

UNEQUAL ERROR PROTECTION TURBO AND LDPC CODES

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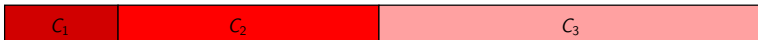
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OVERVIEW

- Unequal Error Protection (UEP)
- UEP **convolutional and Turbo codes**
 - ▶ Pruning vs. puncturing
- Bit-irregular UEP **LDPC codes**
 - ▶ Irregular bit-node profile
 - ▶ Behaviour and optimisation target
 - ▶ Algorithm
- Bit-irregular UEP LDPC codes for **higher order constellations**
 - ▶ Motivation
 - ▶ Extended density evolution
- Summary

UNEQUAL ERROR PROTECTION (UEP)

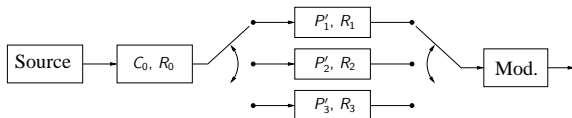
- Data of **different importance** from source encoder



- Multimedia:
 - ▶ Header
 - ▶ DC components, magnitude information
 - ▶ AC components, high resolution data, position information
- Different **protection classes**/levels provided by
 - ▶ Physical layer: modulation, bit loading, power loading, ...
 - ▶ Network layer: network coding, protocols, ...
 - ▶ **Channel coding**: variable rate codes, local properties, ...

CONVENTIONAL UEP CONVOLUTIONAL CODES

- Goal: a code family with **several code rates** $R_i = \frac{k_i}{n_i}$
- Well-known: **Puncturing**
- **Increased code rate** by excluding code bits from transmission



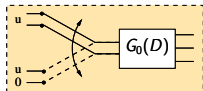
DECODING

- Punctured positions are known to the receiver
- **Content is unknown**, 0 and 1 are assumed equally likely
- **No contribution** to decoding

WHAT ABOUT REDUCING THE CODE RATE?

NEW APPROACH: PRUNING

- **Reduced code rate** by inserting known bits into the info sequence
- Pruning pattern: Insert bits **periodically in certain patterns** (\leftrightarrow puncturing pattern)
- Fixing bits means **pruning paths from the trellis** \rightarrow **sub-codes**

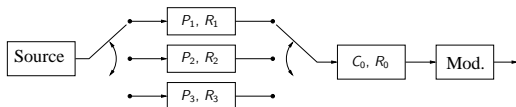
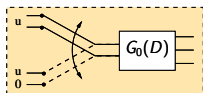


DECODING

- Receiver knows pruned positions **and their content**
- **A-priori knowledge** of pruned positions is **infinite!**
- Turbo codes: **Increased extrinsic information!**

NEW APPROACH: PRUNING

- **Reduced code rate** by inserting known bits into the info sequence
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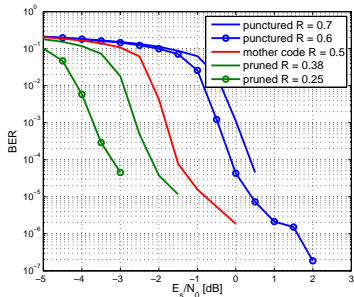
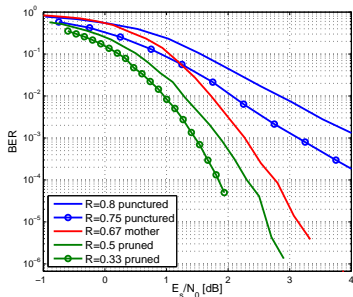


DECODING

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- **A-priori knowledge** of pruned positions is **infinite!**
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TIME-VARIANT PRUNING - RESULTS

- Convolutional (NSC) mother code, $R = 2/3$, $N = 200$
- Punctured: $R = 3/4$ and $R = 4/5$
- Pruned: $R = 1/2$ and $R = 1/3$
- Turbo code with (RSC) mother code, $R = 0.5$, $N = 1000$
- Punctured: $R = 0.6$ and $R = 0.7$
- Pruned: $R = 0.25$ and $R = 0.38$



