

DNN-based Active User Detection for a NB-IoT Compatible Grant Free NOMA System

Progress Seminar

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Summary of previous work

- An algorithm based on 2D fast Fourier transform (FFT) was implemented for random access in narrow band Internet of things (NB-IoT).

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- The algorithm was initially simulated and tested in Matlab.
- In the next phase, Matlab code was translated into C code.

Introduction to grant free NOMA

Definition

Massive machine type communication (mMTC) refers to the connection of a **large number of devices** (10^6 devices per km^2 ^a) to the base station (BS).

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- In mMTC majority of the devices are
 - ▶ Low cost
 - ▶ Battery powered
 - ▶ Transmits low data sporadically.
- Main concern of mMTC is **massive connectivity** of the devices to the BS.

Introduction to grant free NOMA

Existing access method: Grant based OMA

- ▶ **Grant based access** : Each device requests a data transmission slot via a contention-based random access process.
- ▶ **Orthogonal multiple access (OMA)** : Available radio resources are allocated to devices in a non over lapping manner.

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Why grant based access OMA is not suitable for mMTC?

- ▶ Limited number of available radio resources.
- ▶ Signalling overhead.

Introduction to grant free NOMA

Solution : Grant free non-orthogonal multiple access

- ▶ **Grant free** access, allows devices to transmit data without the scheduling process.
- ▶ **NOMA** enables massive connectivity over limited radio resources^a.

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- In typical grant free system
 - ▶ Preamble and data are transmitted simultaneously.
 - ▶ Active user detection plays major role in successful data decoding.

Grant free NOMA system

- A typical uplink mMTC with one BS serving a total of M users is considered.

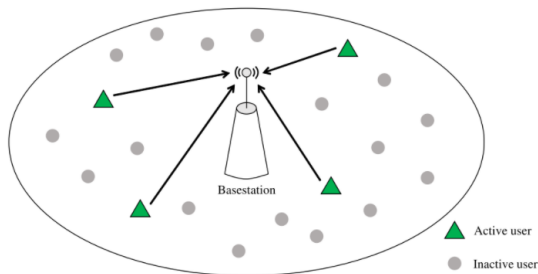


Figure: mMTC uplink scenario where only a few devices are active¹.

- Number of active users : Poisson with mean λ .

¹Wonjun Kim et al. "Deep neural network-based active user detection for grant-free NOMA systems". In: *IEEE Trans. Commun.* (2020).

Grant free NOMA system contd.

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Choosing number of preambles

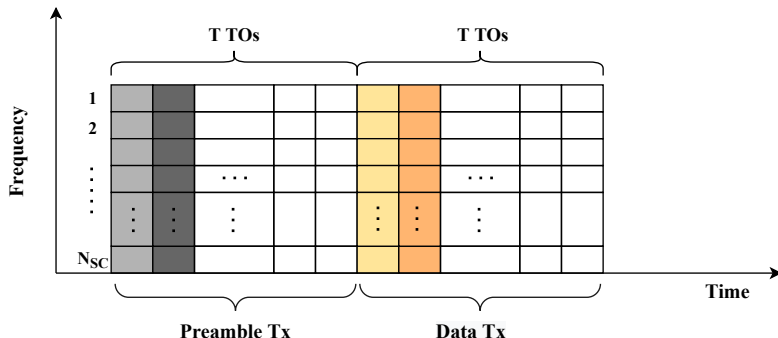
$$N_p = \left\lceil \frac{\lambda^2}{2\log(1/\mathbb{E}[p_{\text{all}}])} \right\rceil$$

N_p : Number of preambles,

$\mathbb{E}[p_{\text{all}}]$: Desired average success (no collision) probability^a.

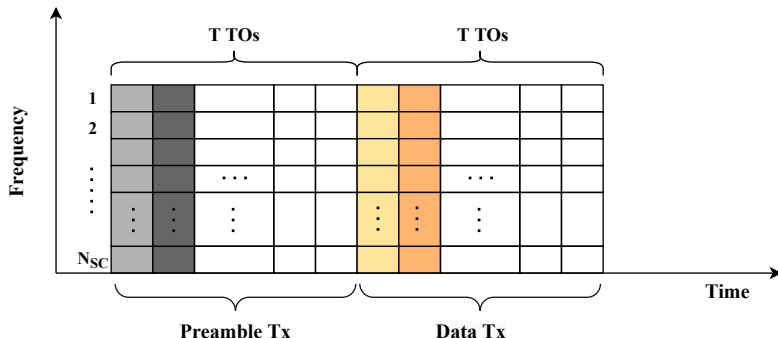
^a[GS Harini et al.](#) "On Preamble-based Grant-Free Transmission in Low Power Wide Area (LPWA) IoT Networks". In: *Proc. IEEE 6th World Forum Internet Things*. IEEE. 2020, pp. 1–6.

