# Indoors object locating with RSSI

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# Why to locate objects

Sailors all over the time have tried to locate themself using stars

It would be nice to know where valuable items were

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# Missing F-35: US military asks for public's help to find jet

18 September 2023

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#### Megan Fisher

BBC News



Watch: This is what the missing F-35 US military jet looks like

https://www.bbc.com/news/world-us-canada-66841194

US military might start to locating

their planes

The US military has asked for the public's help to locate one of its \$100m (£80m) F-35B fighter jets after the pilot ejected from the aircraft.

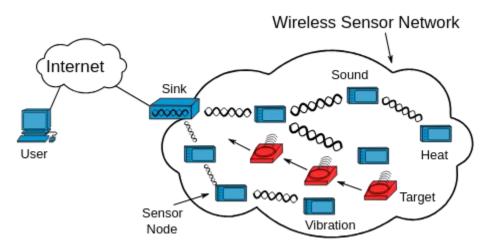
# Wireless sensor network (WSN)

Network of *nodes* 

Sink is a more powerful computer used to connect the WSN to the internet

#### Use cases

- Forest fire detecting
- Earthquake detecting
- Research



https://commons.wikimedia.org/wiki/File:Wireless\_Sensor\_Network General Structure.svg

## **A** node

Small -- About size of a coin

#### Required component

- Central Processing Unit (CPU)
- Communicating unit
- Battery

#### Measurement nodes (Node)

- Lot of sensors
- Unknown position

#### Anchor nodes (Anchors)

- Specialize on measuring own position
- Specialized locating chip
- Mounted on known location

# Time of Arrival (ToA)

Known speed of radio waves

Two very accurate clocks

CPU has a built in clock, but it is very inaccurate

External clock is required

$$d = v \cdot t$$

d is distance

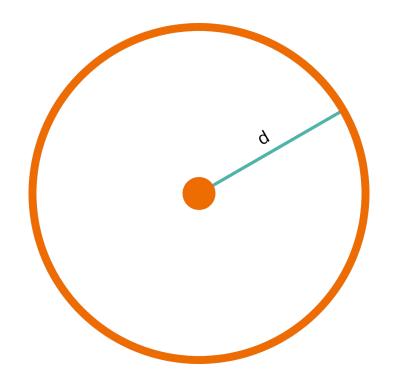
*v* is velocity of the radio wave

*t* is the time it took

# Received Signal Strength Indicator (RSSI)

One can calculate the distance between the node and anchor

Strength is inverse proportionality to the distances square



# Friis equation

$$\frac{P_r}{P_t} = G_t G_r \left(\frac{\lambda}{4\pi d}\right)^2$$

 $P_r$  is received power

 $P_{t}$  is transmitted power

 $G_{t}$  is transmitter gain

 $G_r$  is receiver gain

 $\lambda$  is wavelength of signal

d is distance between the nodes

#### Usually power are given in decibels

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

#### Solve the distance

$$d = \frac{\lambda}{4\pi} 10^{-\frac{P_r^{[dB]} - P_t^{[dB]} - G_t^{[dB]} - G_r^{[dB]}}{20}}$$

## **Trilateration**

#### Three points

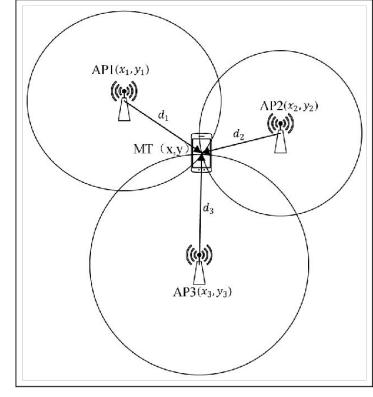
Compute the intersecting point

$$x = \frac{AY_{32} + BY_{13} + CY_{21}}{2(x_1Y_{32} + x_2Y_{13} + x_3Y_{21})}, \quad y = \frac{AX_{32} + BX_{13} + CX_{21}}{2(y_1X_{32} + y_2X_{13} + y_3X_{21})}$$

$$A = x_1^2 + y_1^2 - d_1^2, \quad B = x_2^2 + y_2^2 - d_2^2, \quad C = x_3^2 + y_3^2 - d_3^2$$

$$X_{32} = (x_3 - x_2), \quad X_{13} = (x_1 - x_3), \quad X_{21} = (x_2 - x_1),$$

