



**POLITECNICO**  
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SCUOLA DI INGEGNERIA INDUSTRIALE  
E DELL'INFORMAZIONE

# IoT Challenge #1, Wokwi and Power Consumption

INTERNET OF THINGS

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# 1 | Code explanation



## 2 | Energy consumption estimation

### 2.1. Power consumption estimation

In this section, there is a description of the power consumption of the ESP32 node in all states: deep sleep, idle, sensor reading, WiFi turned on and transmission.

In order to estimate power consumption based on the provided CSV files, we first plotted data using Python and Matplotlib library. Based on the plot, we computed the average power consumption in the analyzed working state.

#### 2.1.1. Power consumption plots

In order to plot the data contained in the provided CSV files, we used the following algorithm.

```
import pandas as pd
import matplotlib.pyplot as plt
# read CSV file
df = pd.read_csv('deep_sleep.csv', parse_dates=['Timestamp'])
# plot data
plt.figure(figsize=(10, 6))
plt.plot(df['Timestamp'], df['Data'], linestyle='--', linewidth=2)
plt.xlabel('Time')
plt.ylabel('Power [mW]')
plt.title('Power consumption deep sleep')
plt.grid(True)
# save and shows
plt.savefig("power_consumption_deep_sleep.pdf", format="pdf",
            bbox_inches="tight")
plt.show()
```

#### Power consumption deep\_sleep.csv

The following plot represents the power consumption when the ESP32 is in deep sleep state (lowest power level), goes into idle mode (medium power level) and finally turns the

WiFi on (highest power level).

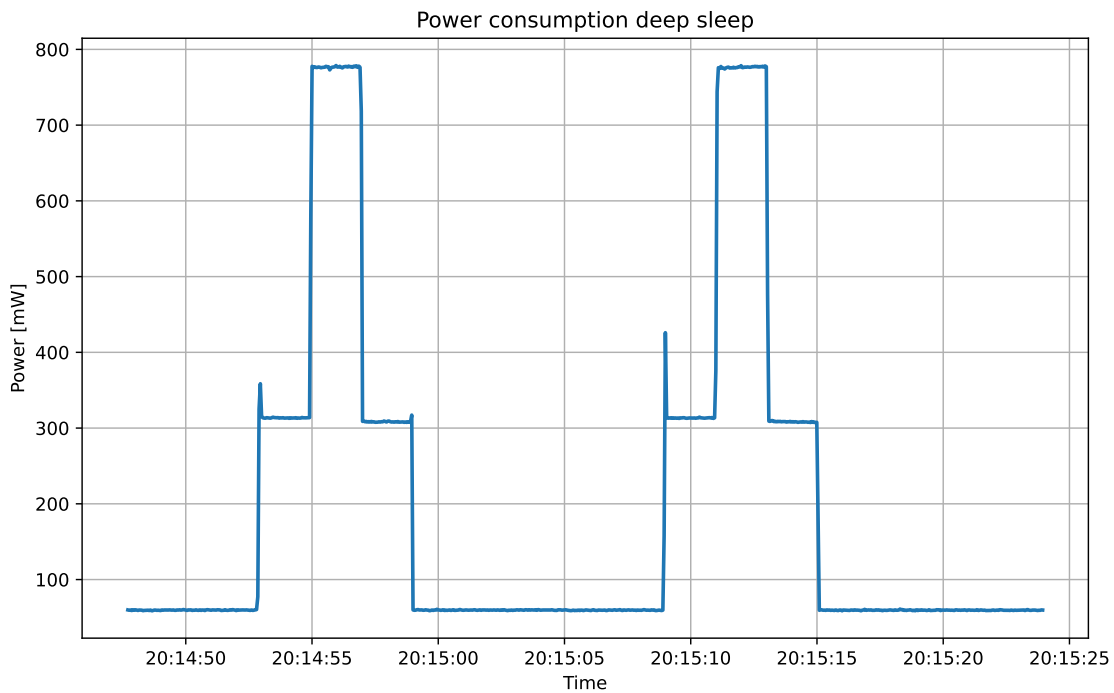


Figure 2.1: Power consumption deep sleep state

### Power consumption sensor\_read.csv

The following plot represents the power consumption when the ESP32 alternates the idle mode and sensor reading mode, in which it performs a measurement using the Ultrasonic Distance Sensor HC-SR04.



