

Three-way_timestamps_Rpi

October 26, 2019

1 Comparison of timestamps in three alternative NMEA data paths

1.1 Rasbian “Buster” - OpenCPN v5.0.0 - Signal K v1.17.0 - DashT v0.5.2

We observe a five minute sampling period stored in InfluxDB database for each of the use case for single value of Apparent Wind Angle:

1. data via Signal K delta TCP channel with Signal K timestamps at its own reception
2. data via Signal K to NMEA-0183 via TCP channel timestamps at reception at the InfluxDB instruments
3. data directly from USB to OpenCPN

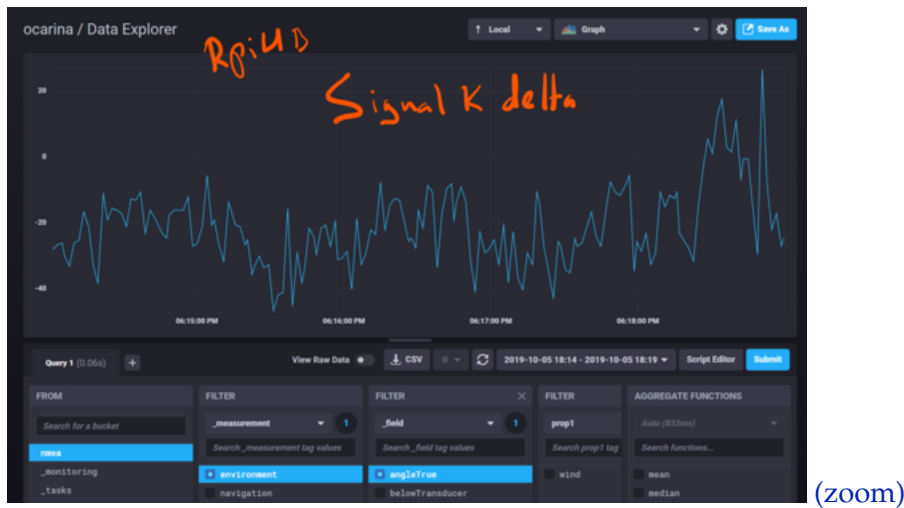
In all above cases the USB is set to 115200 baud at reception on Rasbian “Buster” (Raspberry Pi 4B 4GB) running OpenCPN v5.0.0. Data is originated from Raymarine SeaTalk (4800 baud) and converted to USB in MiniPlex II multiplexer - about 40 values per second are transmitted through this channel but only True Wind Angle timestamp behaviour is observed in this test.



1.2 Numerical comparison

```
[33]: import numpy as np
import pandas as pd
```

1.2.1 Data via Signal K delta TCP channel



```
[34]: df = pd.read_csv("2019-10-05_224308_SignalK_delta_zoom_Rpi.csv", sep=',',  
    ↪header=3)
```

```
[35]: df.head()
```

```
[35]: Unnamed: 0  result  table      _start      _stop \  
0         NaN     NaN      0  2019-10-05T16:14:00Z  2019-10-05T16:19:00Z  
1         NaN     NaN      0  2019-10-05T16:14:00Z  2019-10-05T16:19:00Z  
2         NaN     NaN      0  2019-10-05T16:14:00Z  2019-10-05T16:19:00Z  
3         NaN     NaN      0  2019-10-05T16:14:00Z  2019-10-05T16:19:00Z  
4         NaN     NaN      0  2019-10-05T16:14:00Z  2019-10-05T16:19:00Z  
  
      _time  _value  _field _measurement prop1  
0  2019-10-05T16:14:01.169Z  -28.1  angleTrue  environment  wind  
1  2019-10-05T16:14:03.106Z  -26.6  angleTrue  environment  wind  
2  2019-10-05T16:14:05.045Z  -26.1  angleTrue  environment  wind  
3  2019-10-05T16:14:06.049Z  -30.1  angleTrue  environment  wind  
4  2019-10-05T16:14:07.956Z  -33.1  angleTrue  environment  wind
```

```
[36]: df._value.describe()
```

```
[36]: count      166.000000  
mean       -21.151807  
std         11.759992  
min        -46.600000  
25%        -28.975000  
50%        -22.100000  
75%        -14.725000  
max         26.500000  
Name: _value, dtype: float64
```

```
[37]: df1 = pd.to_datetime(df['_time'])
```

```
[38]: df1.describe()
```

```
[38]: count          166
      unique        166
      top    2019-10-05 16:15:40.881000+00:00
      freq              1
      first  2019-10-05 16:14:01.169000+00:00
      last   2019-10-05 16:18:59.339000+00:00
      Name: _time, dtype: object
```

```
[39]: df2 = df1.astype(np.int64).div(1e6)
```

```
[40]: df3 = df2.diff()
```

```
[41]: df3.describe()
```

```
[41]: count          165.000000
      mean          1807.090909
      std           313.462833
      min           1000.000000
      25%           1863.000000
      50%           1931.000000
      75%           1942.999756
      max           2225.000000
      Name: _time, dtype: float64
```

