## 6G Technologies and Research

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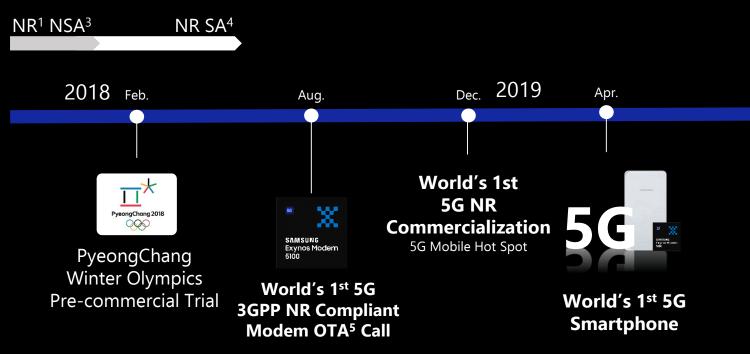
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# **5G Initial Phase**

From Vision to Commercial Deployment

### 3GPP<sup>2</sup> Rel-15



## PHY<sup>6</sup> Technology Highlights

### Scalable OFDM

- 15/30/60 kHz SCS<sup>7</sup> for FR1
- 60/120/240 kHz SCS for FR2

### New Spectrum on mmWave

- Beam Training and Management
- New Channel Coding
  - LDPC<sup>8</sup> and Polar codes

4. Standalone
 5. Over-the-Air
 6. Physical Layer

7. Subcarrier Spacing
 8. Low Density Parity Check

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# **5G Second Phase**

3GPP Rel-16 (2018~2020) and 17 (2020~2022)

## **Use Case Expansion**

NR V2X<sup>1</sup> and Sidelink

Industrial IoT<sup>2</sup> NR Unlicensed (5GHz and 60GHz) NTN<sup>3</sup>

RedCap<sup>4</sup> (for wearables etc.)

## **Efficiency Improvement**

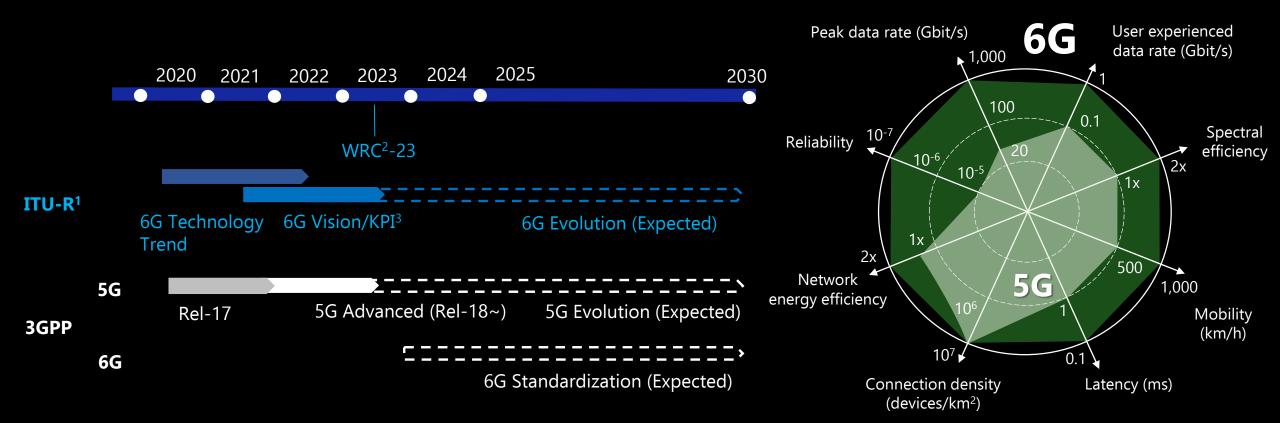
Accurate positioning Integrated access and backhaul (IAB) UE power saving Coverage enhancement (for PUSCH/PUCCH) <u>MIMO<sup>5</sup> enhancement</u>