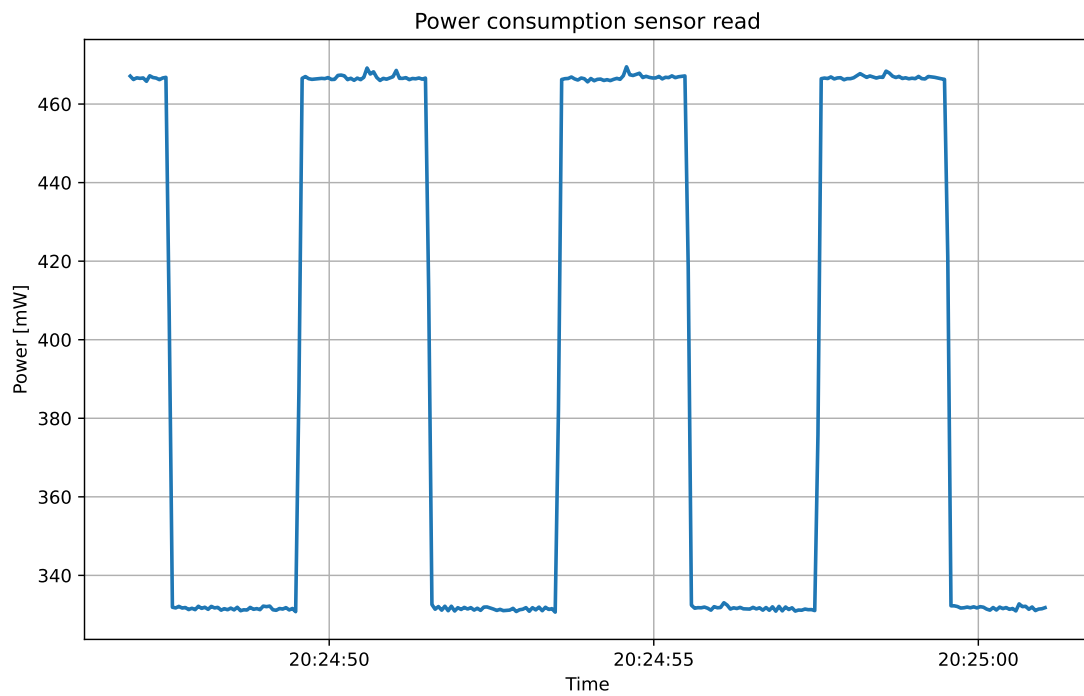


Figure 2.2: Power consumption deep sleep state

### Power consumption transmission\_power.csv

The following plot represents the power consumption when the ESP32 has the WiFi turned on and transmits data using ESP-NOW at 19.5 dBm and 2 dBm.