1.2.2 Data via Signal K to NMEA-0183 converter TCP channel



```
[42]: nf = pd.read_csv("2019-10-05_224509_SignalK_NMEA_TCP_zoom_Rpi.csv", sep=',',
    ↳ header=3)
```

```
[43]: nf.head()
```

```
[43]: Unnamed: 0  result  table      _start      _stop \
0         NaN     NaN      0  2019-10-05T16:23:00Z  2019-10-05T16:28:00Z
1         NaN     NaN      0  2019-10-05T16:23:00Z  2019-10-05T16:28:00Z
2         NaN     NaN      0  2019-10-05T16:23:00Z  2019-10-05T16:28:00Z
3         NaN     NaN      0  2019-10-05T16:23:00Z  2019-10-05T16:28:00Z
4         NaN     NaN      0  2019-10-05T16:23:00Z  2019-10-05T16:28:00Z

      _time  _value  _field _measurement prop1
0  2019-10-05T16:23:00.423Z    19.6  angleTrue  environment  wind
1  2019-10-05T16:23:02.379Z    26.6  angleTrue  environment  wind
2  2019-10-05T16:23:04.295Z    26.1  angleTrue  environment  wind
3  2019-10-05T16:23:05.38Z    26.1  angleTrue  environment  wind
4  2019-10-05T16:23:07.104Z    40.6  angleTrue  environment  wind
```

```
[44]: nf._value.describe()
```

```
[44]: count      188.000000
mean         21.622872
std           9.844945
min           0.500000
25%          13.850000
50%          22.850000
75%          29.600000
max          41.200000
Name: _value, dtype: float64
```

```
[45]: nf1 = pd.to_datetime(nf['_time'])
```

```
[46]: nf1.describe()
```

```
[46]: count              188
unique              188
top    2019-10-05 16:24:50.701000+00:00
freq              1
first    2019-10-05 16:23:00.423000+00:00
last     2019-10-05 16:27:59.431000+00:00
Name: _time, dtype: object
```

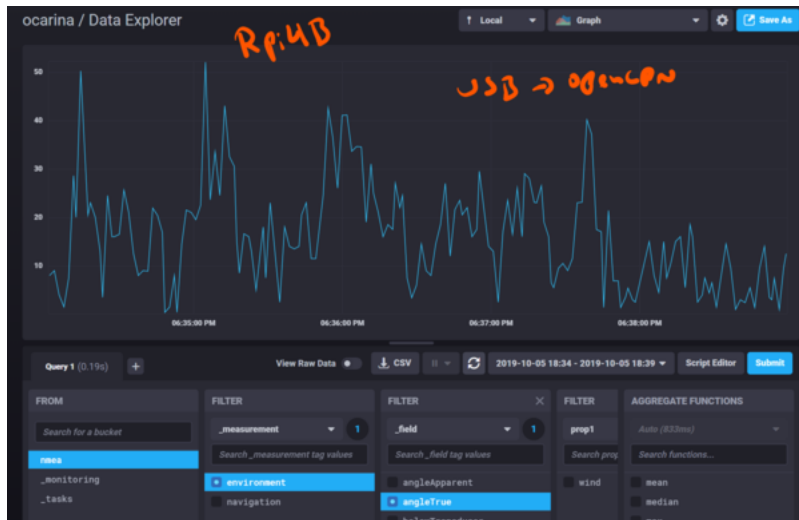
```
[47]: nf2 = nf1.astype(np.int64).div(1e6)
```

```
[48]: nf3 = nf2.diff()
```

```
[49]: nf3.describe()
```

```
[49]: count      187.000000
      mean      1598.973262
      std       422.940687
      min       1000.000000
      25%       1026.500000
      50%       1877.000000
      75%       1928.000000
      max       2036.000000
      Name: _time, dtype: float64
```

