

Implementation and Evaluation of mobile RSSI-based LoRaTM Localization

Bachelor Thesis for B.sc. Informatik

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Motivation



- Apple AirTag™

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- IoT + Industry 4.0

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- Rise of LoRa™

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- Rise of LoRa™
- GPS is insufficient
 - Mountains
 - Underwater
 - Indoor

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- “**Long Range**”
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- Chirp Spread Spectrum Modulation (CSS)
- Low-power (TX: ~20 mA, RX: ~11 mA) [2]

- LoRa™ EU $f_{\text{center}} = 433.05 - 434.79 \text{ MHz}$ and $863 - 870 \text{ MHz}$

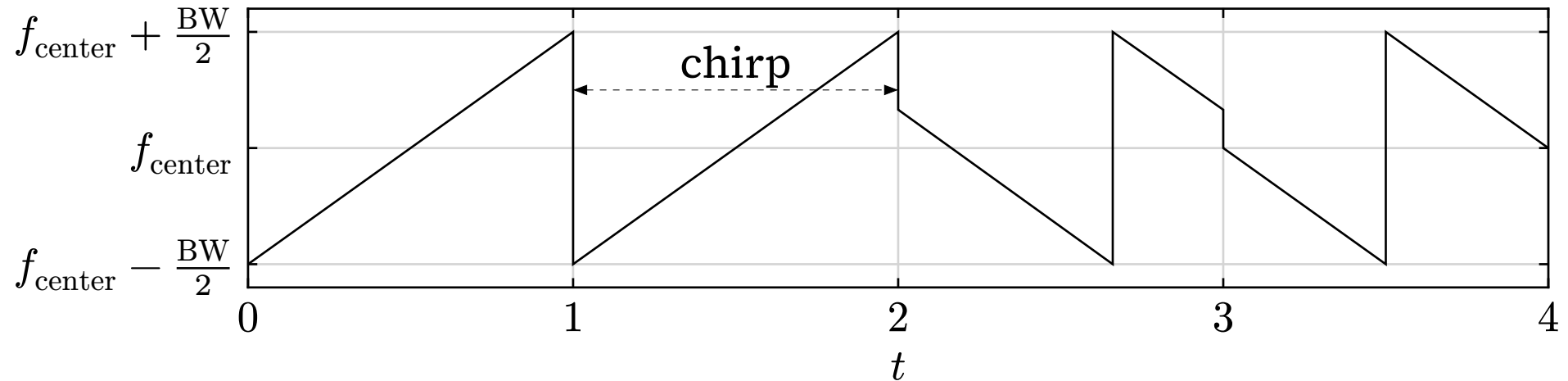


Figure 1: Chirp Spread Spectrum Modulation

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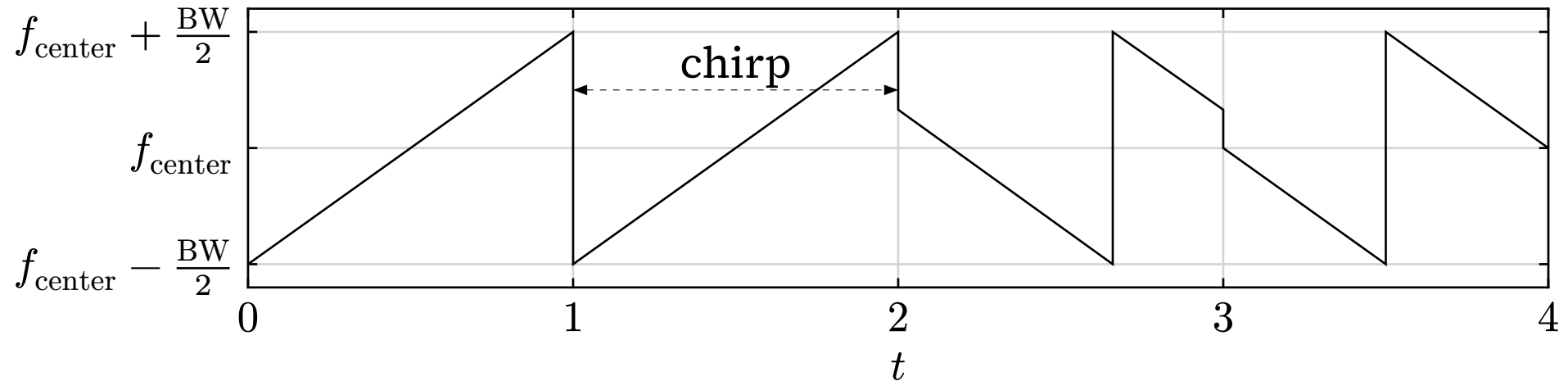