Time-frequency resource grid



- Each TO is associated with set of N preambles, $\mathbb{P} = \{P_1, \dots, P_N\}$.

Time-frequency resource grid



- Each TO is associated with set of N preambles, $\mathbb{P} = \{P_1, \dots, P_N\}$.
- Here N and T are chosen such that $NT \geq N_p$.

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Transmitted signal

Signal transmitted by the user choosing i^{th} preamble sequence,

$$x_i(n) = \frac{1}{N_{\text{FFT}}} \sum_{m=0}^{N_{\text{FFT}}-1} P_i(m) e^{j \frac{2\pi n m}{N_{\text{FFT}}}}$$

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Signal at base station (BS)

Superimposed signal received at the BS in any TO with k active users,

$$y(n) = \sum_{i=1}^k h_i e^{j2\pi r_i n} x_i(n - d_i) + w(n)$$

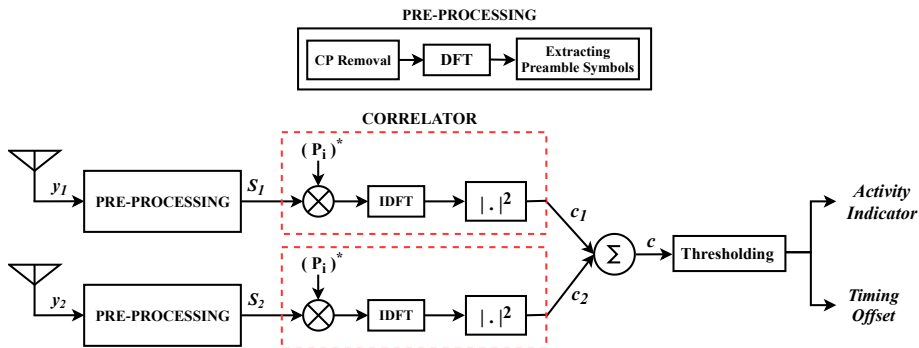
h_i : Flat-fading channel coefficient, r_i : Normalized residual carrier frequency offset and d_i : Timing offset of i^{th} user.

Conventional receiver architecture for AUD

- Given the superimposed signal y , problem at the BS is to **detect all the active users**.

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** The “correlator” block will be **replaced by DNN** in DNN-based AUD.

DNN based AUD

- The AUD at BS is modeled as a **multi label classification** problem.

DNN based AUD

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Binary
Classification



- Spam
- **Not spam**

Multiclass
Classification



- **Dog**
- Cat
- Horse
- Fish
- Bird
- ...

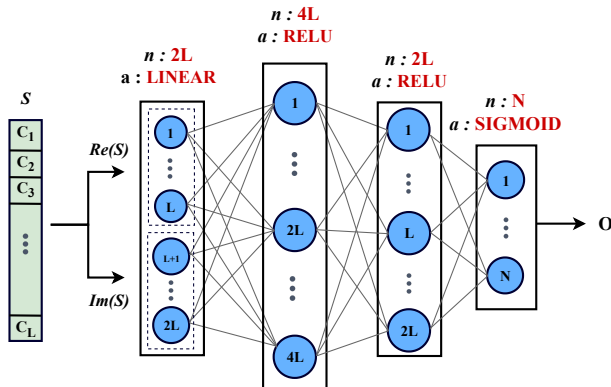
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Source : <https://www.microsoft.com/en-us/research/uploads/prod/2017/12/40250.jpg>