Vehicle-to-everything
 Internet of things
 Non-terrestrial network

Reduced capability devices
 Multiple-input multiple-output

# **B5G and 6G**

## Roadmap and Performance Target



1. International Telecommunication Union Radiocommunication Sector

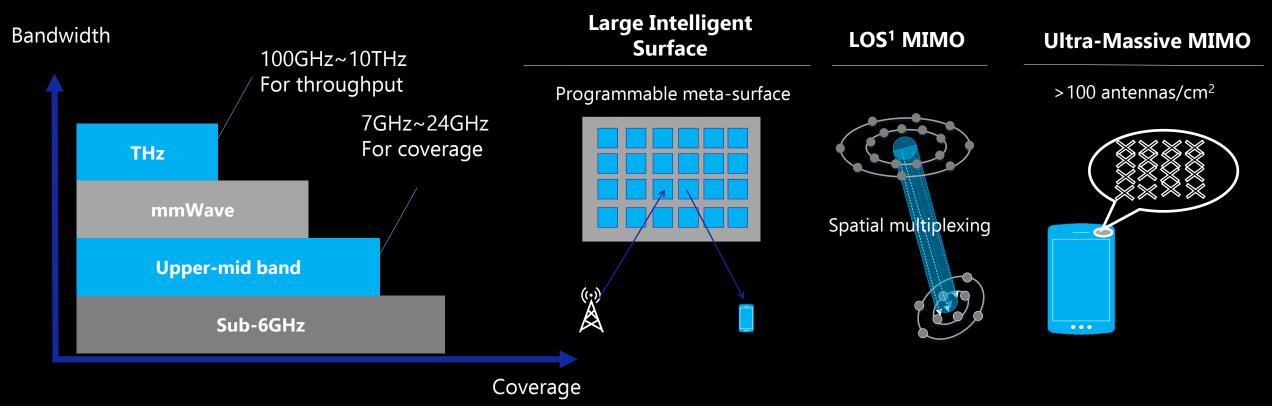
2. World Radiocommunication Conference

3. Key performance indicator

# **6G Technologies**

## More Spectrum and More Antenna

- THz provides large bandwidth, but challenges are there for RFIC/transceiver design and coverage
- Upper-mid band will provide an anchor for coverage
- Large amount of advanced antennas is critical for reliable communication in such 6G spectrum



# **6G Technologies**

**Higher Efficiency** 

## Advanced duplex

- BS<sup>1</sup>-side sub-band full duplex in 5G advanced (no UE<sup>2</sup> full duplex)
- BS and UE true full duplex in 6G; self (internal/reflected) interference cancellation

## Waveform, modulation and coding

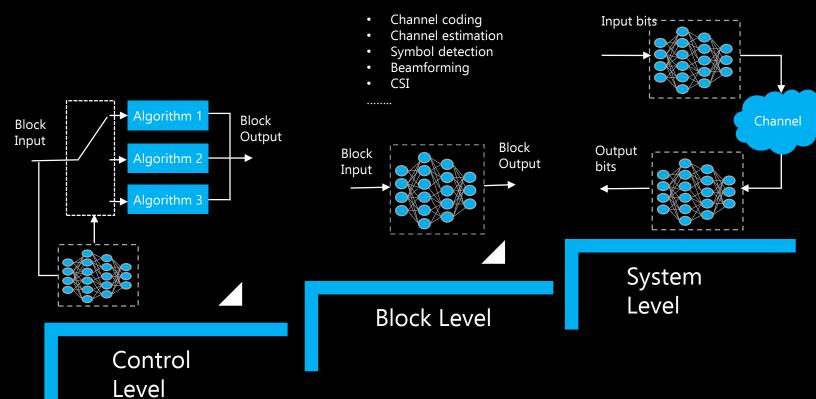
- High phase noise tolerance
- New baseband signal processing for full-digital massive MIMO with low resolution ADC<sup>3</sup>/DAC<sup>4</sup>
- New channel coding for high reliability and low latency

## Energy efficiency

- Near-zero energy communications
- BS and UE energy saving

# **6G Technologies**

## Higher Intelligence



- AI for PHY in 5G advanced
  - CSI
  - Beam management
  - Positioning
- AI native PHY for 6G
- Comprehensive AI in all layers
  - End-to-End AI
  - Split computing
  - Federated learning

## **AI<sup>1</sup> for PHY: Intelligent Modem**

# **Information Theory and Cellular PHY**

- Information theory has provided reference of achievability limit for the rate of communication.
- In many cases, it has also provided specific schemes achieving the optimality which commercial systems have also taken advantages of.
- Point-to-point channel (memoryless, no feedback) Capacity established
  - Source-channel coding separation
  - BICM<sup>1</sup> a backbone of the current cellular system framework
  - Coding across fading
  - LDPC code (spatially coupled) currently used for 5G data channel
  - Polar code currently used for 5G control channel