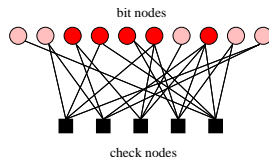
BIT-IRREGULAR UEP LDPC CODES

LDPC CODES

- Block codes with sparse parity-check matrix H
- Irregular bit-node profile $\tilde{\lambda}(x)$
- Group bit nodes into **protection classes**
- Define and **optimise bit-node profile for each class** $\rightarrow \tilde{\lambda}^{(k)}(x)$

$$\begin{aligned}\tilde{\lambda}(x) &= \tilde{\lambda}_2 \cdot x + \tilde{\lambda}_3 \cdot x^2 + \dots \\ &= \tilde{\lambda}_2^{(1)} \cdot x + \tilde{\lambda}_2^{(2)} \cdot x + \tilde{\lambda}_3^{(1)} \cdot x^2 + \tilde{\lambda}_3^{(2)} \cdot x^2 + \dots \\ &= \tilde{\lambda}^{(1)}(x) + \tilde{\lambda}^{(2)}(x)\end{aligned}$$

- **Common check-node profile** $\tilde{\rho}(x)$ assumed for all classes



INHERENT AND ENHANCED UEP OF LDPC CODES

INHERENT UEP

- **Highly connected nodes** are protected better!
- Most connected bit nodes assigned to most important class
- **Maximise average connection degree** of bit nodes in each class

ENHANCED UEP

By density evolution, one can show that a class' error probability is minimised by

- maximising the **average connection degree** $d_{av}^{(k)}$ and
- maximising the **minimum degree** $d_{min}^{(k)}$ of its bit nodes.

OPTIMISATION STRATEGY

Optimisation is done **class after class**, most important class first!

OBJECTIVE FUNCTION AND CONSTRAINTS

- Objective functions: $\max d_{vav}^{(k)}$ and $\max d_{vmin}^{(k)}$
- Constraints:
 - C1 **Rate** constraint
 - C2 **Proportion** distribution constraints
 - C3 **Convergence** constraint
 - C4 **Stability** condition
 - C5 **Minimum bit-node degree** constraint
 - C6 **Previous classes** constraints

USED OPTIMISATION METHOD

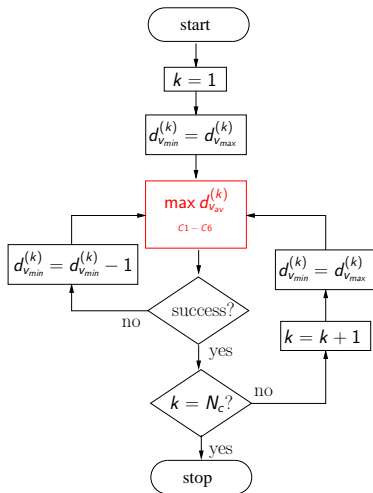
- Linear programming (LIPSOL, simplex)
- **Single objective function**
- Equality/inequality constraints, lower/upper bounds

HIERARCHICAL OPTIMISATION ALGORITHM

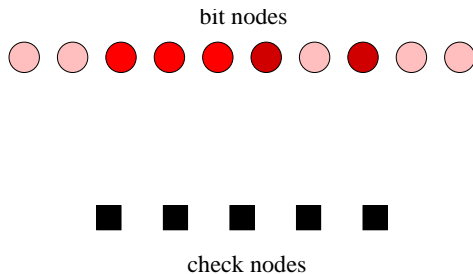
- Given parameters: σ^2 (SNR), $\tilde{\rho}(x)$, N_c , R
- For **each class k** , starting with the most important class
 - Initialisation $d_{v_{min}}^{(k)} = d_{v_{max}}^{(k)}$
 - While optimisation failure
 - Maximise average connection degree of class k :

