

Minimum Distortion Variance Concatenated Block Codes for Embedded Source Transmission

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Outline

- Source quality assessment basics
- Progressive source compression
- Unequal Protection Schemes:
 - Conventional Schemes.
 - Previous work: Concatenated Block Coding
- Few results and issues about the previous work
- Description of the extension scheme (proposed)
 - Optimization of parameters
- Numerical results

Source quality assessment Basics: Image compression

- Given two images I and I' (original and the noisy version), the distortion will be measured by Mean Square Error (MSE):

$$MSE = \frac{1}{L_x \times L_y} \sum_{y=1}^{L_x} \sum_{x=1}^{L_y} [I(x, y) - I'(x, y)]^2$$

where L_x and L_y are dimensions of the image.

- Peak Signal to Noise Ratio (PSNR in dB) is defined to be

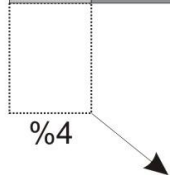
$$PSNR = 10 \times \log_{10} \left(\frac{I_{max}^2}{MSE} \right)$$

where I_{max} is the maximum possible intensity value of the image.

- For monochromatic gray scale image: $I_{max} = 255$
- Lower MSE (larger PSNR) means better image quality.
- “Source rate” means the average number of bits spent per pixel (bpp). For a given PSNR value, the lower the source rate is, the better the compression will be.

Progressive Source Compression

SPIHT Encoded Bit Stream



0.01bpp, PSNR=22.55dB

- Ex: SPIHT image compression algorithm [1]. 4% gives you only a brief description of the source.

[1] A. Said and W. A. Pearlman, "A New Fast and Efficient Image Codec Based on Set Partitioning in Hierarchical Trees," *IEEE Trans. on Circuits and Systems for Video Tech.*, vol. 6, pp.243-250, June 1996.

Progressive Source Compression

SPIHT Encoded Bit Stream



0.01bpp, PSNR=22.55dB

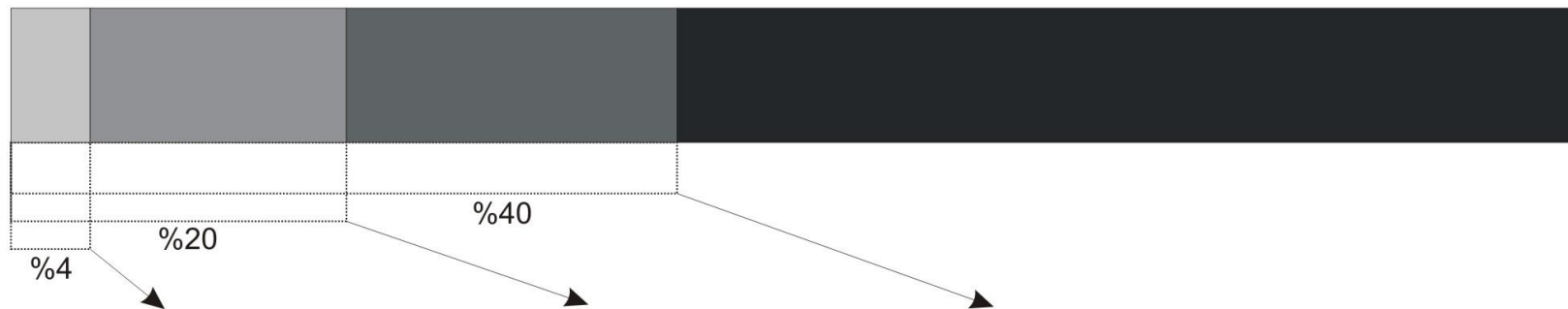


0.05bpp, PSNR=27.17dB

- 20% is good enough to say what the picture looks like.