# **Information Theory and Cellular PHY**

- MIMO AWGN channel Capacity established
  - Successive cancellation receiver with MMSE/whitening popular MIMO detection implementation
  - SVD beamforming and rank selection equal power constraint popular in CSI framework
- Multi-user (MU) MIMO AWGN channel Capacity partially established
  - Broadcast (DL MU) \_ Superposition for single antenna case 4G Rel-13 MUST<sup>1</sup>

— Non-linear precoding for multiple antenna case - 5G candidate

- Multiple access (UL MU) transparently realizable in the current system -> 5G Rel-16 NOMA<sup>2</sup> study
- Interference (DL multi-Tx) -

No capacity yet, the current system mostly does TIN<sup>3</sup> Interference alignment - 4G Rel-11 CoMP<sup>4</sup> and 5G TIN, but as structured noise - 4G Rel-12 NAICS<sup>5</sup>

- Multi-user superposition transmission
  Non-orthogonal multiple access
- 4. Coordinated multi point
- 5. Network-assisted interference cancellation and suppression

3. Treating interference as noise

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## **6G Research**

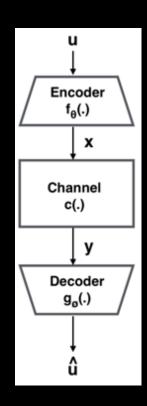
Deep-learning based channel coding

- Optimal algorithms for point-to-point communication are often available.
  - Model mismatch breaks optimality.
  - Optimal algorithms are often too complex.
  - With emergence of AI, we increasingly rely on 'machine' to resolve the issue.

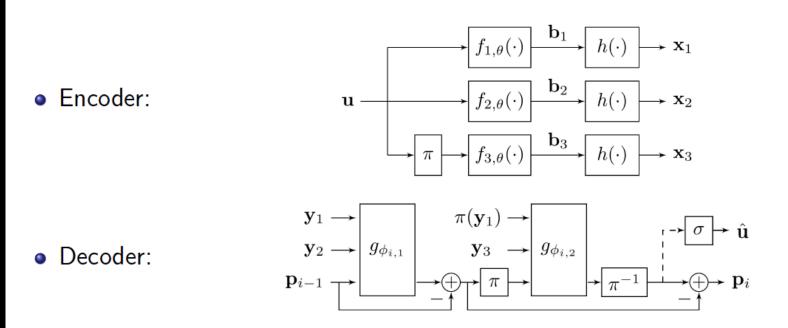
- We will discuss designing of channel codes using deep learning.
  - Interleaver Design and Pairwise Codeword Distance Distribution Enhancement for Turbo Autoencoder, **GLOBECOM 2021**
  - Product AE: Towards Training Larger Channel Codes based on Neural Product Codes, ICC 2022 best paper award

# Turbo Autoencoder (AE)

• System model: y = x + nwhere n is AWGN.



• Turbo AE: based on either CNN<sup>1</sup> or RNN<sup>2</sup> for f or g

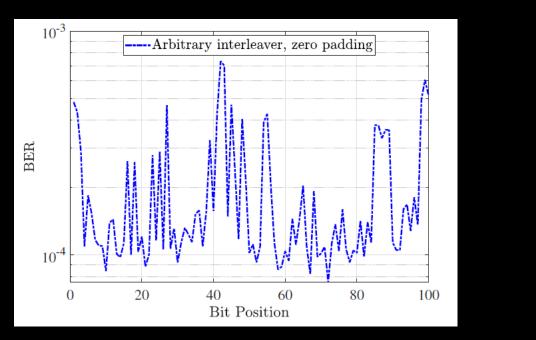


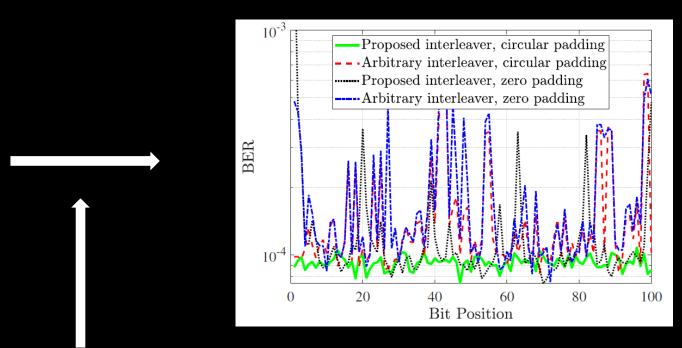
<sup>&</sup>lt;sup>1</sup>Y. Jiang, H. Kim, H. Asnani, S. Kannan, S. Oh, and P. Viswanath, "Turbo autoencoder: Deep learning based channel codes for point-topoint communication channels" in Advances in Neural Information Processing Systems, H. Wallach, H. Larochelle, A. Beygelzimer, F. dAlch ´e-Buc, E. Fox, and R. Garnett, Eds., vol. 32. Curran Associates, Inc., 2019.