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- $2^{\text{Spreading Factor}}$ symbols per chirp (7-12) [3]

Goal of this Thesis

- Evaluate feasibility of a RSSI-based localization system with LoRaTM

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- Implement a simple reference system for evaluation

Implementation

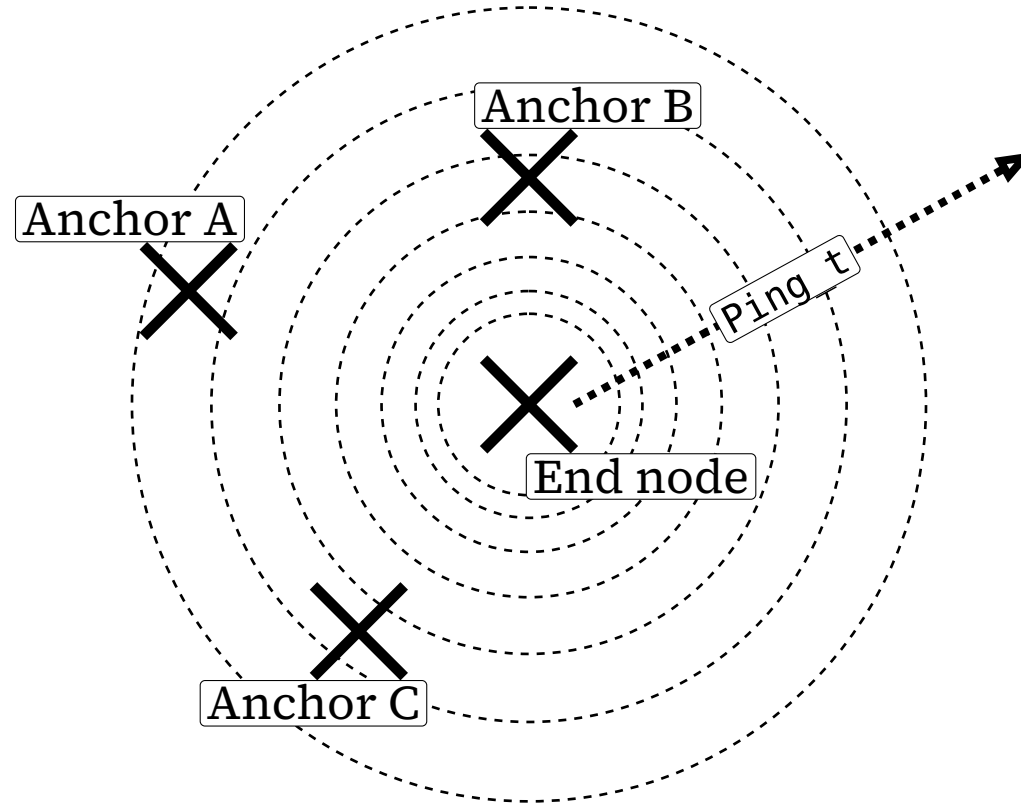


Figure 2: Propagation of 'Ping_t' message



Figure 3: NUCLEO-WL55JC

