# testbench

October 24, 2022

# 1 AtMonSat Detection Algorithm Test Bench

Datapoints from a dataset are send to the C++ implementation for evaluation and results are recorded.

## 1.0.1 Settings

Microcontroller / processor frequency (this value is used to select some settings (e.g. baudrate) and define the filename to store the results)

```
[]: microcontroller_frequency = '146' # [MHz]
```

### Connection type

A connection can be done to an implementation compiled for the host computer using pipes or to an implementation running on a microcontroller (e.g. STM32H743) using serial communication.

```
[]: # connection_type = 'local'
connection_type = 'remote'
```

## Settings for a remote connection

connection\_baudrate is the baudrate to use for the communication with the microntroller. This setting must match the baurate defined in the firmware.

connection\_port is the interface on the local machine that connects to the remote microcontroller.

## Settings for a local connection

connection local command is a list with the command as the first member.

```
[]: connection_local_command = ['../atmonsat']
```

### Dataset to transmit

All datasets are located in the datasets directory. The use\_dataset variable defines which dataset to use.

These measurements have been taken from the EPS of a cubesat. TEMP\_0 to TEMP\_3 are MPPT temperatures TEMP\_4 to TEMP\_8 are voltage regulator temperatures and TEMP\_9 represents the battery temperature.

All temperatures have a resolution of 1°C.

Several ranges have been defined for each dataset. - 'normal': range where no anomaly happens. - 'abnormal': range where one or multiple anomalies have been introduced. - 'experiment': normal + abnormal range.

The use range variable can be used to select what range of datapoints to use for evaluation.

```
[]: use_dataset = '2022.07.20'
use_range = 'experiment'
```

# Number of datapoints

limit\_number\_of\_datapoints limits the number of datapoints from dataset to be transmitted for evaluation. - None: all datapoints will be transmitted - any integer > 0: Maximum number of datapoints to transmit

```
[]: limit_number_of_datapoints = None # <int> or None
```

#### Threshold hold-off

A hold-off can be activated that suppresses any further detection after an anomaly detection during the number of specified iterations.

threshold\_hold\_off can take any positive integer value. 0 disables the postprocessing.

```
[]: threshold_hold_off = 60
```

Output directory (where to store the results). A directory will be created inside this output directory for each measurement.

```
[]: output_directory = './results'
```

Create output directory

```
[]: import datetime
import os
timestamp_string = datetime.datetime.now().strftime("%Y%m%d-%H%M%S")
output_directory = output_directory + os.path.sep + timestamp_string
os.mkdir(output_directory)
print("Results store in {}".format(output_directory))
```

Results store in ./results/20221024-130021

Save experiment settings for later reference into meta.json

```
[]: import json

settings = {
    'timestamp': timestamp_string,
```

```
'dataset': use_dataset,
    'range': use_range,
    'frequency': microcontroller_frequency,
    'connection_type': connection_type,
    'limit': limit_number_of_datapoints,
    'threshold_hold_off': threshold_hold_off,
}

print('Settings:')
for k, v in settings.items():
    print(" - {} = {}".format(k, v))

filename = output_directory + os.path.sep + 'settings.json'
with open(filename, 'w') as file:
    json.dump(settings, file)
    print("Settings saved to '{}'".format(filename))
```

## Settings:

```
- timestamp = 20221024-130021
- dataset = 2022.07.20
- range = experiment
- frequency = 146
- connection_type = remote
- limit = None
- threshold_hold_off = 60
Settings saved to './results/20221024-130021/settings.json'
```

#### 1.0.2 Includes

```
[]: %matplotlib inline
  import pandas as pd
  import matplotlib.pyplot as plt
  import numpy as np
  import os
  import datetime
  import time
  import pickle
  import sys

enable_example = False
  %run dataset.ipynb
  %run protocol.ipynb
```

## 1.0.3 Logger

Information, warning and errors are getting logged into atmonsat\_testbench.txt. The variable display\_log is a boolean that defines if the same information should be visualized on screen.

```
[]: display_log = True
```

The following function is getting called by the communication protocol and can be used to visualize data exchange between the host and the target.

```
[]: def comment(s):
    # print(s)
    pass
```