$\max d_{v_{av}}^{(k)}$

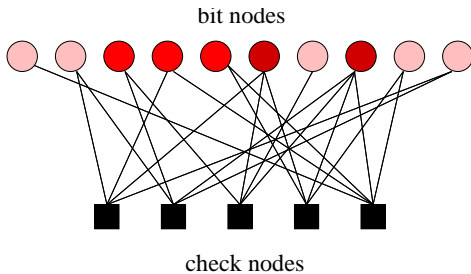
fulfilling constraints C1-C6
 - $d_{v_{min}}^{(k)} = d_{v_{min}}^{(k)} - 1$
 - End
- End



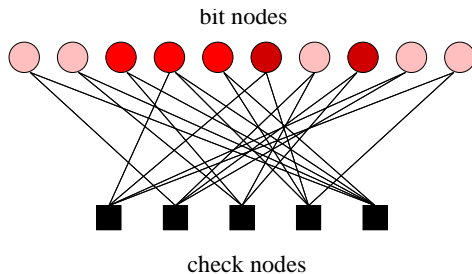
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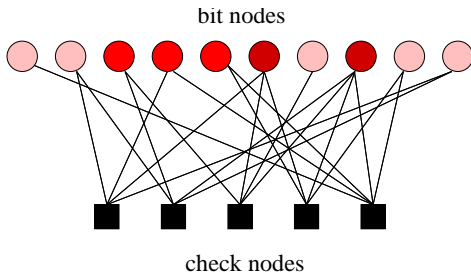
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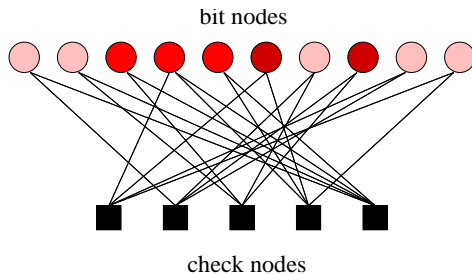
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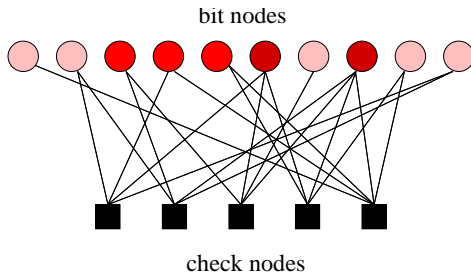
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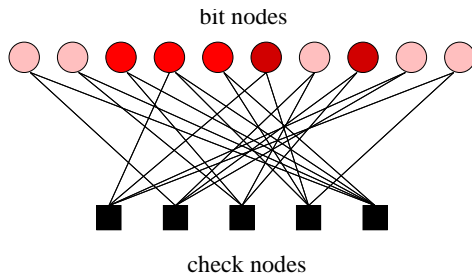
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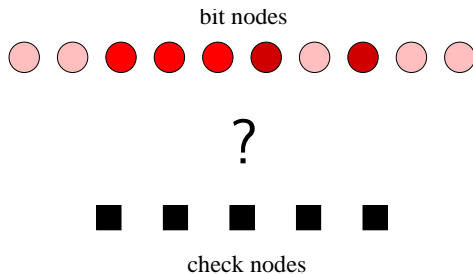
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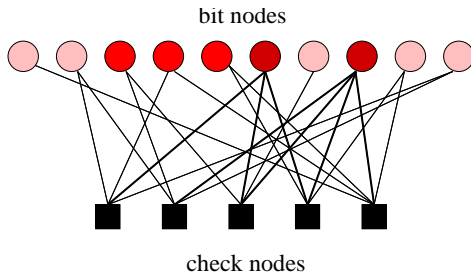