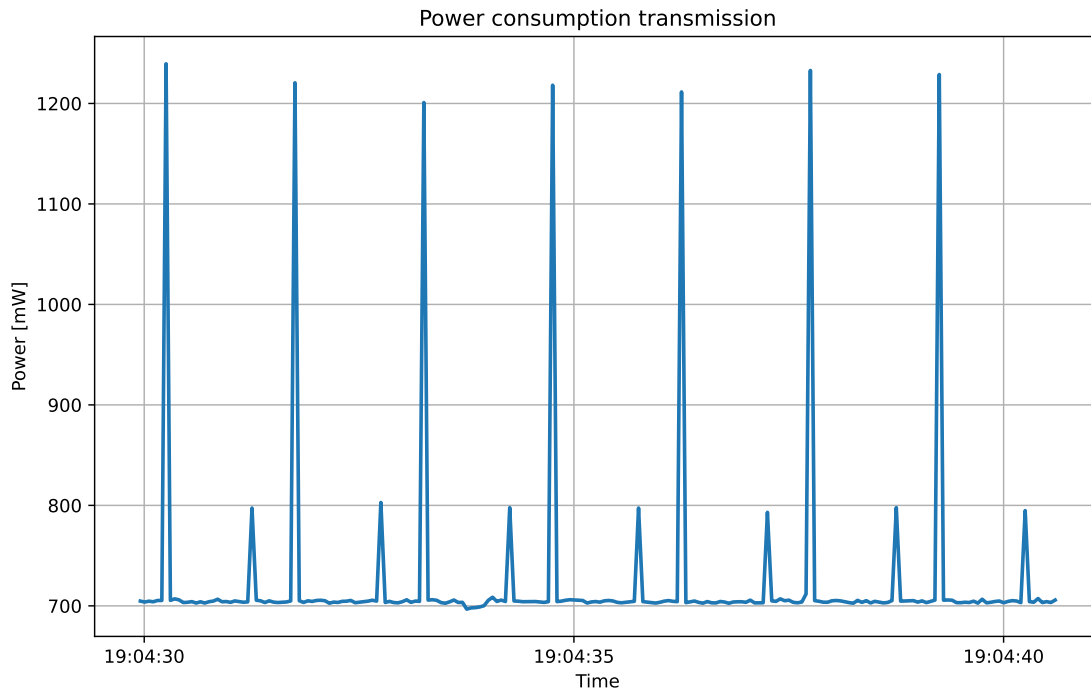


Figure 2.3: Power consumption deep sleep state

### 2.1.2. Average power consumption values

In order to find a numerical estimation of the power consumption in different states, we used some Python algorithms, using the library Pandas. Starting from the plot we selected the range of values regarding a particular state, discarding the others, and computed the average power consumption. For some states, the values of power can be found in multiple CSV files; in such cases, we merged different files in order to have a more accurate estimation.

In the following sections we also report two example of algorithms used to compute the average, both using a single dataset and multiple datasets.

#### Power consumption in deep sleep state

Data to estimate the deep sleep mode is contained in the dataset `deep_sleep.csv`. We filtered all values of power above 100mW, referring to other states as shown by the plot, and computed the average.

```
import pandas as pd
# read CSV file
dataset_deep_sleep = pd.read_csv("deep_sleep.csv", parse_dates=['
                                Timestamp'])
# filter data with power < 100mW, referring to deep sleep mode
```

```

deep_sleep_data = dataset_deep_sleep[dataset_deep_sleep["Data"] < 100]
# compute average
deep_sleep_avg_power = deep_sleep_data["Data"].mean()
# print average
print("Average power consumption deep sleep: ", deep_sleep_avg_power)

```

We obtained an average consumption in deep sleep of 59.66 mW.

## Power consumption in idle state

The power consumption in idle state can be extracted both by `sensor_read.csv` and `transmission_power.csv`. We used both datasets, merging all values and computing the average. We considered only values between 200 mW and 500 mW, that refer to idle mode.

```

import pandas as pd
# read CSV file
dataset_deep_sleep = pd.read_csv("deep_sleep.csv", parse_dates=['
                                Timestamp'])
dataset_sensor_reading = pd.read_csv("sensor_read.csv", parse_dates=['
                                Timestamp'])
# filter data with power between 200 mW and 500 mW, referring to idle
                                mode
idle_data_deep_sleep = dataset_deep_sleep[(dataset_deep_sleep['Data'] >=
                                200) & (dataset_deep_sleep['Data']
                                <= 500)]
# filter data with power <= 400 mW, reffering to idle mode
idle_data_sensor_reading = dataset_sensor_reading[dataset_sensor_reading
                                ['Data'] <= 400]
# merge values
idle_merged = pd.concat([idle_data_deep_sleep, idle_data_sensor_reading]
                        , ignore_index=True)
# compute average
idle_avg_power = idle_merged['Data'].mean()
# print average
print("Average power consumption idle: ", idle_avg_power)

```

We obtained an average consumption in idle of 322.62 mW.

## Power consumption in measurement state

Data representing power consumption in measurement state is contained in the dataset `sensor_read.csv`. We computed the average value considering values of power above 400 mW and obtained an average value of 465.18 mW. The algorithm is very similar to deep

sleep mode, thus we don't report it.

### Power consumption in WiFi on state

We estimated the power consumption when the WiFi is turned on and the ESP32 is not transmitting based on `deep_sleep.csv` and `transmission_power.csv`, merging all value between 600 mW and 750 mW, as shown for the idle mode. The average power consumption when WiFi is turned on is 724.58 mW.

### Power consumption when transmitting at 2 dBm

Finally, we estimated the average power consumption when ESP32 is transmitting at 2 dBm, based on `transmission_power.csv`. We compute the average of values between 750 mW and 900 mW, obtaining an average value of 797.29 mW.

### Summary power consumption

The summary of power consumption of ESP32 in different states is reported below.

State	Power
$P_{idle}$	322.62 mW
$P_{measurement}$	465.18 mW
$P_{WiFi}$	724.58 mW
$P_{transmission}$	797.29 mW
$P_{deep\_sleep}$	59.66 mW

Table 2.1: States power table

## 2.2. States duration estimation

We estimated the duration of different states by simulating the ESP32 on Wokwi and measuring them. We saved the timestamps in different points of execution, computed the duration of each state and printed it on the serial log.