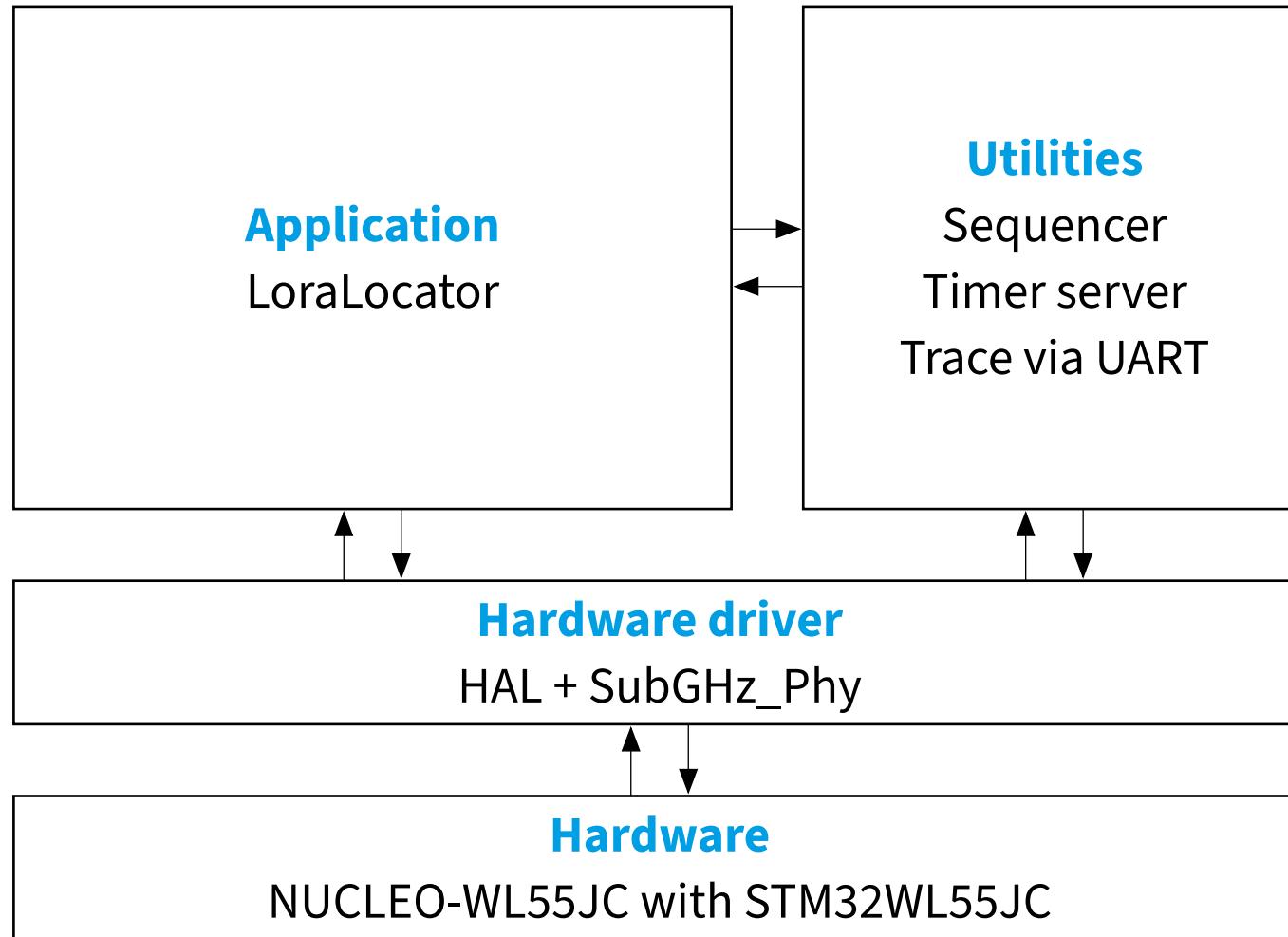


Figure 4: Firmware architecture

- Path-Loss Model [4]

$$\text{RSSI}(d) = P_0 - 10n \cdot \log_{10} \left(\frac{d}{d_0} \right) \quad (1)$$

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- find coefficients with linear regression

$$y = ax + b \quad (3)$$

$$y = \text{RSSI}(d), \quad a = -10n, \quad x = \log_{10} \left(\frac{d}{d_0} \right), \quad b = P_0 \quad (4)$$