Progressive Source Compression

SPIHT Encoded Bit Stream



0.01bpp, PSNR=22.55dB



0.05bpp, PSNR=27.17dB

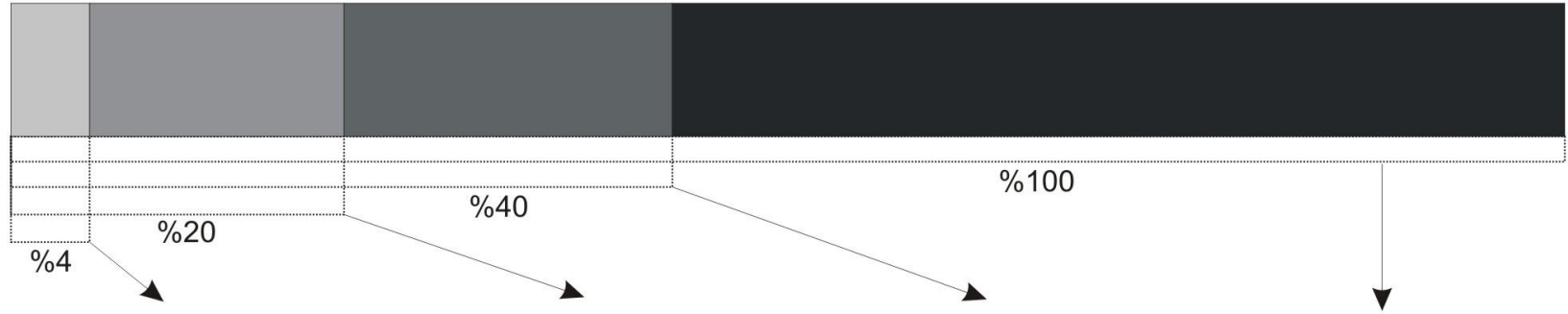


0.1bpp, PSNR=29.81dB

- At 40%, it begins to refine the image.

Progressive Source Compression

SPIHT Encoded Bit Stream



0.01bpp, PSNR=22.55dB



0.05bpp, PSNR=27.17dB



0.1bpp, PSNR=29.81dB



0.25bpp, PSNR=33.68dB

- At 100%, it gives more refinement but no major difference from 40%.

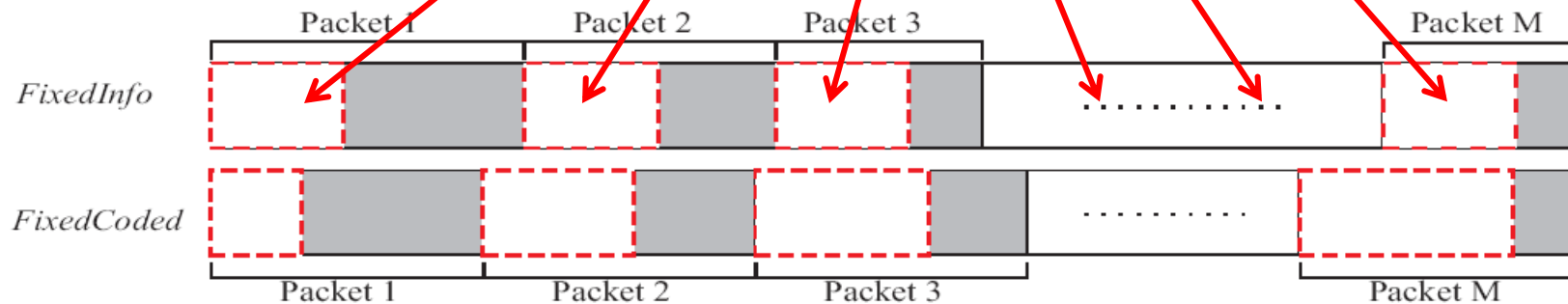
Progressive Source Compression

- We consider progressive type of encoders.
 - Embedded image encoders: EZW, SPIHT, JPEG2000 etc.
 - Image compression using singular value decomposition (SVD).
- **Result: Very sensitive to bit errors.**
- Protection and performance improvement is achieved by error correction coding.
- **Way to go:** Unequal error protection (UEP) is beneficial for progressively encoded sources. This can be provided by several known techniques.
- We consider a concatenated coded scheme.