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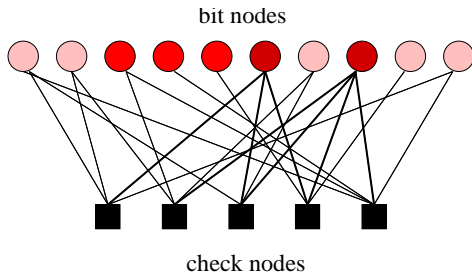
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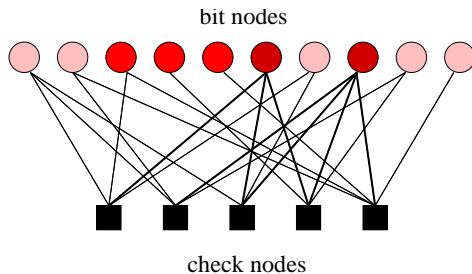
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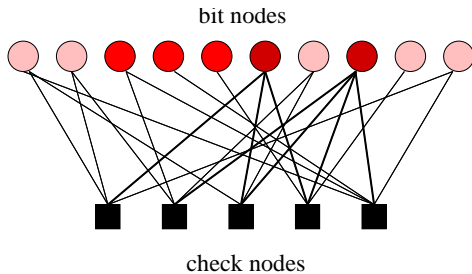
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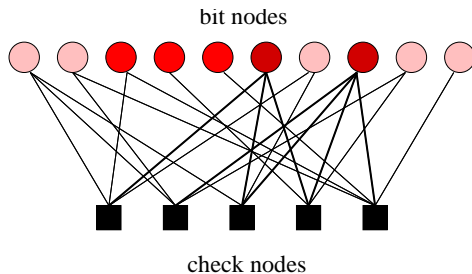
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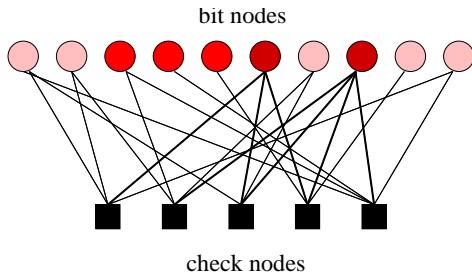
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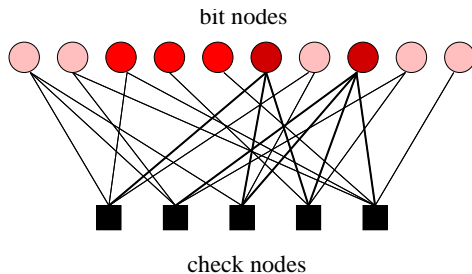
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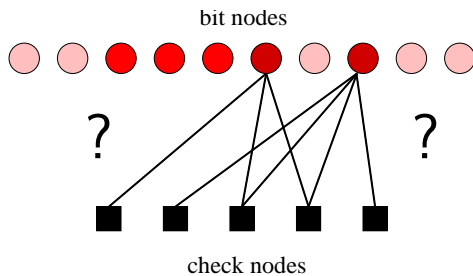
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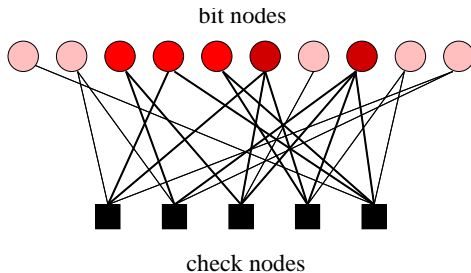