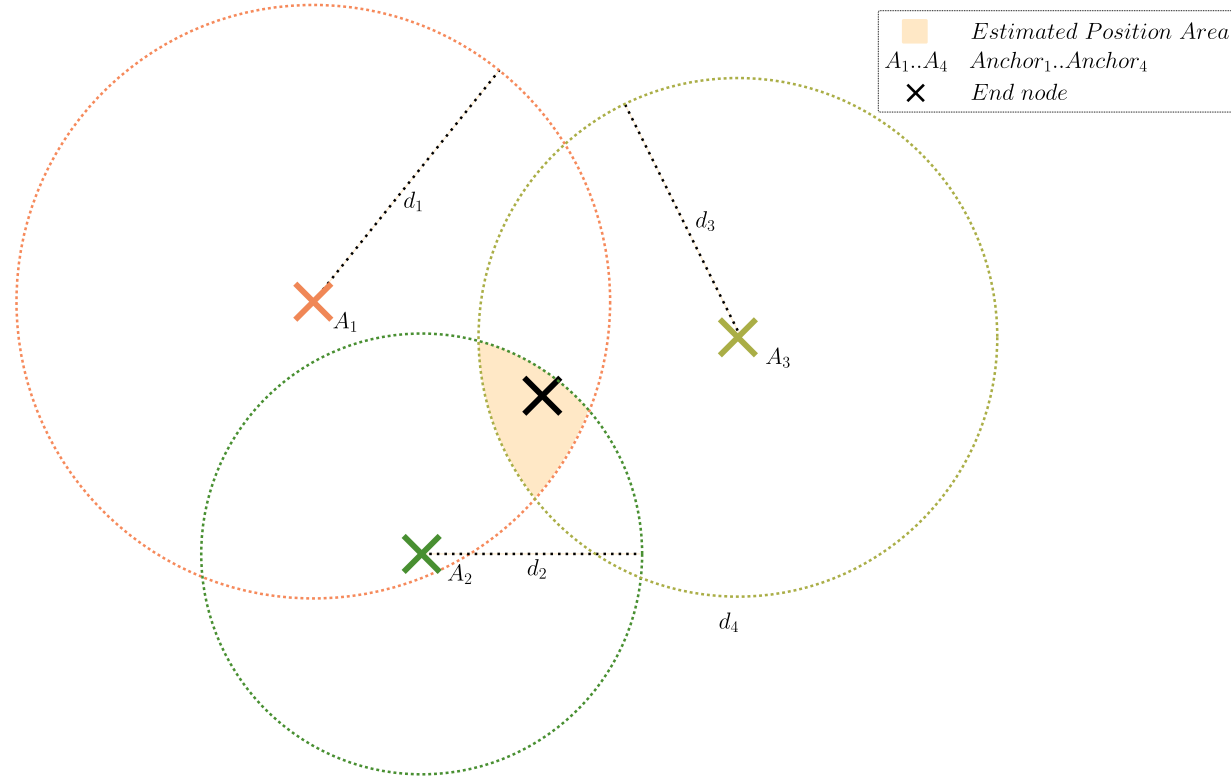


Figure 4: Position estimation error with trilateration

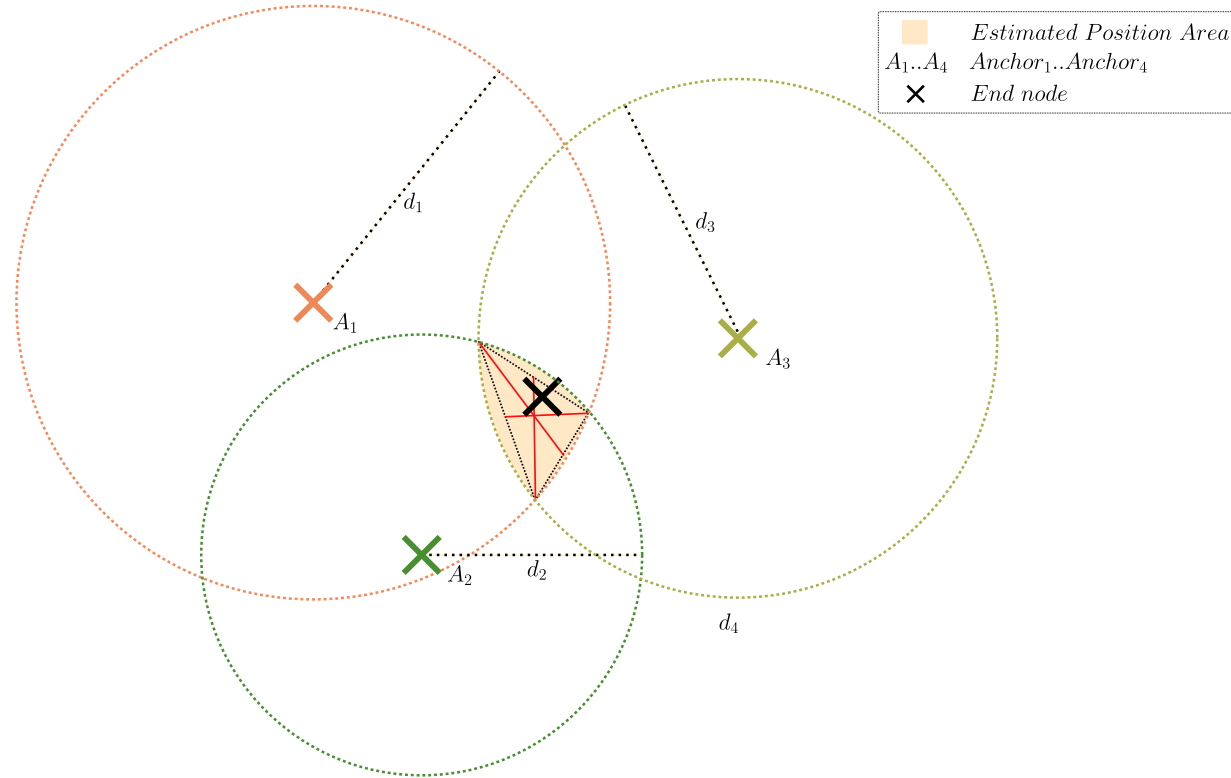


Figure 5: Modified trilateration

Evaluation



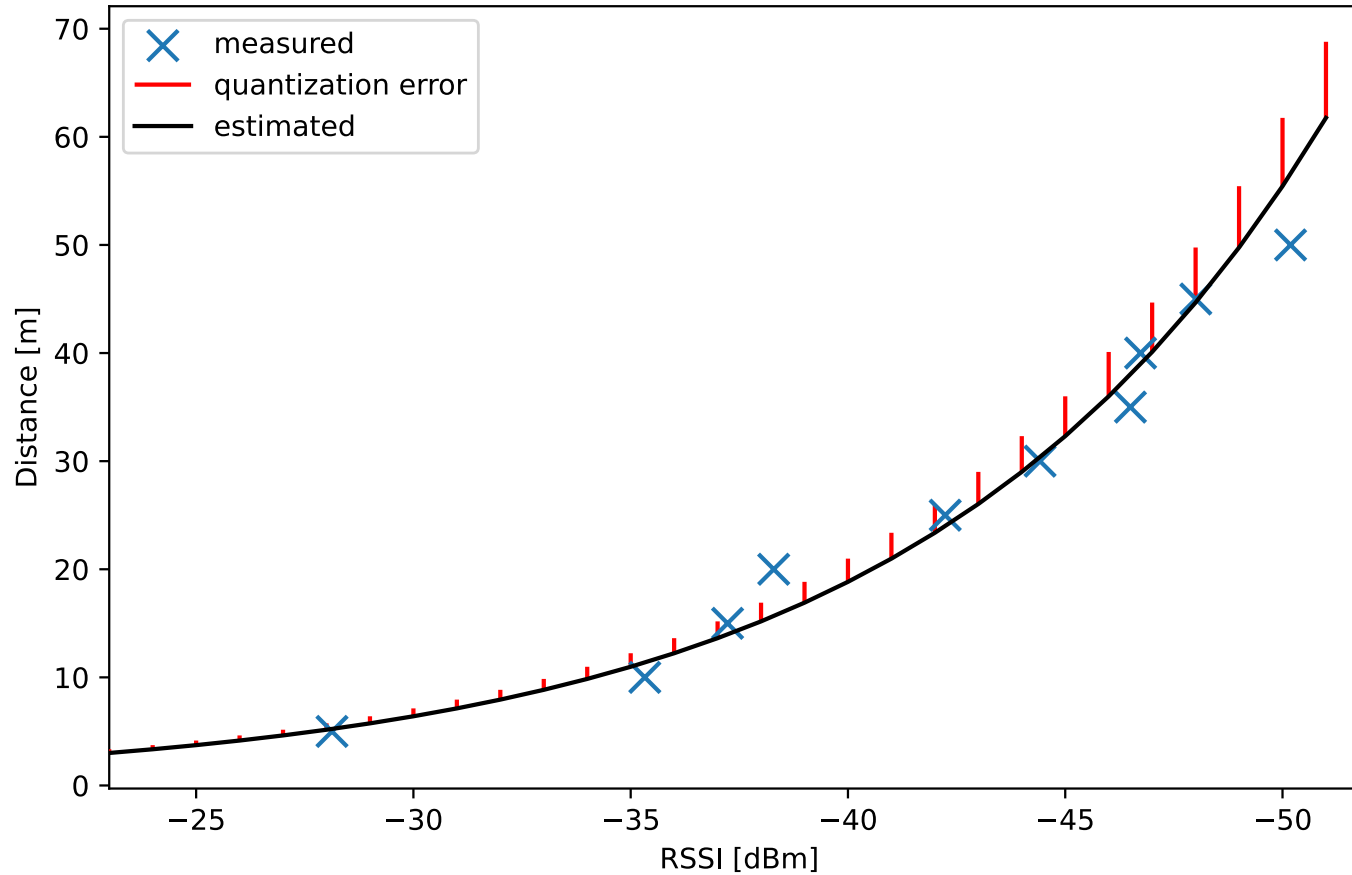