Unequal Error Protection Schemes: REVIEW

❖ Two major methods:

Progressive bit stream

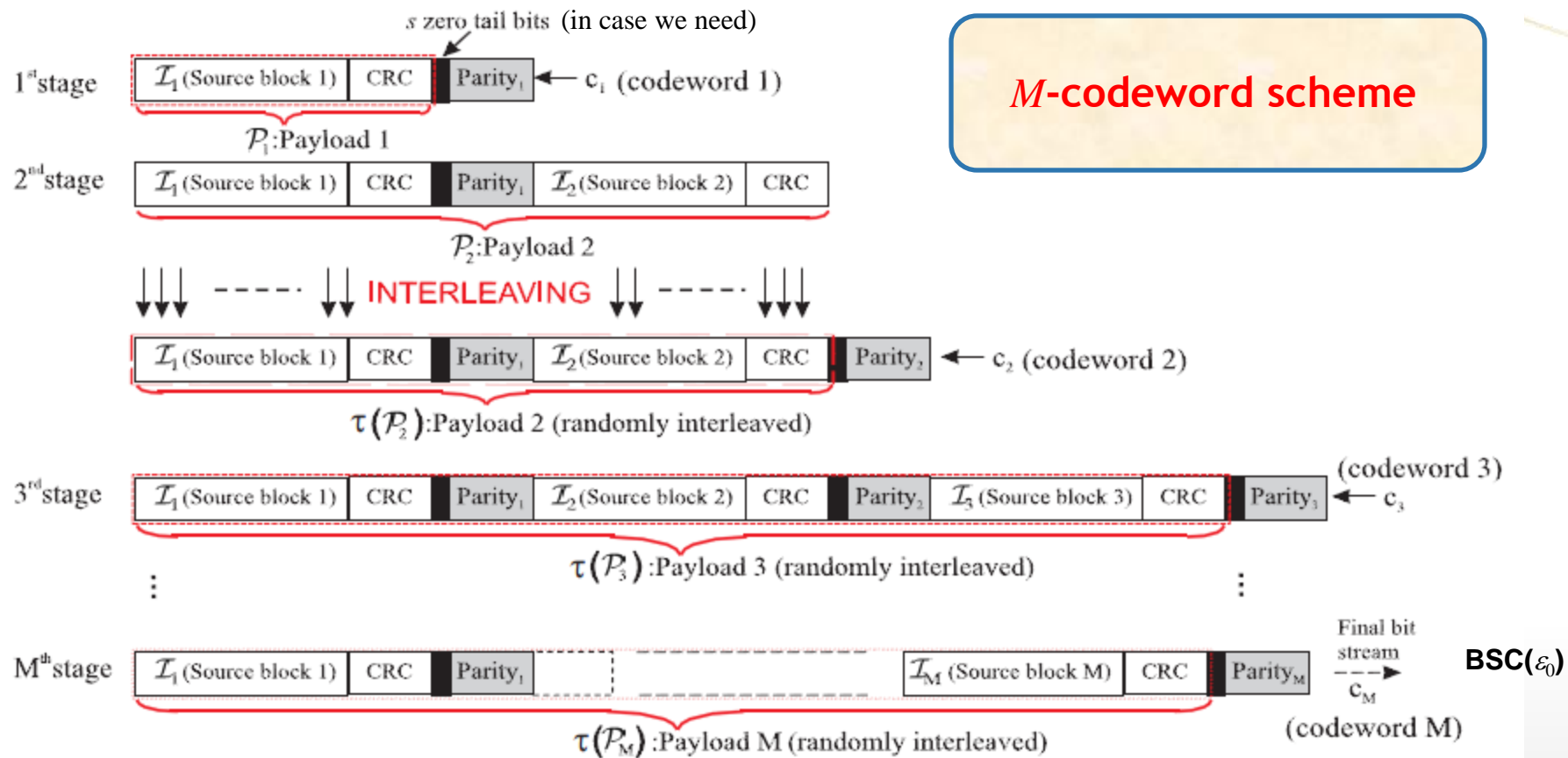


- *FixedInfo*, single channel code rate for all the packets.
- *FixedCoded*, single channel code rate for all the packets.
- *FixedInfo & FixedCoded*, different channel code rates for each packet.