## 1.0.4 Preparation

Load the specified dataset

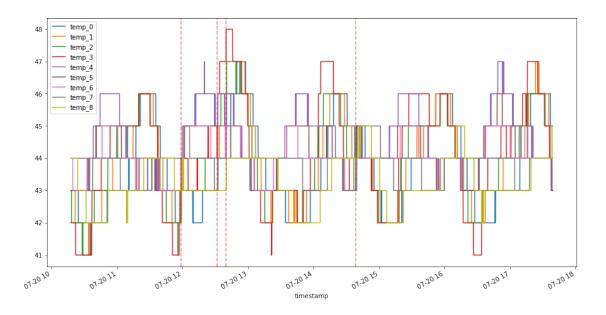
```
[ ]: dataset = Dataset(use_dataset)[use_range]
```

## **Features**

Features to be packed into a datapoint and send to the implementation. The order of the features must correspond to the ordering of a datapoint in the C++ implementation

Visualize the selected range of the dataset for inspection

[]: <\_\_main\_\_.Dataset at 0x7f743f143ac0>



## **Datapoints**

Prepare a list of datapoint commands to be send to the implementation.

```
[]: commands = []
for index, row in dataset.dataframe[feature_columns].iterrows():
    if not limit_number_of_datapoints is None:
        if len(commands) > limit_number_of_datapoints:
            break
    commands.append(CommandDatapointINT8(data=row.to_numpy(dtype=np.int8)))
```

#### **Datastore**

Datastore is a class that collects incoming distance and detection responses from the implementation.

```
return self.__df
    def save(self, filename="distance_detection.csv", sep=";"):
        self.__df.to_csv(filename, sep=sep)
    def __store(self):
        if self.__distance_arrived and self.__detection_arrived:
            # Store received row
            self.__df.loc[self.__row] = [time.time(),
                                          self.__recorded_distance, self.
 →__recorded_detection]
            self.__row = self.__row + 1
            self.__distance_arrived = False # reset flag
            self.__detection_arrived = False # reset flag
    def store_distance(self, distance):
        # logging.info("Malanobis distance={}".format(distance))
        self.__recorded_distance = distance
        if self.__distance_arrived:
            logging.error(
                "Recevied distance a second time before detection. This should \sqcup
 onot happen.")
        self.__distance_arrived = True
        self.__store()
    def store_detection(self, detection):
        # logging.info("Detection={}".format(detection))
        self.__recorded_detection = detection
        if self.__detection_arrived:
            logging.error(
                "Recevied detection a second time before distance. This should \sqcup
 →not happen.")
        self.__detection_arrived = True
        self.__store()
datastore = DataStore()
```

## **ExecutionTimeStore**

ExecutionTimeStore is a class that collects incoming execution time information.

```
@property
def dataframe(self):
    return self.__df

def response_handler_callback_execution_time(self, duration):
    # logging.info("Execution time={}".format(duration))
    self.__df.loc[self.__row] = [time.time(), float(duration*1e-9)]
    self.__row = self.__row + 1

def save(self, filename="execution_times.csv", sep=";"):
    self.__df.to_csv(filename, sep=sep)

execution_time_store = ExecutionTimeStore()
```

response\_handler\_callback\_remote\_exception is a response handler that is getting called when an exception occured in the implementation (local or remote)

```
[]: def response_handler_callback_remote_exception(str):
    text = "A remote exception occured: {}".format(str)
    logging.fatal("Exception={}".format(text))
    raise Exception(text)
```

### Communication response handlers

response\_handlers is a list that defines all allowed responses from the microcontroller and what handlers to call after reception.

#### Connect to the target

Connection can be done on a local target (compiled for the host) or a remote target over serial communication.

## Initialize target

This command initializes or resets the implementation. It has been implemented as a command as the initialization code might try to report errors which is a communication from target->host that is only allowed in reply to a command.

```
[]: connection.emit(CommandInitialize())
```

Comment=target initialized.