DNN architecture



DNN Training Parameters	
Loss function	Binary Cross Entropy
Optimizer	ADAM
Batch size	1024
Epochs	75
Learning rate	0.001

n : Number of Neurons
 a : Activation Function

** $O = [o_1, \dots, o_N]$, where $o_i \in [0, 1]$.

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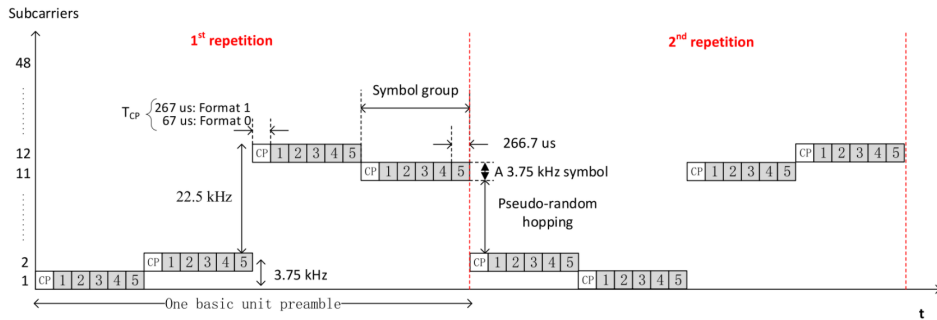
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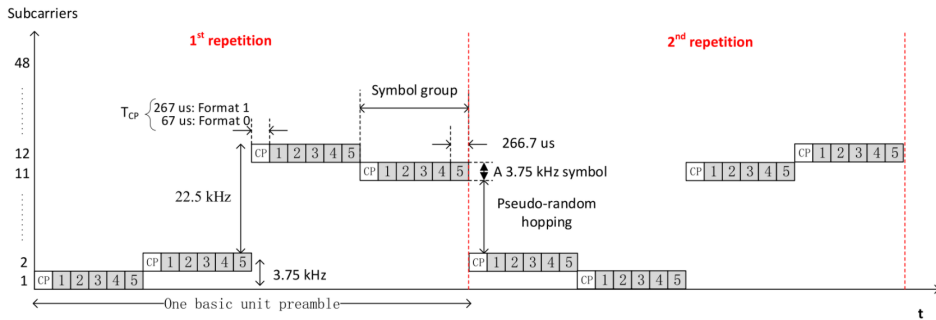
- ▶ Tested on dataset from different SNR values.
- ▶ Outputs from two antenna chain (before thresholding) is averaged to get final output.

Simulation parameters



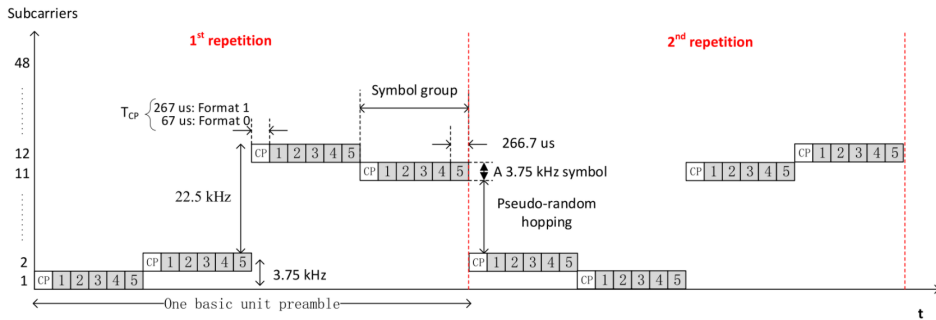
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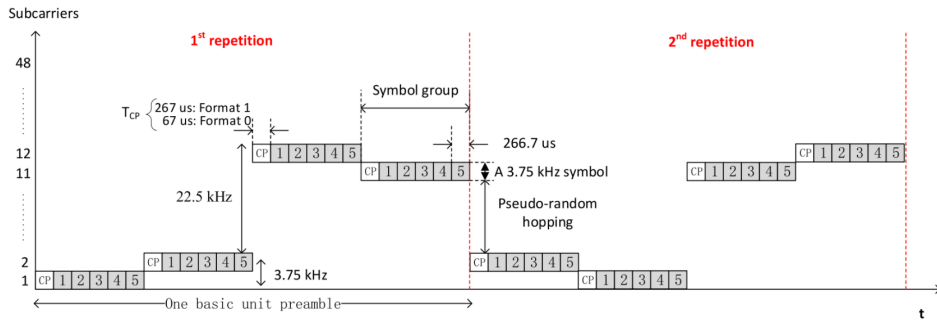
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- ▶ Each TO is of 1 OFDM symbol duration.
- ▶ With this setting $K = 2$ i.e., $\mathbb{P}(k \leq 2) \geq 0.99$.