■ Error Correction Codes include:

- Conventional Block Codes (BCH, Golay, etc),
- Rate-Compatible Punctured Convolutional (RCPC) Codes,
- Rate-Compatible (RC) Turbo codes, RC-LDPC codes
- Reed Solomon (RS) codes.

Concatenated Block Coding for embedded bit stream transmissions



- Find the number of source blocks M , the rate of channel codes based on a bit budget constraint (Transmission rate) and a target error rate using minimum average distortion criterion.

Few results...

Use 512 X 512 *Lena* Image



RCPC codes with rates:

$$\mathcal{C} = \{8/9, 4/5, 2/3, 4/7, 1/2, 4/9, 2/5, 4/11, 1/3, 4/13, 2/7, 4/15, 1/4\}$$

$\varepsilon_0 = 0.1$ and transmission rate (r_{tr}) = 0.3bpp ($0.3 \times 512 \times 512 = 79643$ bits)

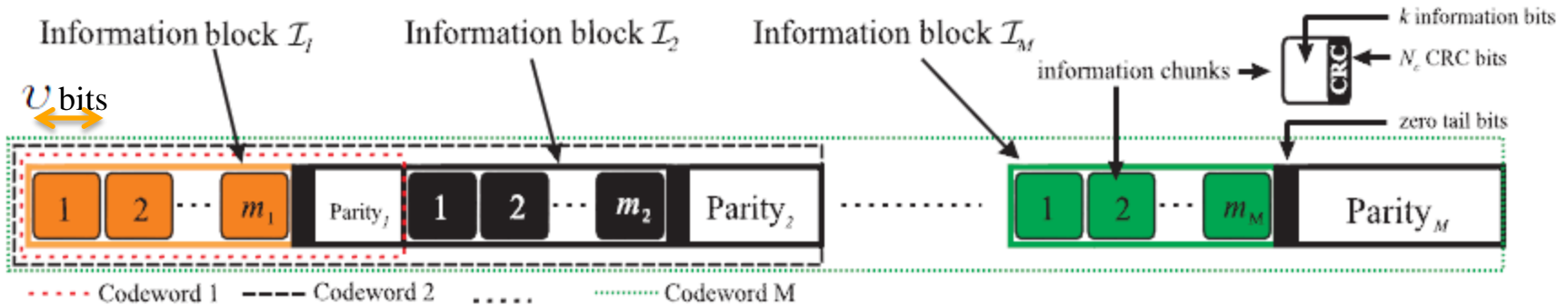
bit budget

M	r_1	r_2	r_3	r_4	r_5	PSNR (dB)
1	1/4	-	-	-	-	20.44
2	2/3	1/3	-	-	-	28.45
3	8/9	4/5	4/13	-	-	28.71
4	8/9	8/9	4/5	1/3	-	28.79
5	1	8/9	8/9	4/5	1/3	28.75

Observations

- In an optimal setting, this coding scheme results in four or five source blocks.
- Number of reconstruction levels is five or six.
- **Result:** User dissatisfaction due to large quality fluctuations.
- We consider a broadcast scenario.
 - One server, multiple receivers with varying channel conditions.
- Minimum average distortion.
 - Sufficient for point-to-point communication.
 - Minimum average does not imply minimum variance.
- **Result:** User dissatisfaction due to unfair service quality.