Figure 6: RSSI vs. Distance data from experiment 02_1

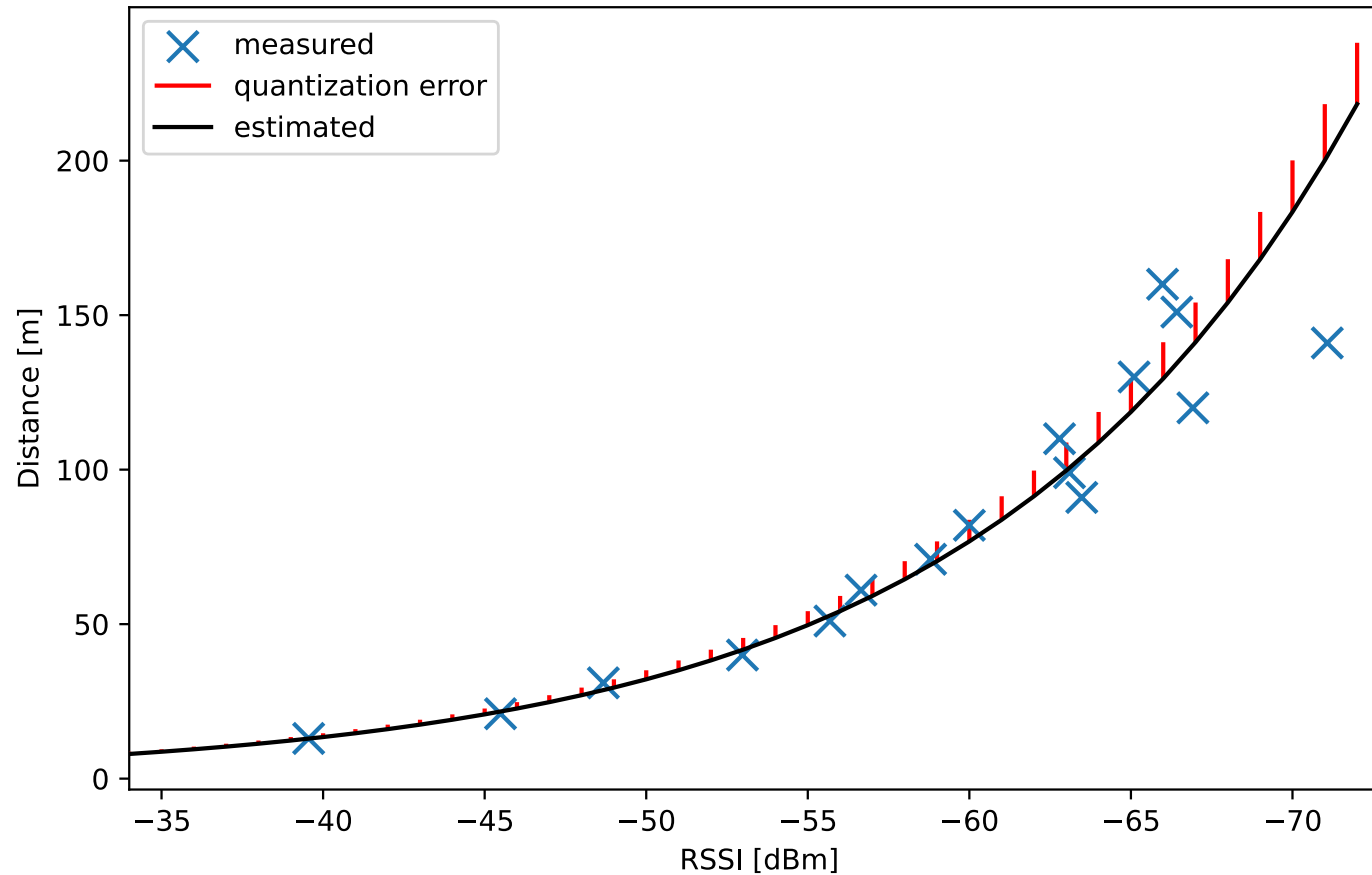


Figure 7: RSSI vs. Distance data from experiment 07

ID	Range	n	P_0	$\mu(e_r)$	$\sigma(e_r)$	e_{\max}
01_2	5 m – 50 m	2.0873	−15.8050	0.0183	±0.1967	+11.0087 m
02_1	5 m – 50 m	2.1331	−12.8029	0.0052	±0.1051	+6.5532 m
02_2	5 m – 50 m	1.7558	−17.6398	0.0323	±0.2710	+19.4399 m
07	10 m – 160 m	2.6444	−10.1465	0.0088	±0.1454	+60.4273 m