[]: True

## Upload coefficients

Upload coefficients stored in a save\_variables.pkl to the implementation. The uploaded coefficients are: - detection\_threshold - mahalanobis inverse covariance matrix - mahalanobis mean vector

```
[]: filename = 'saved_variables.pkl'
     if os.path.exists(filename):
         with open(filename, 'rb') as file:
             variables_dict = pickle.load(file)
             # logging.info("Threshold: {}".
      →format(float(variables_dict['detection_threshold'])))
             # logging.info("Matrix: {}".format(np.
      →array(variables dict['mahalanobis matrix']).flatten()))
             # logging.info("Mean: {}".format(np.
      →array(variables_dict['mahalanobis_mean']).flatten()))
             connection.emit(CommandFloat(id=PROTOCOL_COMMAND_ID_THRESHOLD,
      ⇔data=float(variables dict['detection threshold'])))
             connection.
      -emit(CommandMatrix(id=PROTOCOL_COMMAND_ID_MAHALANOBIS_INVERSE_COVARIANCE_MATRIX,
                                           data=np.
      →array(variables_dict['mahalanobis_matrix']).flatten()))
             connection.emit(CommandMatrix(id=PROTOCOL_COMMAND_ID_MAHALANOBIS_MEAN,
```

```
data=np.

Garray(variables_dict['mahalanobis_mean']).flatten()))
```

Comment=treshold updated

Comment=mahalanobis inverse covariance matrix has been updated Comment=mahalanobis mean matrix has been updated

## Upload hold-off

```
[]: connection.emit(CommandUInt32(id=PROTOCOL_COMMAND_ID_THRESHOLD_HOLD_OFF, ⊔

data=threshold_hold_off))
```

Comment=hold-off updated

[]: True

#### 1.0.5 Processing

Take time before sending all datapoints to estimate the total required time.

```
[]: start_time = time.time()
```

#### Upload the datapoints

```
[]: connection.emit(commands)
```

[]: True

# Total required time

This is the time of the evaluation + testbench and communication overhead. It gives an uppper bound on the duration.

Duration of evaluation + communication + testbench: 67.5136046409607 [s]

Store the received data and extend the dataset with responses

```
print("Number of missing detection responses: {} / {}".format(missing_values, __
 →len(dataset.dataframe.index)))
dataset.dataframe['detection'] = list(datastore.dataframe['detection']) +
→[0]*missing values
dataset.dataframe['distance'] = list(datastore.dataframe['distance']) +__
 →[0]*missing_values
dataset.dataframe['detection_recording_timestamp'] = list(datastore.

→dataframe['detection_recording_timestamp']) + [0]*missing_values

missing_values = len(dataset.dataframe.index) - len(execution_time_store.

dataframe['execution_time'])
print("Number of missing execution_time responses: {} / {}".

→format(missing_values, len(dataset.dataframe.index)))

dataset.dataframe['execution_time'] = list(execution_time_store.

dataframe['execution_time']) + [0]*missing_values

dataset.dataframe['execution_time_recording_timestamp'] =__
 ⇔list(execution_time_store.dataframe['execution_time_recording_timestamp']) + ∪
 →[0]*missing_values
```

Number of missing detection responses: 7 / 5164 Number of missing execution\_time responses: 0 / 5164

#### Save the dataset with the results to CSV

```
[]: filename = output_directory + os.path.sep + "testbench_data.csv" print("Dataset and results have been stored into: {}".format(filename)) dataset.save_dataframe_as_csv(filename=filename)
```

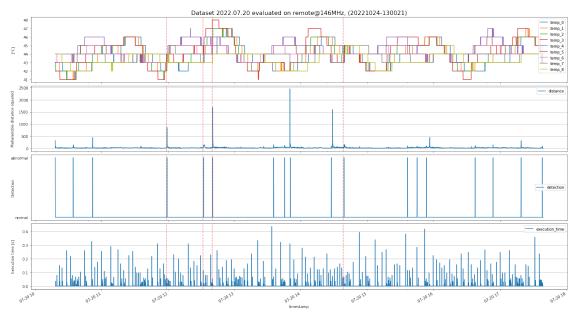
Dataset and results have been stored into: ./results/20221024-130021/testbench data.csv