Extension System and Optimization



- M codewords. Each information block is chopped.
- Number of reconstruction levels: $\sum_{l=1}^M m_l + 1$
- This extensions increases the redundancy due to CRC.
 - Less space for source bits:

$$\sum_{l=1}^M \mathcal{I}_l - (m_l - 1)N_r \leq \sum_{l=1}^M \mathcal{I}_l$$

Extension System and Optimization

- Original Problem: A code allocation policy π allocates the channel code $c_\pi^{(i)} \in \mathcal{C}$ to be used in the i -th stage of the algorithm.
- Let $\bar{D}_\pi(n)$ denote the n -th moment of the distortion at the receiver using policy π .
- Let N_s be the number of source samples B is the bit budget.
- Minimum Average Distortion Problem:

$$\min_{\pi, \xi, v} \bar{D}_\pi(1) \text{ such that } r_{tr} = \frac{1}{N_s} \sum_{i=1}^M \frac{m_i v}{\prod_{j=i}^M r_\pi^{(j)}} \leq B$$

$$\xi = \{m_1, \dots, m_M\}$$

Extension System and Optimization

- Constrained Minimization of Distortion Variance:

$$\min_{\pi, \xi, v} \sigma_{\pi}^2 \text{ such that } r_{tr} = \frac{1}{N_s} \sum_{i=1}^M \frac{m_i v}{\prod_{j=i}^M r_{\pi}^{(j)}} \leq B, \bar{D}_{\pi}(1) \leq \gamma_D$$

$$\sigma_{\pi}^2 = \bar{D}_{\pi}(2) - \bar{D}_{\pi}^2(1)$$

- **Assume:** σ_{π}^2 is a non-increasing function of $\bar{D}_{\pi}(1)$ using policy π
- Minimization of Second moment of Distortion: Set $\bar{D}_{\pi}(1) = \gamma_D$

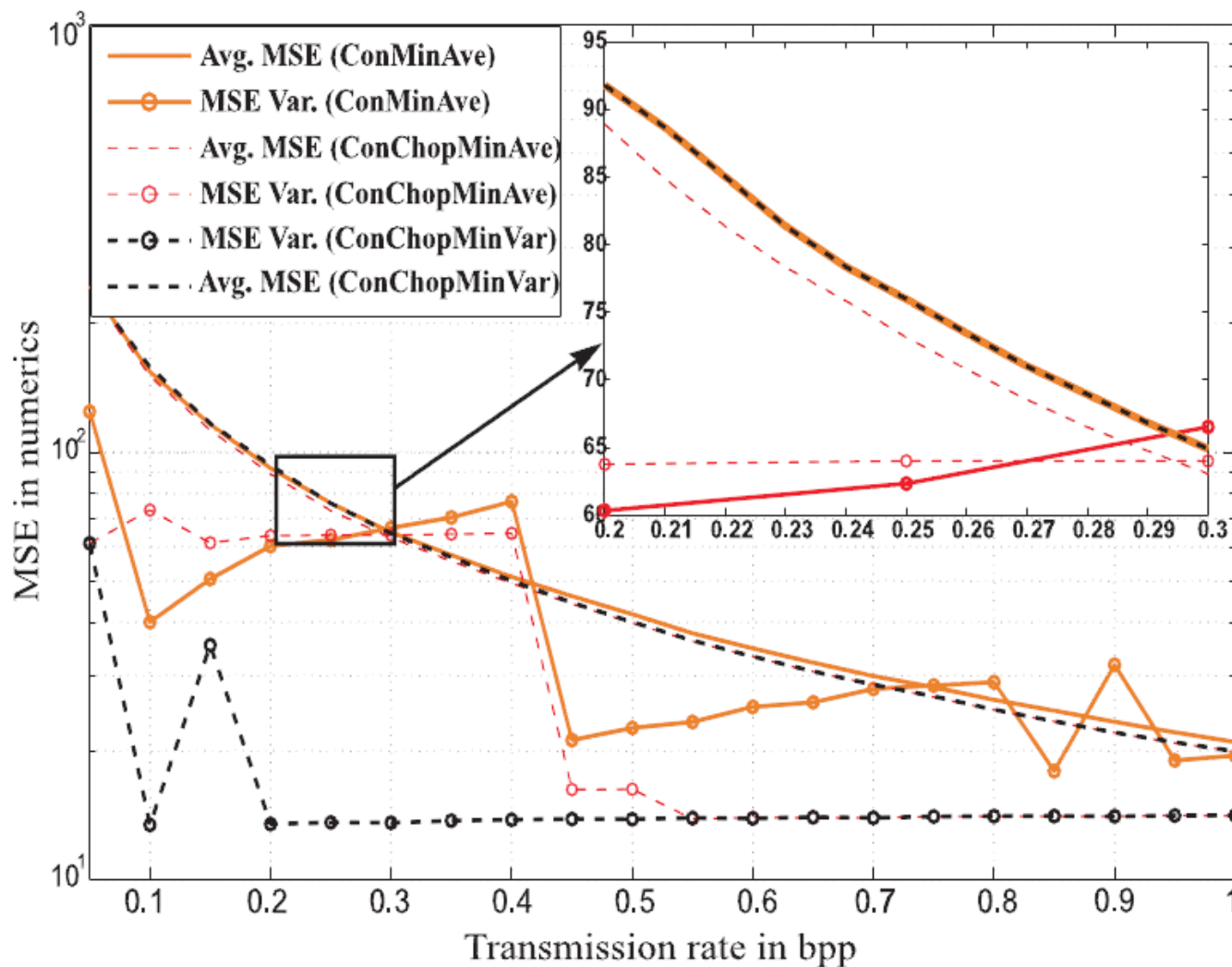
$$\min_{\pi} \bar{D}_{\pi}(2) \text{ subject to } r_{tr} = \frac{1}{N_s} \sum_{i=1}^M \frac{m_i v}{\prod_{j=i}^M r_{\pi}^{(j)}} \leq B$$