1.2.3 Data without Signal K directly from USB



(zoom)

```
[50]: of = pd.read_csv("2019-10-05_224657_USB_to_0_zoom_Rpi.csv", sep=',', header=3)
```

```
[51]: of.head()
```

```
[51]:   Unnamed: 0  result  table      _start      _stop \
0         NaN     NaN     0  2019-10-05T16:34:00Z  2019-10-05T16:39:00Z
1         NaN     NaN     0  2019-10-05T16:34:00Z  2019-10-05T16:39:00Z
2         NaN     NaN     0  2019-10-05T16:34:00Z  2019-10-05T16:39:00Z
3         NaN     NaN     0  2019-10-05T16:34:00Z  2019-10-05T16:39:00Z
4         NaN     NaN     0  2019-10-05T16:34:00Z  2019-10-05T16:39:00Z
```

```
      _time      _value      _field _measurement prop1
0  2019-10-05T16:34:01.719Z      8.1  angleTrue  environment  wind
1  2019-10-05T16:34:03.658Z      9.1  angleTrue  environment  wind
2  2019-10-05T16:34:05.585Z      4.1  angleTrue  environment  wind
3  2019-10-05T16:34:07.545Z      1.6  angleTrue  environment  wind
4  2019-10-05T16:34:09.462Z      7.6  angleTrue  environment  wind
```

```
[52]: of._value.describe()
```

```
[52]: count      172.000000
      mean       16.232558
      std        10.655020
      min        0.500000
      25%        7.900000
      50%       15.600000
      75%       22.225000
      max       52.100000
      Name: _value, dtype: float64
```

```
[53]: of1 = pd.to_datetime(of['_time'])
```

```
[54]: of1.describe()
```

```
[54]: count                172
      unique                172
      top    2019-10-05 16:36:01.753000+00:00
      freq                        1
      first  2019-10-05 16:34:01.719000+00:00
      last   2019-10-05 16:38:58.943000+00:00
      Name: _time, dtype: object
```

```
[55]: of2 = of1.astype(np.int64).div(1e6)
```

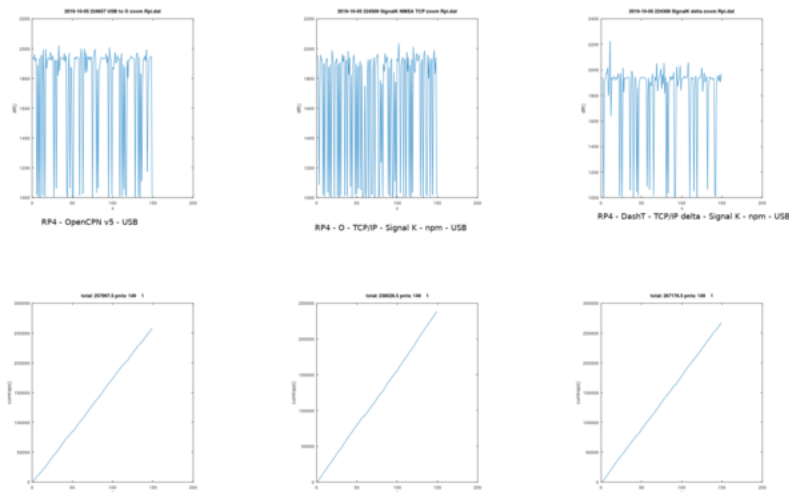
```
[56]: of3 = of2.diff()
```

```
[57]: of3.describe()
```

```
[57]: count      171.000000
      mean      1738.152047
      std       366.632300
      min      1000.000000
      25%      1825.499634
      50%      1925.000000
      75%      1942.500000
      max      2025.999756
      Name: _time, dtype: float64
```

1.3 Graphical comparison

GNU Octave v5.1/MATLAB script [jitterplots.m](#) was developed to present the jitter in graphical plot format, including the cumulative difference. The datafiles were truncated to hold 150 measurement points each. A helper script [raspberryyjitter.m](#) was used to produce the following plots.