[]: <\_\_main\_\_.Dataset at 0x7f743f143ac0>

## 1.0.6 Postprocessing

#### Visualize the dataset and the results

```
axs[1].yaxis.set_label_coords(*ylabel_position)
dataset.plot(columns=['distance'], ax=axs[1], grid=True).
 →plot_anomalies(ax=axs[1])
axs[1].grid(axis='x')
axs[2].set_ylabel("Detection")
axs[2].yaxis.set label coords(*ylabel position)
axs[2].set_yticks([1.0, 0.0], ["abnormal", "normal"])
dataset.plot(columns=['detection'], ax=axs[2]).plot_anomalies(ax=axs[2])
axs[3].set_ylabel("Execution time [s]")
axs[3].yaxis.set_label_coords(*ylabel_position)
dataset.plot(columns=['execution_time'], ax=axs[3], grid=True).
 →plot_anomalies(ax=axs[3])
axs[3].grid(axis='x')
fig.subplots_adjust(hspace=0.05)
filename = output_directory + os.path.sep + "testbench_results.png"
plt.savefig(filename)
```

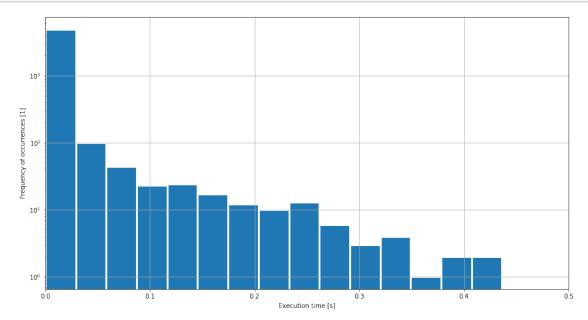


## Execution time statistics

```
"sample_size": len(column)
                  }
    print("Execution time / datapoint statistics")
    for k, v in execution_time_statistics.items():
        print("{} \t= {} [s]".format(k, v))
    print("")
    print("Description")
    print(column.describe())
    statistics = {"execution_time": execution_time_statistics}
    filename = output_directory + os.path.sep + "execution_time_statistics.json"
    with open(filename, 'w') as file:
        json.dump(statistics, file)
        print("Statistics stored to '{}'".format(filename))
    Execution time / datapoint statistics
    mean
           = 0.006044625484120837 [s]
    median = 1.00000000000000000e-06 [s]
           = 0.031117882778141696 [s]
           min
           = 0.43631000000000003 [s]
    max
            = 31.214446000000002 [s]
    sum
                   = 5164 [s]
    sample_size
    Description
    count
             5164.000000
               0.006045
    mean
               0.031118
    std
    min
               0.000001
    25%
               0.000001
    50%
                0.000001
    75%
                0.000002
                0.436310
    max
    Name: execution_time, dtype: float64
    Statistics stored to './results/20221024-130021/execution_time_statistics.json'
    Histogram of excution duration
[]: if connection_type == "local":
        ax=column.plot.hist(column=["execution_time"], bins=15, logy=True,__
      wxlim=[0,0.001], figsize=(15,8), edgecolor='white', linewidth=3, grid=True)
        ax=column.plot.hist(column=["execution_time"], bins=15, logy=True, xlim=[0,_
      ⇔0.5], figsize=(15,8), edgecolor='white', linewidth=3, grid=True)
```

```
ax.set_xlabel("Execution time [s]")
ax.set_ylabel("Frequency of occurrences [1]")

filename = output_directory + os.path.sep + "testbench_histogram.png"
plt.savefig(filename)
```



## Communication and testbench overhead

Calculate time overhead due to communication and testbench

```
[]: time_overhead = total_evaluation_time - column.sum()
    print("Total duration: {} [s]".format(total_evaluation_time))
    print("Evaluation duration: {} [s]".format(column.sum()))
    print("Overhead (communication and testbench): {} [s]".format(time_overhead))
```

Total duration: 67.5136046409607 [s]

Evaluation duration: 31.214446000000002 [s]

Overhead (communication and testbench): 36.29915864096069 [s]

### 1.0.7 Finalize

# Terminate the connection

[]: connection.close()