.000000e-02	-1.000000e-02	-6.000000e-02
75%	6.000000e-01	4.500000e-01	4.600000e-01
max	1.000000e+00	1.000000e+00	1.000000e+00

3.7 Power-Split-Ratio & Phase Difference

```
[80]: modality = pd.read_csv(filename, skiprows=skip, usecols=[columns[0],  
↪ columns[20], columns[21]], sep=sep)
```

```
[81]: modality[columns[0]] = pd.to_datetime(modality[columns[0]])  
modality.set_index(columns[0], inplace=True)
```

```
[82]: modality.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
DatetimeIndex: 2369654 entries, 2025-02-20 13:28:29.460000 to 2025-02-27  
13:28:39.060000  
Data columns (total 2 columns):  
#   Column                Dtype  
---  ----  
0   Power-Split-Ratio      float64  
1   Phase Difference[°]     float64  
dtypes: float64(2)  
memory usage: 54.2 MB
```

```
[83]: modality.isnull().sum()
```

```
[83]: Power-Split-Ratio      0  
Phase Difference[°]      0  
dtype: int64
```

```
[84]: modality.describe()
```

```
[84]:
```

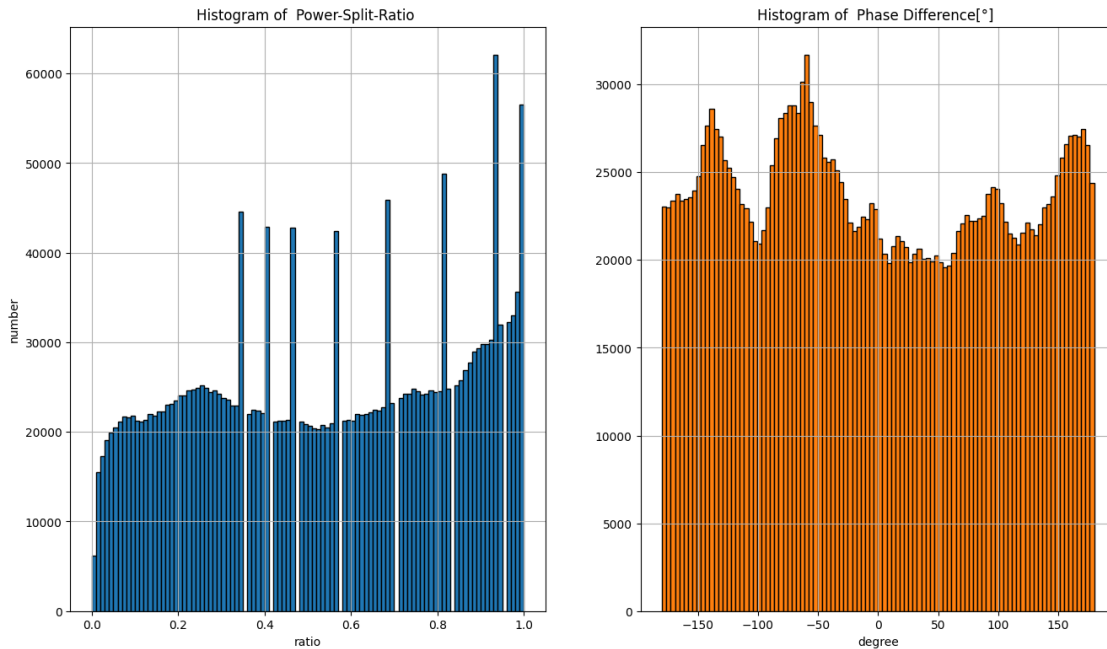
	Power-Split-Ratio	Phase Difference[°]
count	2.369654e+06	2.369654e+06
mean	5.314885e-01	-3.031721e+00
std	2.957300e-01	1.053810e+02
min	0.000000e+00	-1.800000e+02