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Bandwidth, B	180 KHz
Subcarrier spacing, Δf	3.75 KHz
Number of subcarriers, N_{SC}	48
N_{FFT} and N_{CP}	512 and 32 samples
Antenna configuration	1 Tx; 2 Rx
Channel model	EPA 1 Hz
Timing offset	$\text{rand}(0, N_{CP})$ samples
Frequency offset	$\text{rand}(-200, 200)$ Hz
Average success probability	0.9
Number of TOs, T	40
Preamble length, L	47
Average number of users, λ	18

Performance metrics

All-user success probability:

Probability of all active users in a TO being detected correctly.

Per-user success probability:

Probability of successful detection of an individual user.

False alarm probability:

Probability of wrongly detecting a user when only noise is present.

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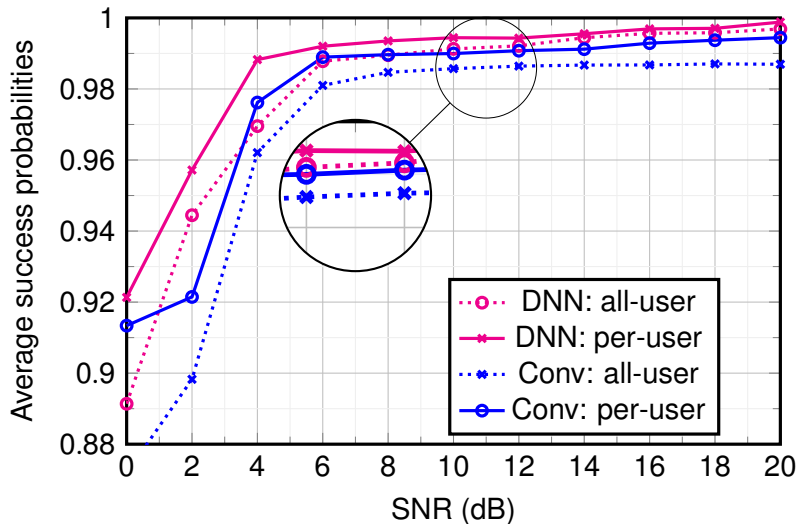
Performance requirements of NB-IoT

- ▶ Per-user success probability ≥ 0.99
- ▶ False alarm probability ≤ 0.001

at 6 dB SNR for CVA 2 and approximately 12 dB SNR for CVA 1 in extended pedestrian A (EPA) channel^a.

^a[3GPP](#). *LTE; Evolved Universal Terrestrial Radio Access (E-UTRA); Base Station (BS) radio transmission and reception*. [TS 36.104](#). [3GPP](#), 2017.

Results



** It is observed that false alarm probability of both methods were below 0.001 for $\text{SNR} \geq 6\text{dB}$.

Computational complexity

- ▶ Number of real floating point operations required is considered.

²Steven G. Johnson et al. “A Modified Split-Radix FFT With Fewer Arithmetic Operations”. In: *IEEE Trans. Signal Process.* (2007).

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Conventional method

- ▶ $\approx N \times (6L + \frac{34}{9} \times N_{\text{FFT}} \log_2(N_{\text{FFT}}))^2$

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$$\approx 32L^2 + 4LN + K \times (6L + \frac{34}{9} \times N_{\text{FFT}} \log_2(N_{\text{FFT}}))$$

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- ** It is observed that the proposed method is ≈ 7 times computationally less complex compared to conventional method.

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Summary

- A GF-NOMA scheme compatible with NB-IoT was presented.

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- DNN-based AUD was discussed, which has the following advantages
 - ▶ Works well in the presence timing and frequency offsets.
 - ▶ Its performance is comparable or better than the conventional AUD scheme for flat-fading (and low mobility) channels.
 - ▶ Its complexity is much smaller than the conventional scheme.
 - ▶ It meets the performance criteria of NB-IoT in both CVA 1 and 2.

NPRACH detection algorithm

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Proposed GF-NOMA scheme

- ▶ Extending the proposed solutions to CVA 3 of NB-IoT.
- ▶ Exploring adaptive threshold based multi label classification.
- ▶ Exploring different preamble lengths.
- ▶ Designing a unified DNN based architecture incorporating collision detection along with AUD.
- ▶ Analyzing the suitability of the proposed method in NTN scenario.

Thank you!