(zoom)

1.4 Miscellaneous observations

1.4.1 Applying moving average



(zoom)

```
[58]: af = pd.read_csv("2019-10-05_224657_USB_to_0_zoom_Rpi_moving_average.csv",
    ↳ sep=',', header=3)
```

```
[59]: af.head()
```

```
[59]: Unnamed: 0  result  table  _start  _stop  \
0          NaN    NaN      0  2019-10-05T16:34:00Z  2019-10-05T16:39:00Z
1          NaN    NaN      0  2019-10-05T16:34:00Z  2019-10-05T16:39:00Z
2          NaN    NaN      0  2019-10-05T16:34:00Z  2019-10-05T16:39:00Z
3          NaN    NaN      0  2019-10-05T16:34:00Z  2019-10-05T16:39:00Z
4          NaN    NaN      0  2019-10-05T16:34:00Z  2019-10-05T16:39:00Z
```

```
      _time  _value  _field  _measurement  prop1
0  2019-10-05T16:34:33.669Z  18.570  angleTrue  environment  wind
```

```

1 2019-10-05T16:34:35.605Z 18.795 angleTrue environment wind
2 2019-10-05T16:34:37.541Z 18.745 angleTrue environment wind
3 2019-10-05T16:34:39.468Z 18.995 angleTrue environment wind
4 2019-10-05T16:34:41.444Z 19.365 angleTrue environment wind

```

```
[60]: af._value.describe()
```

```

[60]: count      153.000000
      mean        16.781928
      std         5.240853
      min         5.600000
      25%        13.300000
      50%        17.200000
      75%        20.085000
      max        27.700000
      Name: _value, dtype: float64

```

1.5 Summary of results

data path	timestamp	standard deviation	maximum time difference
1 Signal K delta	at source	313 ms	2225 ms
2 Signal K NMEA TCP	at reception	423 ms	2036 ms
3 USB to OpenCPN	at reception	367 ms	2025 ms

1.6 Conclusion

Judged by a human eye there is no difference between the three methods - the needles and values are jumping back and forth as always!

The difference will come apparent when we want to eliminate that jumping by applying some statistical and continuous algorithms on the received time series data. The accuracy of the time stamps is, of course important for any time series analysis.

Based on the graphs we make an assumption that the upstream system, the Miniplex multiplexer receives the wind data from SeaTalk with 1Hz (1/s) frequency and that it adds some jitter into the system but that jitter is not possible to measure. However, that particular jitter source is the same to all three methods observed, allowing the comparison

On Raspberry with Raspian Linux we can observe:

1. Signal K NMEA-0183 conversion and TCP/IP feed to OpenCPN which the passes the data to the plug-in loses less packages than the other two solutions - it receives more packages in a period. However, jitter-wise it the worst of the three - about every other second a data frame is missed.
2. Somewhat surprisingly, the direct connection of OpenCPN to the USB source is performing more badly than when it receives the data indirectly through Signal K using TCP/IP. The

jitter is less important, however.

3. Jitter-wise, the direct connection to the Signal K is the best, as expected. However, on this platform it is losing more packages than the two other methods. This can be explained partly by the low performance of the CPU on this platform - there is probably a saturation either on the npm-thread on Signal K side sending the delta-values or the on the POSIX communication thread of the Dashboard-Tactics plug-in receiving them, or on both.

On this platform the Signal K TCP feed to OpenCPN is the method to use if the number of samples is preferred, wxWidgets and OpenCPN TCP/IP implementation works very well here.

The direct Signal K connection's advantage is the reduced jitter also in this platform, not to mention the functional enhancements it provides. It can be used together with the Signal K TCP/IP feed since the requirement was precisely to reduce jitter for the key parameters.

Albeit the OpenCPN direct USB connection on a modest Raspberry Pi 4B board allows similar performances than more powerful i7 CPU based processor under Window 10, a similar level of jitter with lesser performance would suggest to prefer the usage of Signal K and TCP/IP. Of course, a direct TCP/IP connection to the Miniplex multiplexer is possible but not studied. Performance-wise this can be expected to provide similar results but since there is no performance penalty by using Signal K it would be pity not to profit from the extra functionality it provides.

Finally, let's consider the following statement from *Numerical Recipes In C: The Art of Scientific Computing* (ISBN 0-521-43108-5) concerning irregularly sampled data, where the values f_i are not uniformly spaced in time:

If the change in f across the full width of the $N = nL + nR + 1$ point window is less than $\sqrt{N}/2$ times the measurement noise on a single point, then the cheap method can be used.

The "*cheap method*" is to ignore the jitter. But as the above graphics illustrates, were are constantly switching, here for the wind data, between 1Hz and 0.5Hz for f_i !