Table 1: Experiments for distance estimation evaluation

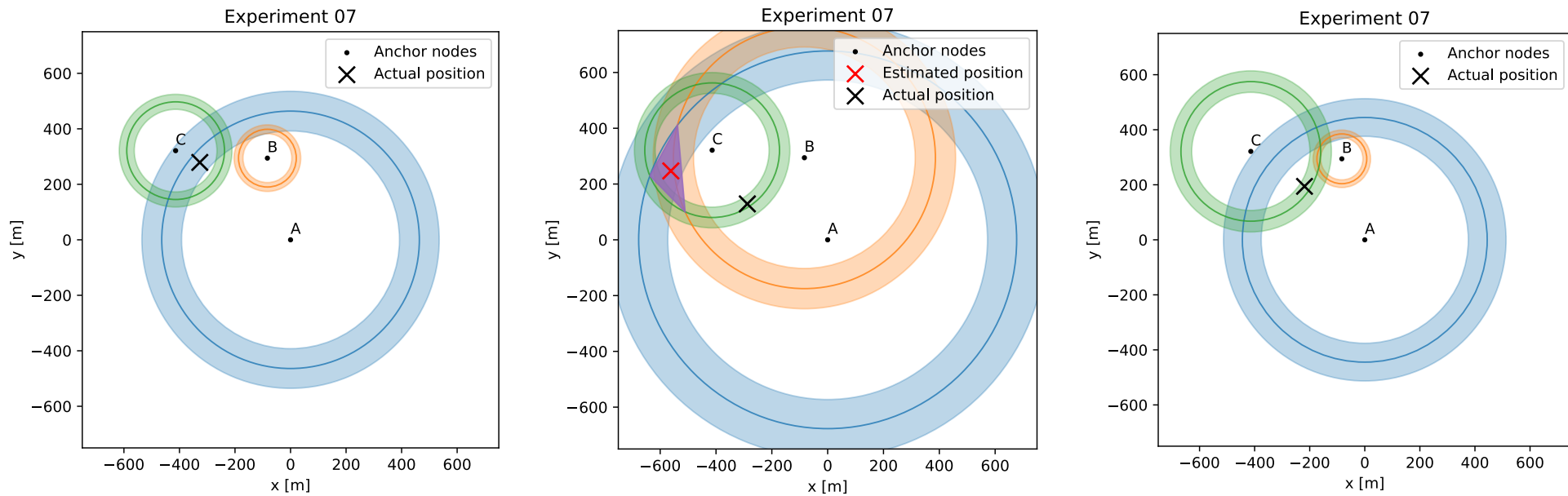


Figure 8: Localization – Experiment 07

Distance model	Position	Anchor ID	RSSI [dBm]	Distance [m]	Estimated Position	Actual Position
07	pos2	A	−85.0000	677.0604 ± 104.4287	$x: -561.83$ $y: 246.99$	$x: -287.79$ $y: 128.70$
		B	−80.7910	469.3139 ± 72.3862		
		C	−73.15	241.30 ± 37.22		

Table 2: Experiments for localization evaluation

Conclusion

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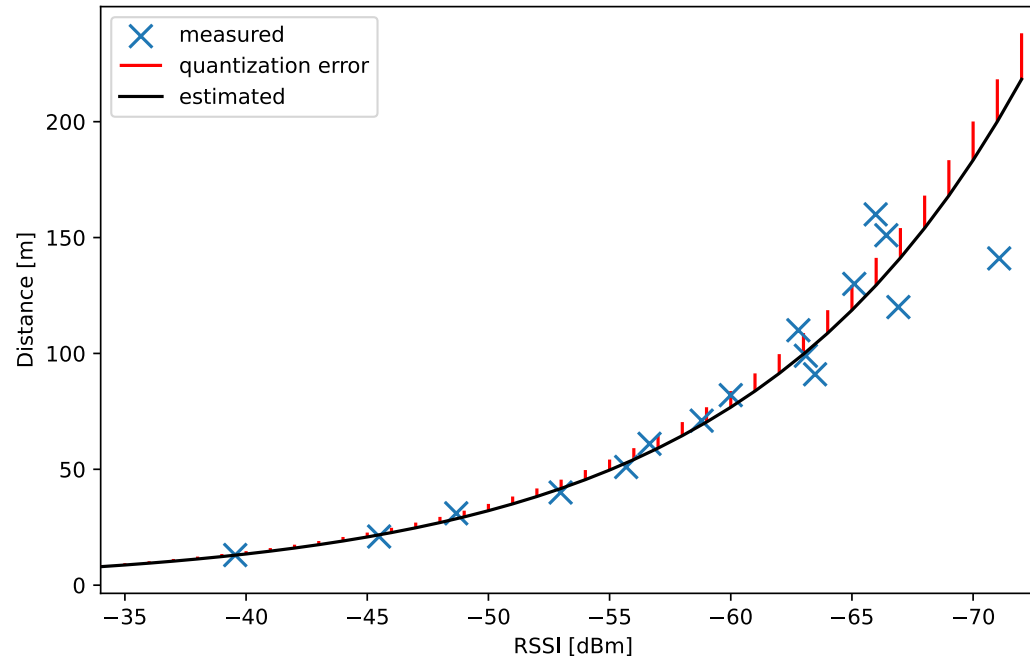