25%	2.700000e-01	-9.181000e+01
50%	5.400000e-01	-1.049000e+01
75%	8.000000e-01	9.062000e+01
max	1.000000e+00	1.800000e+02

```
[85]: plt.figure(figsize = (16,9))
plt.subplot(1, 2, 1)
plt.hist(modality[columns[20]], bins=100, color='tab:blue', edgecolor='black')
plt.grid()
plt.xlabel('ratio')
plt.ylabel('number')
plt.title(f'Histogram of {columns[20]}')

plt.subplot(1, 2, 2)
plt.hist(modality[columns[21]], bins=100, color='tab:orange', edgecolor='black')
plt.grid()
plt.xlabel('degree')
plt.title(f'Histogram of {columns[21]}')

plt.show()
```



```
[86]: first_second = modality.loc[(modality.index >= start_time) & (modality.index <
    ↪end_time_second)]

fig, ax1 = plt.subplots(figsize=(16, 9))
```

```

ax1.plot(first_second.index, first_second[columns[20]], color='tab:blue',
        ↪label=columns[20])
ax1.set_xlabel('time')
ax1.set_ylabel('ratio')
ax1.grid()

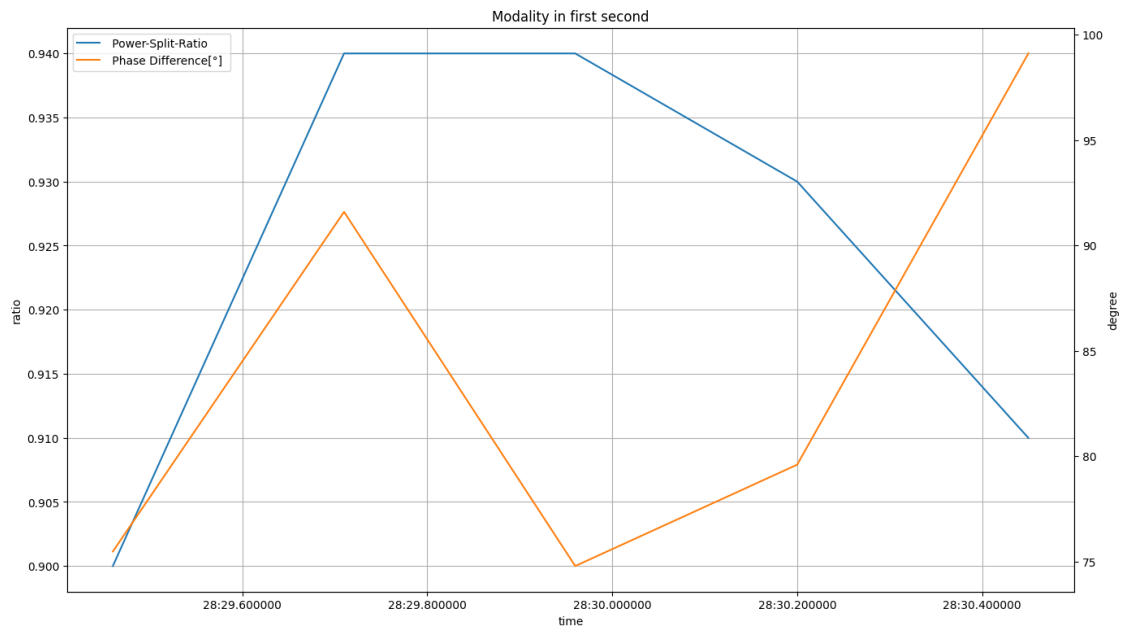
ax2 = ax1.twinx()
ax2.plot(first_second.index, first_second[columns[21]], color='tab:orange',
        ↪label=columns[21])
ax2.set_ylabel('degree')

plt.title('Modality in first second')

lines_1, labels_1 = ax1.get_legend_handles_labels()
lines_2, labels_2 = ax2.get_legend_handles_labels()
plt.legend(lines_1 + lines_2, labels_1 + labels_2, loc='best')

plt.show()

```



```

[87]: first_minute = modality.loc[(modality.index >= start_time) & (modality.index <
        ↪end_time_minute)]

fig, ax1 = plt.subplots(figsize=(16, 9))