We simulated the execution for about a hundred cycles and saved the serial log. We provide few examples of entries of the serial log, in which the durations are in microseconds.

```
entry 0x400805dc
Idle duration: 882
```

```
Measurement duration: 24116
Sending duration: 64
WiFi duration: 189158
ets Jul 29 2019 12:21:46
```

```
entry 0x400805dc
Idle duration: 846
Measurement duration: 24212
Sending duration: 65
WiFi duration: 188548
ets Jul 29 2019 12:21:46
```

```
entry 0x400805dc
Idle duration: 845
Measurement duration: 24211
Sending duration: 64
WiFi duration: 187548
ets Jul 29 2019 12:21:46
```

We computed the duration of different states using a Python algorithm to extract the values from the log, using regexes, filtering outliers and computing the average.

In the measurement of the sending duration, during which the ESP32 uses WiFi, we noticed the presence of some values completely different from the others, caused by the simulation on Wokwi; we filtered them out to get a more accurate estimation.

Finally, we observed that the duration of the measurement with the HC-SR04 ultrasonic distance sensor varies based on the resulting distance: the higher the distance, the longer the duration. In order to have a realistic duration, we gathered data with different values of measured distance.

```
import re
import statistics
# open log file
with open("time_measurements_log.txt", "r") as file:
    log_text = file.read()
# extract values from log using a regex
idle_values = list(map(int, re.findall(r"Idle duration:\s*(\d+)",
                                       log_text)))
measurement_values = list(map(int, re.findall(r"Measurement duration:\s*
                                              *(\d+)", log_text)))
sending_values = list(map(int, re.findall(r"Sending duration:\s*(\d+)",
```

```

log_text)))
wifi_values = list(map(int, re.findall(r"WiFi duration:\s*(\d+)",
log_text)))
# filter wrong values, given by the simulation
wifi_values = [value for value in wifi_values if value <= 200000]
# compute averages
avg_idle = statistics.mean(idle_values) if idle_values else 0
avg_measurement = statistics.mean(measurement_values) if
measurement_values else 0
avg_sending = statistics.mean(sending_values) if sending_values else 0
avg_wifi = statistics.mean(wifi_values) if wifi_values else 0
#print values
print("Average Idle duration:", avg_idle)
print("Average Measurement duration:", avg_measurement)
print("Average Sending duration:", avg_sending)
print("Average WiFi duration:", avg_wifi)

```

The duration of different states is reported in the following table.

State	Duration
$T_{idle}$	833.70 $\mu s$
$T_{measurement}$	18647.79 $\mu s$
$T_{WiFi}$	188640.86 $\mu s$
$T_{transmission}$	59.60 $\mu s$
$T_{deep\_sleep}$	$48 \times 10^6 \mu s$

Table 2.2: States duration table

## 2.3. Energy consumption estimation

Battery size 18593 J

Average power consumption Average power consumption deep sleep: 59.66093555093555 mW => 59.66 mW Average power consumption idle: 322.62464743589743 mW => 322.62 mW Average power consumption measurement: 465.18097744360904 mW => 465.18 mW Average power consumption WiFi on: 724.5793571428571 mW => 724.58 mW Average power consumption transmission at 2dBm: 797.2942857142858 mW => 797.29 mW

$$E_{idle} = P_{idle} \cdot T_{idle} = 322.62 \text{ mW} \cdot 833.70 \text{ } \hat{\text{A}}\text{tS} = 2.69 \times 10^{-4} \text{ J}$$

$$E_{measurement} = P_{measurement} \cdot T_{measurement} = 465.18 \text{ mW} \cdot 18647.79 \text{ } \hat{\text{A}}\text{tS} = 8.67 \times 10^{-3} \text{ J}$$

$$E_{WiFi} = P_{WiFi} \cdot (T_{WiFi} - T_{transmission}) = 724.58 \text{ mW} \cdot (188640.86 \text{ } \hat{\text{A}}\text{tS} - 59.60 \text{ } \hat{\text{A}}\text{tS}) = 0.136642 \text{ J}$$

$$E_{transmission} = P_{transmission} \cdot T_{transmission} = 797.29 \text{ mW} \cdot 59.60 \text{ } \hat{\text{A}}\text{tS} = 4.7518484 \times 10^{-5} \text{ J}$$

$$E_{deepsleep} = P_{deep\_sleep} \cdot T_{deep\_sleep} = 59.66 \text{ mW} \cdot 48 \times 10^6 \text{ } \hat{\text{A}}\text{tS} = 2.86386 \text{ J}$$

$$E = E_{idle} + E_{measurement} + E_{WiFi} + E_{transmission} + E_{deep\_sleep} = 3.00931 \text{ J}$$

6.87168 7.65929 8.15501





## 3 | Improvements



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