Figure 1: RSSI vs. Distance data from experiment 07

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→ potential for future improvements is visible

- different localization systems with LoRaTM already exist
 - RSSI-based [5], [6]
 - timing-based [7]

Future Work



- improve RSSI to distance estimation

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 - integrate GPS receiver into hardware for automated data acquisition

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 - integrate GPS receiver into hardware for automated data acquisition
 - use more than 4 devices for model fitting
 - analyze the cause for the RSSI variation at higher distances
 - try different Path-Loss model
- try different localization algorithm
 - weighted least-mean-square method [5], [8]
 - DV-Hop [9]

Thank you

- firmware available on GitHub:
<https://github.com/moseschmiedel/lora-locator>
- data available on GitHub:
<https://github.com/moseschmiedel/bachelor-thesis>

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- [8] R. Sumathi and R. Srinivasan, “RSS-based location estimation in mobility assisted wireless sensor networks,” in *Proceedings of the 6th IEEE International Conference on Intelligent Data Acquisition and Advanced Computing Systems*, Sep. 2011, pp. 848–852. doi: [10.1109/IDAACS.2011.6072891](https://doi.org/10.1109/IDAACS.2011.6072891).
- [9] H. Safa, “A novel localization algorithm for large scale wireless sensor networks,” *Computer Communications*, vol. 45, pp. 32–46, Jun. 2014, doi: [10.1016/j.comcom.2014.03.020](https://doi.org/10.1016/j.comcom.2014.03.020).

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Figure 1: LiPo Battery 35

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```
1 typedef struct {
2     // attribute used to discriminate
3     // between packet types
4     PacketType_t packet_type;
5     // ID of the device
6     // sending the `Ping_t`
7     uint8_t device_id;
8     // ID of the `Ping_t`, `device_id`
9     // combined with this should be unique
10    uint8_t packet_id;
11 } Ping_t;
```

Listing 1: Packet type `Ping_t`

```
1 // Note that this packet does not need a `packet_type` discriminator
  because
2 // it is the only type that is 4 bytes long.
3 typedef struct {
4     // ID of the anchor sending the `AnchorResponse_t`
5     Device_t anchor_id;
6     // ID of the `Ping_t` that triggered this `AnchorResponse_t`
7     uint8_t packet_id;
8     // RSSI of `Ping_t` measured by the anchor node
9     int16_t recv_rssi;
10 } AnchorResponse_t;
```

Listing 2: Packet type `AnchorResponse_t`

```
1 typedef struct {  
2     // attribute used to discriminate between packet types  
3     PacketType_t packet_type;  
4     // ID of the anchor this `Ack_t` is addressed to  
5     Device_t receiver_id;  
6     // ID of the `Ping_t` that triggered the communication  
7     uint8_t packet_id;  
8 } Ack_t;
```

Listing 3: Packet type `Ack_t`

Mode	Current [mA]	Power consumption [mW]
Transmitting	27	135
Receiving	8.5	42.5
Sleep	2.9	14.5

Table 1: Power consumption

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Table 1: Power consumption

$$P_{\text{tx}} = \frac{2 \cdot T_{\text{tx}}}{1\,000\text{ ms}} \cdot 135\text{ mW} \quad (1)$$

$$P_{\text{rx}} = \frac{400\text{ ms}}{1\,000\text{ ms}} \cdot 42.5\text{ mW} \quad (2)$$

$$P_{\text{tx}} = \frac{1000 - (2 \cdot T_{\text{tx}} + 400\text{ ms})}{1\,000\text{ ms}} \cdot 14.5\text{ mW} \quad (3)$$

$$P_{\text{total}} = P_{\text{tx}} + P_{\text{rx}} + P_{\text{sleep}} = 30.7\text{ mW} \quad (4)$$

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Figure 1: LiPo Battery

- lifetime of ~1303 h or 54.3 days with battery capacity of 40000 mWh

$$P_r^{[\text{dB}]} = P_t^{[\text{dB}]} + G_t^{[\text{dBi}]} + G_r^{[\text{dBi}]} + 20 \log_{10} \left(\frac{\lambda}{4\pi d} \right) \quad (6)$$