```

```

ax1.plot(first_minute.index, first_minute[columns[20]], color='tab:blue',
        label=columns[20])
ax1.set_xlabel('time')
ax1.set_ylabel('ratio')
ax1.grid()

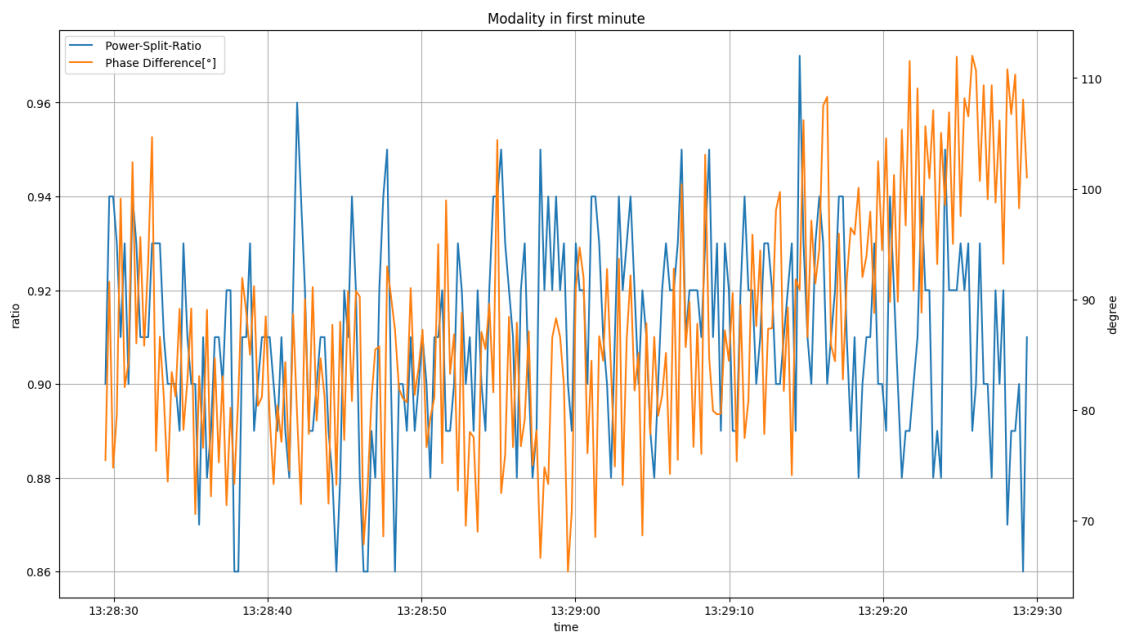
ax2 = ax1.twinx()
ax2.plot(first_minute.index, first_minute[columns[21]], color='tab:orange',
        label=columns[21])
ax2.set_ylabel('degree')

plt.title('Modality in first minute')

lines_1, labels_1 = ax1.get_legend_handles_labels()
lines_2, labels_2 = ax2.get_legend_handles_labels()
plt.legend(lines_1 + lines_2, labels_1 + labels_2, loc='best')

plt.show()

```



```

[88]: first_hour = modality.loc[(modality.index >= start_time) & (modality.index <
        end_time_hour)]

fig, ax1 = plt.subplots(figsize=(16, 9))

ax1.plot(first_hour.index, first_hour[columns[20]], color='tab:blue',
        label=columns[20])

```

```

ax1.set_xlabel('time')
ax1.set_ylabel('ratio')
ax1.grid()

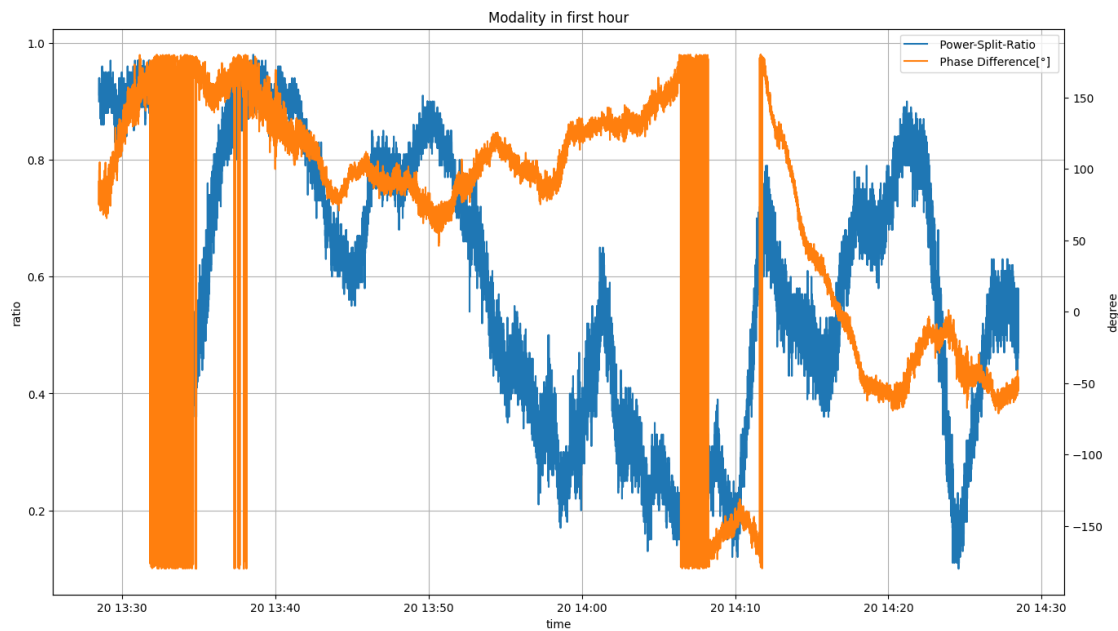
ax2 = ax1.twinx()
ax2.plot(first_hour.index, first_hour[columns[21]], color='tab:orange',
        ↪label=columns[21])
ax2.set_ylabel('degree')

plt.title('Modality in first hour')

lines_1, labels_1 = ax1.get_legend_handles_labels()
lines_2, labels_2 = ax2.get_legend_handles_labels()
plt.legend(lines_1 + lines_2, labels_1 + labels_2, loc='best')

plt.show()