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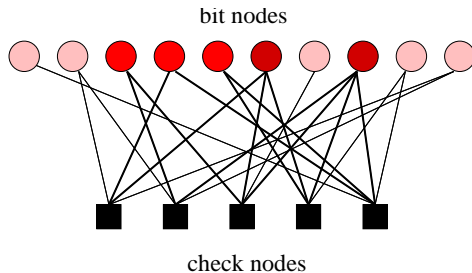
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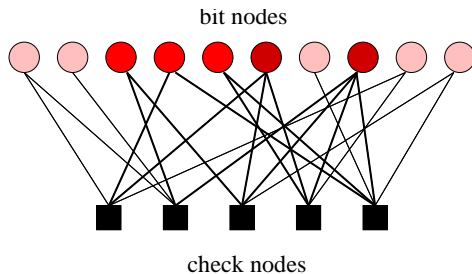
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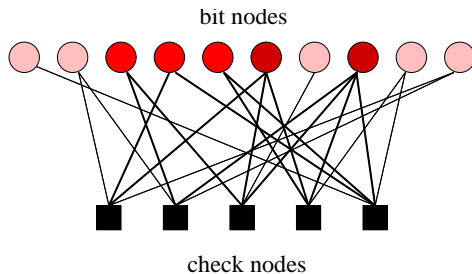
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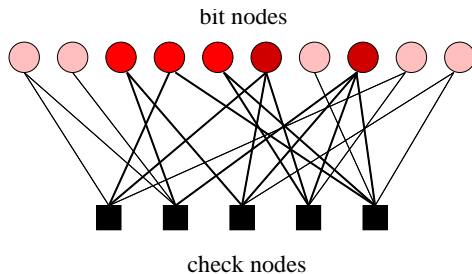
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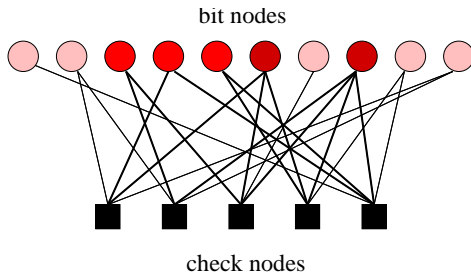
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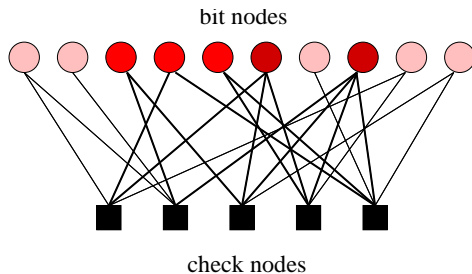
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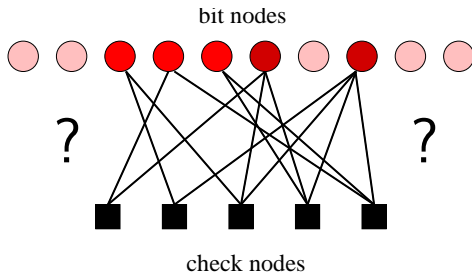
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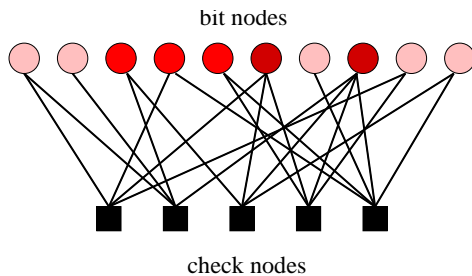
HIERARCHICAL OPTIMISATION ALGORITHM