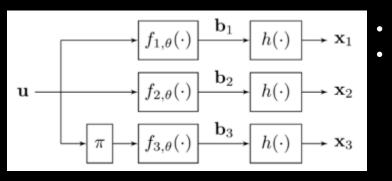
2. Recurrent neural network

<sup>1.</sup> Convolutional neural network

# Turbo AE: Resolving Non-Uniform BER<sup>1</sup>



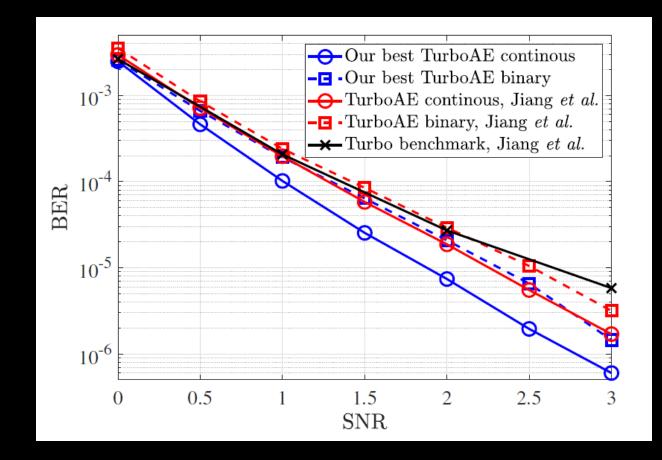




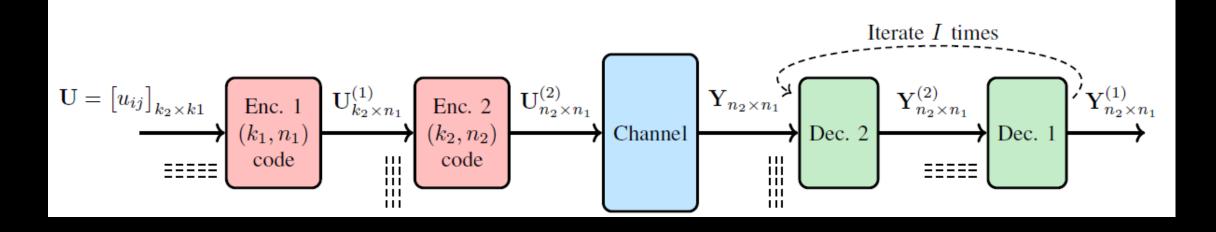
- Circular padding instead of zero padding for CNN encoder f and decoder g. Interleaver  $\pi(i) = i\delta \mod K$ ,  $i = 0, 1 \dots, K - 1$ .
  - Ensuring certain symmetry  $(\pi(i) \pi(i+k)) \mod K = (\pi(j) \pi(j+k)) \mod K$

# Enhanced Average BER

- A regularizer ensuring good minimum distance 'T'.
  - Loss = BCE +  $\lambda \sum_{1 \le i, j \le 2^K, i \ne j} (T d(c_i, c_j))_+$ ,  $c_i$  is 'i'th codeword,  $(x)_+ = \max\{0, x\}$ .
- As a result, we achieved enhanced average BER performance (K=100).