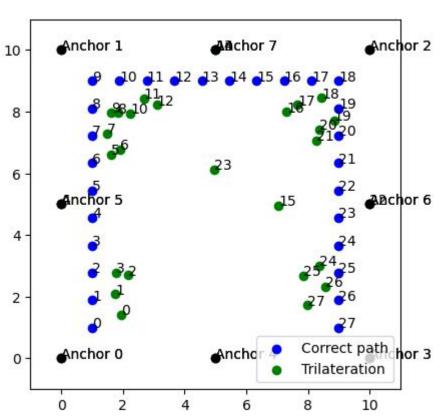
$$Y_{32} = (y_3 - y_2), \quad Y_{13} = (y_1 - y_3), \quad Y_{21} = (y_2 - y_1).$$



https://www.researchgate.net/figure/ll ustration-of-trilateration\_fig1\_3240615 72

## **Point estimation with Trilateration -- 8 nodes**

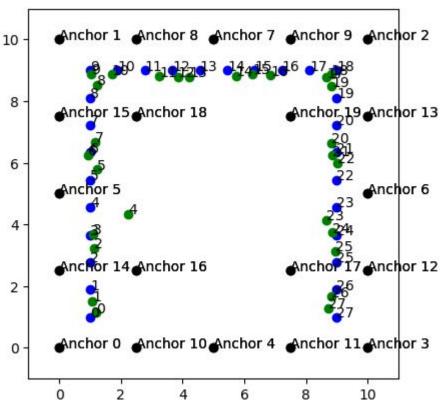
Root mean square error 1.474132706



## Add more anchor nodes -- 20 nodes

Root mean square error 0.442750141

More anchors -> Less nodes



## Kalman filter

R. Kalman published a revolutionary paper in 1960

Recursive solution to a discrete-time linear optimization problem

Recursive -> No huge memory consumption

Works in two part

First part is the so called prediction part

Second part is the so called update part

## **Kalman filter -- Prediction**

The Prediction step, predicts a new state based on the state model

$$\bar{X}_k = A\hat{X}_{k-1}$$

$$P_k^- = AP_{k-1}A^T + Q_k$$

A is state state-transition matrix  $ar{X}_k$  is the new state prediction  $ar{X}_{k-1}$  is the last updated state  $P_{\iota}^{-}$  is temporary covariance matrix  $P_{k-1}$  is covariance matrix  $Q_k$  is process noise matrix

## Kalman filter -- Update

The Update step, updates the new state based on the data available

$$K_{k} = P_{k}^{-}H^{T} (HP_{k}^{-}H^{T} + R)^{-1}$$

$$\hat{X}_{k} = \bar{X}_{k} + K_{k} (z_{k} - H\bar{X}_{k})$$

$$P_{k} = (I - K_{k}H)P_{k}^{-}$$

 $K_k$  is kalman gain Measurement model matrix  $X_k$  is the state estimation R measurement noise matrix  $P_k$  is covariance matrix  $|\mathcal{Z}_k|$  is the measurement matrix

### Kalman filter -- In our case

$$A = \begin{bmatrix} 1 & 0 & T_s & 0 \\ 0 & 1 & 0 & T_s \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$X = \begin{bmatrix} x & y & \dot{x} & \dot{y} \end{bmatrix}^T$$

$$z = \begin{bmatrix} x & y \end{bmatrix}^T$$

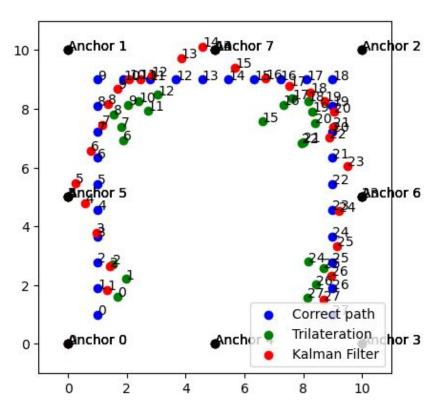
$$H = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}$$

