

# Wireless IoT technology

## ① LoRa/LoRaWAN

Range: {urban: 2~5  
(km) rural: 15

Speed: 50 kbps

power consumption: •••

Application: Precision farming, manufacturing automation, pipeline monitoring

## ② Sigfox

Range: {urban: 3~10  
(km) rural: 30~50

Speed: 300bps

power consumption: •

Application: Predictive maintenance, capacity planning, demand forecasting

## ③ NB-IoT

Range: {urban: 1~5  
(km) rural: 10~15

Speed: 250 kbps

power consumption: •

Application: Electric metering, manufacturing automation, retail PoS, Emergency Command Center, GIS monitor center, DMA Pipe Network

## ④ Zigbee

{ Zigbee Router  
Zigbee Coordinator  
Zigbee Device

Zigbee Device Objectives (ZDO)

provides the interface between application objects, device profiles, and APS layer in Zigbee devices.

## ⑤ MQTT: Message Queuing Telemetry Transport

Features: Publish and subscribe to topics

3 quality of service: { 0 Best effort to deliver a msg  
1 Deliver at least once  
2 Deliver exactly once

Application: Home Automation, Health care, Mobile phone apps, Industrial automation, Automotive

## Edge Computing

Benefits and use cases

- Latency - extremely low latency is essential for near real-time control
- Bandwidth - limited bandwidth
- Connectivity - reduce cloud connectivity costs
- Keeps data private by doing more computations locally
- Time sensitive compute-intensive workloads

## Energy Models

k = no. of bits

E<sub>elec</sub>: 50 nJ/bit E<sub>amp</sub>: 100 pJ/bit/m<sup>2</sup> energy loss due to channel gain

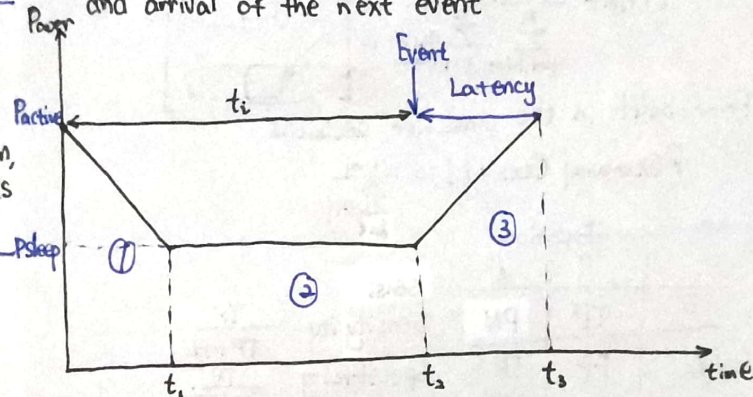
Transmitter  $\Rightarrow E_{Tx}(k,d) = E_{elec} * k + [E_{amp} * k * d^2]$   
Receiver  $\Rightarrow E_{Rx}(k,d) = E_{elec} * k$

## • Start Up Energy

Energy consumption dominated by start-up transient when packet size is small.

## • State transitions

t<sub>i</sub> = time between end of processing of previous event and arrival of the next event

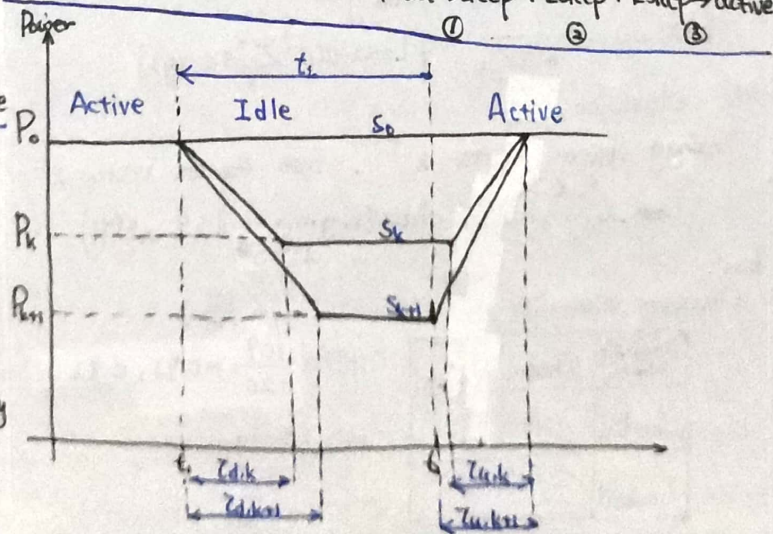


$$① \text{ Exact} \rightarrow \text{sleep} = (P_{\text{active}} + P_{\text{sleep}}) * \frac{t_1}{2}$$

$$② \text{ Esleep} = P_{\text{sleep}} * \frac{t_2 - t_1}{2}$$

$$③ \text{ Esleep} \rightarrow \text{active} = (P_{\text{active}} + P_{\text{sleep}}) * \frac{t_3 - t_2}{2}$$

∴ sleep if  $P_{\text{active}} * t_1 > \text{Exact} \rightarrow \text{sleep} + \text{Esleep} + \text{Esleep} \rightarrow \text{active}$



## • Energy Savings E<sub>s</sub>

$$E_{s,k} = (P_0 - P_k) t_i - \left[ \frac{P_0 - P_k}{2} \right] t_{i,k} - \left[ \frac{P_0 + P_k}{2} \right] t_{u,k}$$

Transition is only justified when  $E_{s,k} > 0$

## • Transition time threshold

$$T_{th,k} = \frac{1}{2} \left[ t_{i,k} + \left( \frac{P_0 + P_k}{P_0 - P_k} \right) t_{u,k} \right]$$

Worthwhile going to mode k if  $t_i > T_{th,k} \Rightarrow S_0 \rightarrow S_k$

States: Active(s<sub>a</sub>), Ready(s<sub>r</sub>), Monitor(s<sub>m</sub>), Look(s<sub>l</sub>), Sleep(s<sub>s</sub>)

## WSN Routing

• Minimum Transmission Energy (MTE) routing: neighbour selection.