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BIT-IRREGULAR UEP LDPC CODES FOR HIGHER ORDER CONSTELLATIONS

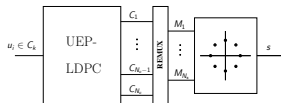
MOTIVATION

- **Heterogeneous bit-error probabilities**
- Example 8-PSK:

$$P_{b,d_0} = 1/2 \cdot P_s$$

$$P_{b,d_1} = P_{b,d_2} = 1/4 \cdot P_s$$

- Conventional LDPC-design assumes homogeneous channel noise variance σ^2 in density evolution
- Noise variance is **not any more equal** for all bits!



TASKS FOR IMPROVED DESIGN

- Find a way to **treat bit positions differently**
- Find **equivalent noise variances** for the bit positions

IMPROVED DESIGN

EXTENDED DENSITY EVOLUTION



$$I_{CV}^{(l-1)} = 1 - \sum_{t=2}^{d_{cmax}} \rho_t \cdot J \left((t-1) \cdot J^{-1} \left(1 - I_{VC}^{(l-1)} \right) \right)$$



$$I_{VC}^{(l)} = \sum_{i=2}^{d_{vmax}} \lambda_i \cdot J \left(\frac{2}{\sigma^2} + (i-1) \cdot J^{-1} \left(I_{CV}^{(l-1)} \right) \right)$$

IMPROVED DESIGN

EXTENDED DENSITY EVOLUTION



$$I_{cv}^{(l-1)} = 1 - \sum_{t=2}^{d_{cmax}} \rho_t \cdot J \left((t-1) \cdot J^{-1} \left(1 - I_{vc}^{(l-1)} \right) \right)$$



$$I_{vc}^{(l)} = \sum_{k=1}^{N_c} \sum_{i=2}^{d_{vmax}} \lambda_i^{(k)} \cdot J \left(\frac{2}{\sigma^2} + (i-1) \cdot J^{-1} \left(I_{cv}^{(l-1)} \right) \right)$$

IMPROVED DESIGN

EXTENDED DENSITY EVOLUTION



$$I_{CV}^{(l-1)} = 1 - \sum_{t=2}^{d_{cmax}} \rho_t \cdot J \left((t-1) \cdot J^{-1} \left(1 - I_{VC}^{(l-1)} \right) \right)$$



$$I_{VC}^{(l)} = \sum_{j=1}^{N_s} \sum_{k=1}^{N_c} \sum_{i=2}^{d_{vmax}} \lambda_i^{(k,j)} \cdot J \left(\frac{2}{\sigma_j^2} + (i-1) \cdot J^{-1} \left(I_{CV}^{(l-1)} \right) \right)$$

IMPROVED DESIGN

EXTENDED DENSITY EVOLUTION



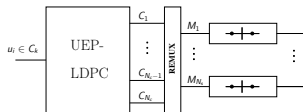
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EQUIVALENT NOISE VARIANCES σ_j^2

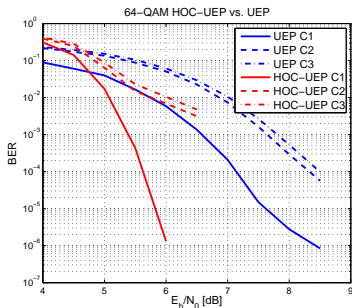
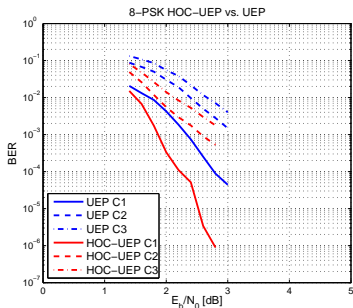
- Gray labelling: $\log_2 M$ equivalent BPSK channels
- Equivalent noise variances based on bit-error probabilities
- $\sigma^2 \stackrel{8\text{-PSK}}{\equiv} P_s \xrightarrow{\text{Gray}} P_{b,d_i} \stackrel{\text{BPSK}}{\equiv} \sigma_j^2$



UEP LDPC CODES FOR HOC - RESULTS

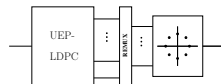
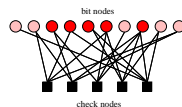
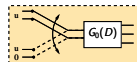
We compare

- UEP codes **designed for BPSK** (UEP) with
- UEP codes **designed for 8-PSK/64-QAM** (HOC-UEP).
- Both schemes are used with higher order constellations.
- $N_c = 3$, $n = 4096$, $R = 0.5$, $N_{it} = 20$



SUMMARY

- Pruned UEP convolutional and Turbo codes
 - Pruning through **inserting zeros** into the info sequence
 - Set of BER curves in both directions
- Bit-irregular UEP LDPC codes
 - **Different bit-node profiles** for protection classes
 - Hierarchical optimisation algorithm
- UEP LDPC codes for higher order constellations
 - Awareness of **heterogeneous noise variance**
 - **Extended density evolution**



Thank you!