```



```

[89]: first_day = modality.loc[(modality.index >= start_time) & (modality.index <
        ↪end_time_day)]

fig, ax1 = plt.subplots(figsize=(16, 9))

ax1.plot(first_day.index, first_day[columns[20]], color='tab:blue',
        ↪label=columns[20])
ax1.set_xlabel('time')
ax1.set_ylabel('ratio')

```

```

ax1.grid()

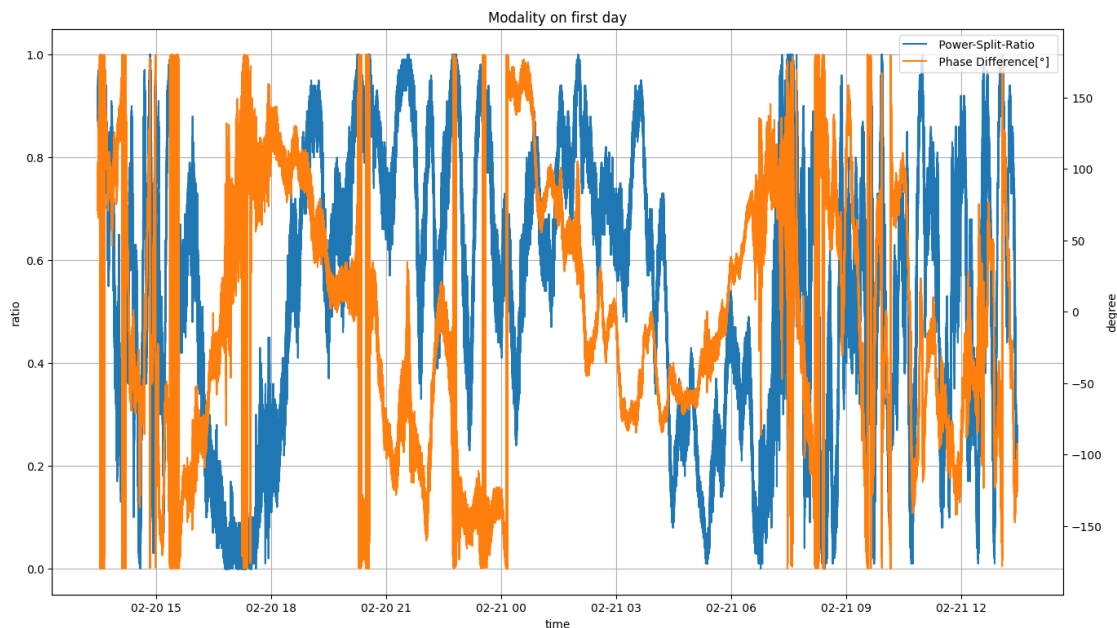
ax2 = ax1.twinx()
ax2.plot(first_day.index, first_day[columns[21]], color='tab:orange',
        ↪label=columns[21])
ax2.set_ylabel('degree')

plt.title('Modality on first day')

lines_1, labels_1 = ax1.get_legend_handles_labels()
lines_2, labels_2 = ax2.get_legend_handles_labels()
plt.legend(lines_1 + lines_2, labels_1 + labels_2, loc='best')

plt.show()

```



```
[90]: modality_seconds_resample = modality.resample('s').mean()
```

```

[91]: fig, ax1 = plt.subplots(figsize=(16, 9))

ax1.plot(modality_seconds_resample.index,
        ↪modality_seconds_resample[columns[20]], color='tab:blue', label=columns[20])
ax1.set_xlabel('time')
ax1.set_ylabel('ratio')
ax1.grid()

ax2 = ax1.twinx()

```

```

ax2.plot(modality_seconds_resample.index,
        ↪modality_seconds_resample[columns[21]], color='tab:orange',
        ↪label=columns[21])
ax2.set_ylabel('degree')

plt.title('Secondly resampled modality of full measurement')

lines_1, labels_1 = ax1.get_legend_handles_labels()
lines_2, labels_2 = ax2.get_legend_handles_labels()
plt.legend(lines_1 + lines_2, labels_1 + labels_2, loc='best')

plt.show()

```