Numerical Results

- We compare the following systems:
 - **ConMinAve**: Concatenated block coding with minimum average distortion criterion. Let d^* be the minimum distortion. (**Original System [1]**)
 - **ConChopMinAve**: Extension scheme with minimum average distortion criterion.
 - **ConChopMinAve**: Extension scheme with minimum distortion variance criterion subject to a minimum average distortion constraint $\gamma_D \leq d^*$
- We use a 512 X 512 monochromatic images Lena and Goldhill using SPIHT and JPEG2000 compression algorithms.
- Let us set $v = 850$, $M = 2$, and use RCPC codes [1].
- A BSC with crossover probability $\varepsilon_0 = 0.05$.
- Our distortion metric is MSE and we present the mean MSE and MSE variance for all three systems.

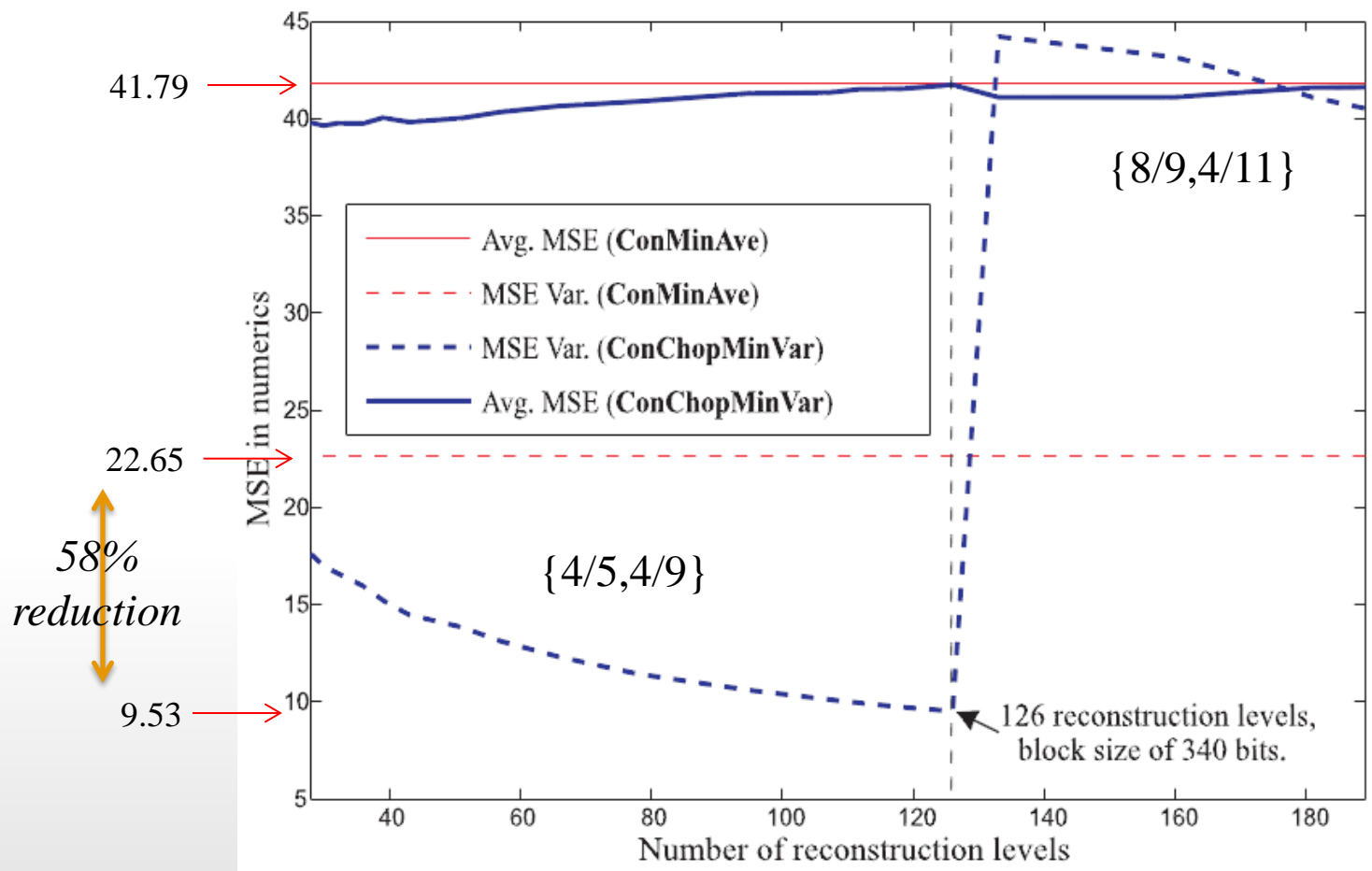
[1] S. S. Arslan, P. C. Cosman and L. B. Milstein, “Concatenated Block Codes for Unequal Error Protection of Embedded Bit Streams,” Submitted to IEEE Trans. on Image Processing.

Numerical Results



Numerical Results