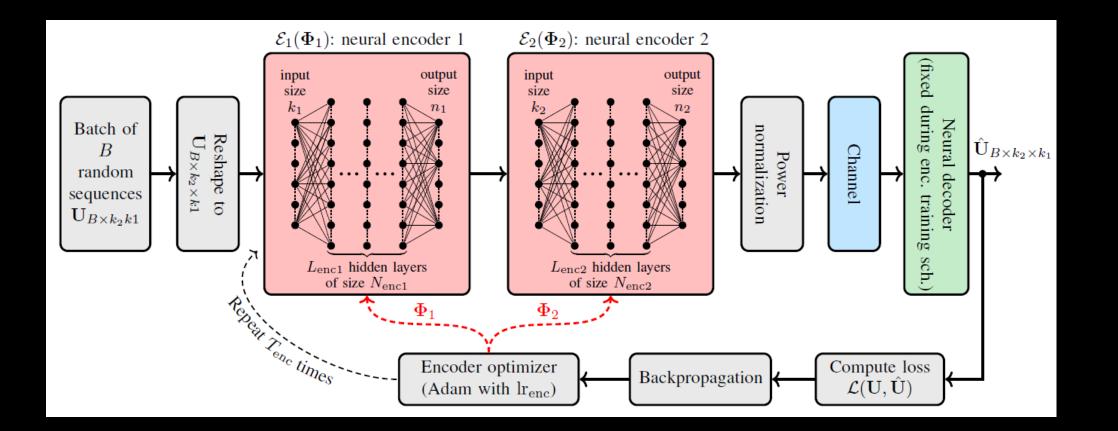


## Product Autoencoder



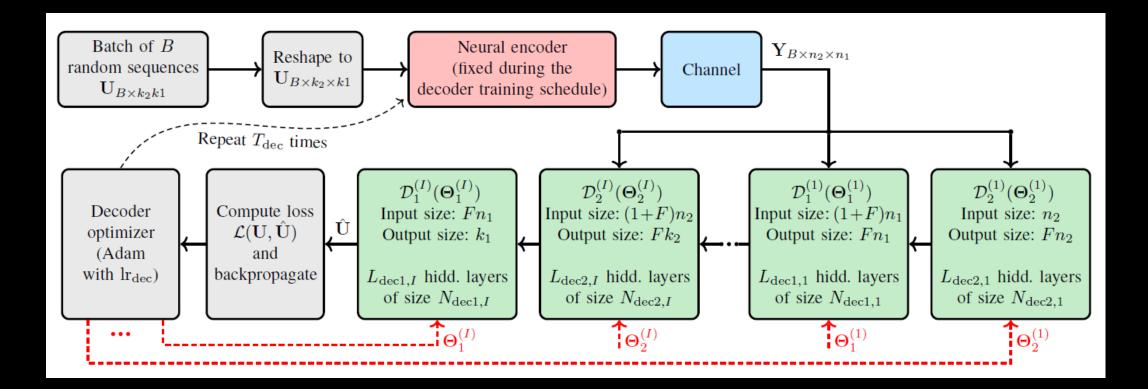
- The main motivation is to design a trainable large block length codes using smaller length component codes.
- Enc. 1 encodes each row using an  $(n_1, k_1)$  code.
- Enc. 2 encodes each column using an  $(n_2, k_2)$  code.
- Dec 2 encodes the columns while Dec. 1 decodes the rows.
- The decoder iterates the process several times (*I* times).

## Product AE: Encoder training



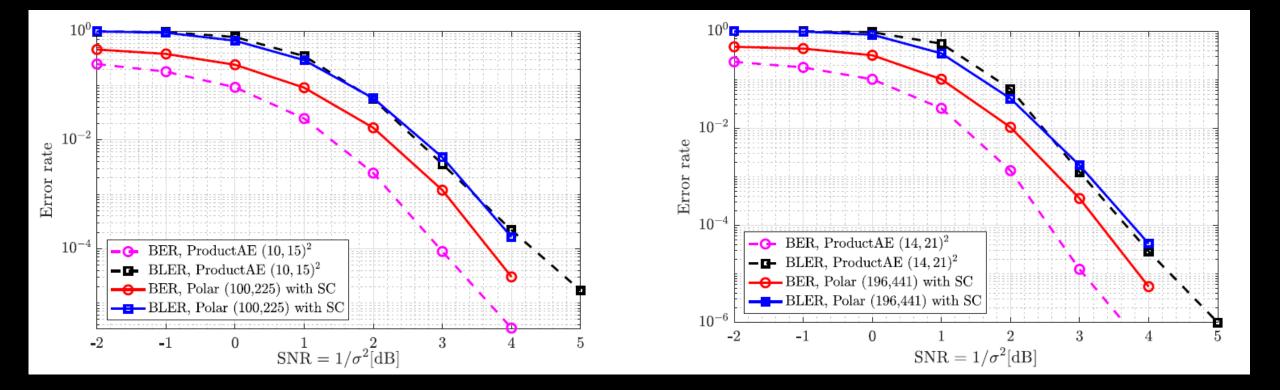


## Product AE: Decoder training



## Product AE: Results

- Product AE shows better BER performance than polar codes.
  - The current BCE loss function is not designed for BLER performance.



# Thank you!