For 3-nodes A, B, C, A would transmit to node C through B iff  $ETX(d=d_{AB}) + ETX(d=d_{BC}) < ETX(d=d_{AC})$

## LEACH

- The job of cluster head is to collect data from their surrounding nodes and pass it on to the base station.
- Cluster head role is rotated to share energy load
- Cluster head receives the data and performs signal processing functions on the data and transmits data to the remote base station

The LEACH network has two phases

## ① Set-Up phase

- Cluster Heads (CH) are chosen.
- Nodes join clusters.

## ② The Steady-State phase

- The cluster heads and clusters are maintained.
- Data is transmitted from nodes to CHs.
- CHs may do some processing and data aggregation before sending the data to the base station (BS).
- Each sensor node elects itself to be cluster head with a certain prob at the beginning of each round.
- Node have not already been cluster heads may become CH



Decision to be cluster heads

$$\text{Threshold : } T(r) = \frac{P}{1 - P \times (r \bmod P^{-1})} \quad \forall \text{ nodes } \in G$$

$$T(r) = 0 \quad \forall \text{ nodes } \notin G$$

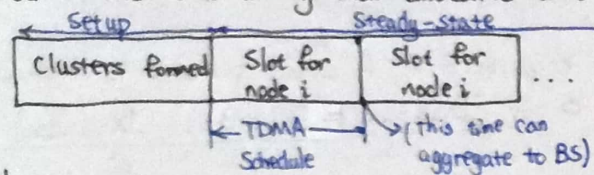
P is the cluster-head probability

G is the set of nodes that weren't CHs previous rounds

$$P = \frac{\text{no. of CH}}{\text{no. of nodes}} = \frac{k}{N}$$

n is the random number between 0 and 1, if  $n < T(r)$ , then that node become cluster head. Each node will be cluster head once in an interval of  $\frac{1}{P}$  rounds.

Nodes send data during their allocated time slot



Cluster members will send data according to DSSS code & a TDMA schedule to minimize inter-cluster interference

Advantages of LEACH

- ① LEACH helps to balance the load in the networks as compared to direct transmission
- ② LEACH is completely distributed, requiring no control information from the base station
- ③ Nodes do not need global topology information

Disadvantages of LEACH

- ① Nodes must have data to send in the allotted time
- ② Bad clusters can be formed due to the random nature of cluster information

Data fusion

Various forms of CSP

- Single Node, Multiple Modality (SN, MM)
  - No communication burden
- Multiple Node, Single Modality (MN, SN)
  - Higher communication burden

- Multiple Node, Multiple Modality (MN, MM)
  - Highest communication and computational burden

Classification.

For a  $m \times m$  confusion matrix

- Probability of detection for class m

$$PD_m = \frac{n_{mm}}{\sum_{j=1}^m n_{mj}} \quad m \begin{bmatrix} \text{ } \end{bmatrix}$$

- Probability of false alarm for class m

$$PFA_m = \frac{\sum_{k=1, k \neq m}^M \frac{n_{km}}{\sum_{j=1}^M n_{kj}}}{\sum_{k=1, k \neq m}^M \frac{n_{km}}{\sum_{j=1}^M n_{kj}}}$$

- Prior belief in the classifier decisions

$$P(x \in w_m | C(x) = j) = \frac{n_{mj}}{\sum_{i=1}^M n_{ij}}$$

Binary Classification

	$\hat{P}$	$\hat{N}$
P	TP	FN
N	FP	TN

$$\text{Precision} = \frac{TP}{TP+FP}$$

$$\text{Sensitivity} = \frac{TP}{TP+FN}$$

$$\text{Specificity} = \frac{TN}{TN+FP}$$

$$\text{Accuracy} = \frac{TP+TN}{TP+TN+FP+FN}$$

Gaussian Classifiers

$$\text{mean vector } \mu_j = E_j[x], \quad Z_j = E_j[(x - \mu_j)(x - \mu_j)^T]$$

Likelihood function for class j

$$P(x | w_j) = \frac{1}{\pi^N |Z_j|} \cdot \exp[-(x - \mu_j)^T Z_j^{-1} (x - \mu_j)]$$

NN classifier

Training feature vector  $x^{ctr}$ , test feature vector  $x^c$

$$CNN(x_1, \dots, x_k) = \text{class}(\arg \min_{x^{ctr} \in S^{Tr}} \|x^c - x^{ctr}\|)$$

Ex:

Gaussian classifier

	$w_m$	Wheeled	Tracked
120		109	11
120		22	98

$$PD = \frac{109}{120}, 0.82, \text{ avg} = 0.86$$

$$PFA = \frac{22}{120}, 0.09$$

Collaborative Signal & Information Processing (CSIP)

IDSQ: Information Driven Sensor Query

(Static case)  $\uparrow$  without need to have the sensor data first

Measurement/Observation model

$$Z_i(t) = h(x(t), \lambda_i(t)) \quad \textcircled{1} \lambda_i = [x_i, \sigma_i^2]^T$$

$$Z_i = \frac{a}{\|x_i - x\|^2} + w_i$$

a is target amplitude  
d is attenuation coefficient

Information Utility