$$R = \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix}$$

## Point estimation with Kalman filter -- 8 nodes

Root mean square error without kalman filter 1.042912802

Root mean square error with kalman filter 0.8047597031

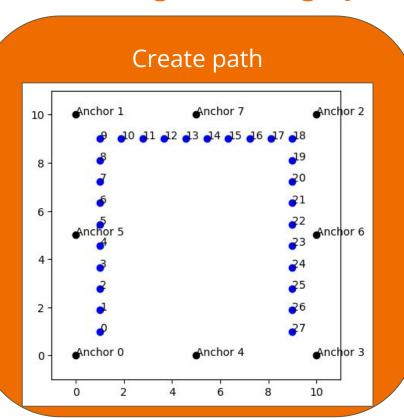


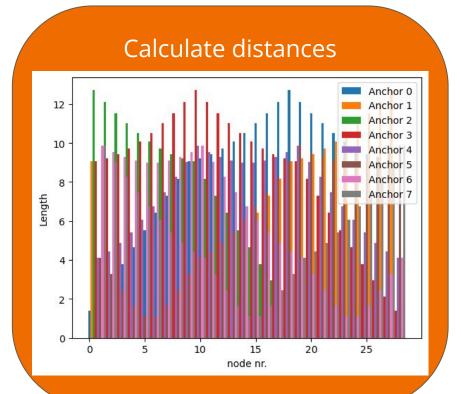
## **How produce data**

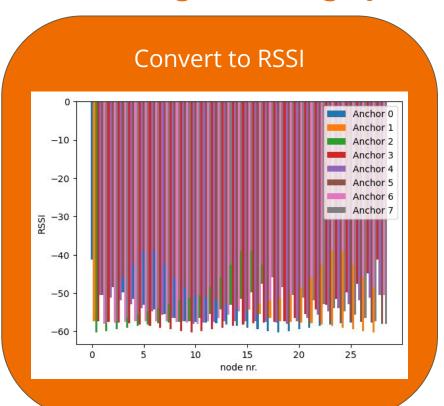
Hopefully a whole testing lab -- unfortunately not

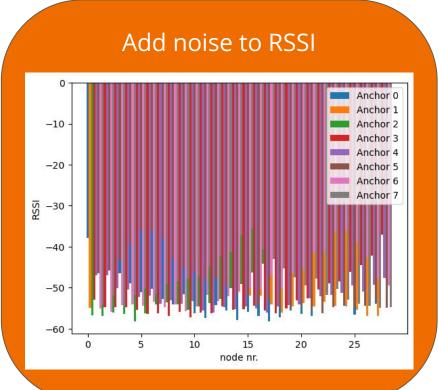
Needed to produce it synthetically.

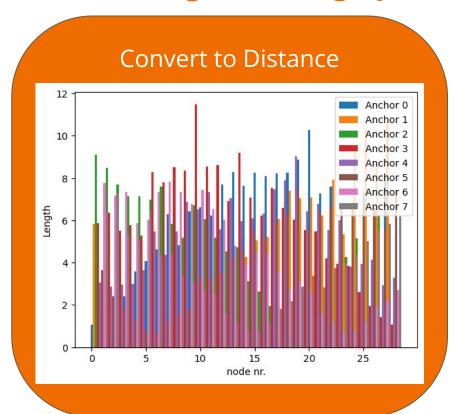
Five step process

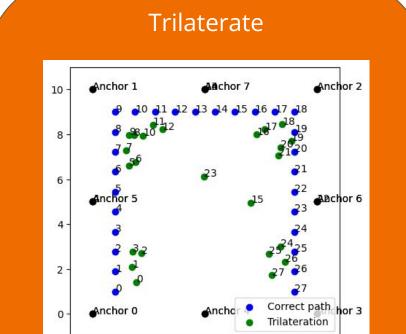


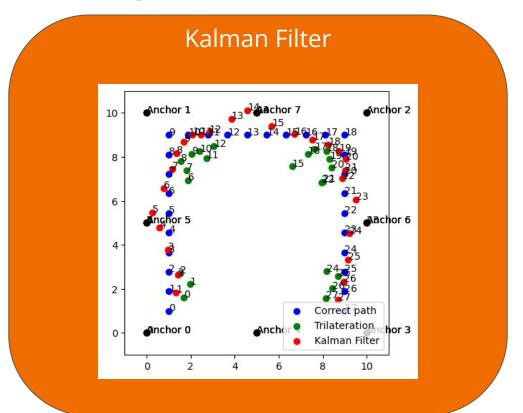












## **Conclusion**

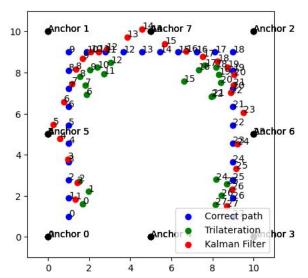
It is possible to locate and track objects indoors with RSSI.

Kalman filter can be utilized to improve measurements

Adding more anchors improves the accuracy on the cost of nodes

Root mean square error without kalman filter 1.042912802

Root mean square error with kalman filter 0.8047597031



# **Questions and Sources**

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