- Let us vary v , to increase/decrease the number of reconstruction levels.
- Set $M = 2$.



Numerical Results

Image	$r_{tr}(\text{bpp})$	Results (Std. dev.)	Channel raw BER (ϵ_0)			Image	Channel raw BER (ϵ_0)	
			0.1	0.05	0.01		0.05	0.01
Lena (SPIHT)	0.25	ConMinAve	89.9	62.33	52.68	Goldhill (JPEG2000)	73.95	59.95
		ConChopMinVar	19.62	12.79	8.58		24.25	17.11
		Percentage decrease	78.17%	79.48%	83.71%		67.20%	71.46%
	0.5	ConMinAve	26.75	22.65	16.34		99.92	18.73
		ConChopMinVar	16.33	9.53	7.66		24.55	16.87
		Percentage decrease	38.95%	57.92%	53.12%		75.43%	9.93%
	0.8	ConMinAve	34.77	28.99	15.11		24.55	16.87
		ConChopMinVar	16.03	4.92	2.65		9.18	17.01
		Percentage decrease	53.9%	83.03%	82.46%		73.93%	30.29%

- Dramatic improvements can be obtained while maintaining the good mean distortion characteristics.
- Similar results can be observed using RC-LDPC codes.

References

- [1] A. Said and W. A. Pearlman, “A New Fast and Efficient Image Codec Based on Set Partitioning in Hierarchical Trees,” *IEEE Trans. on Circuits and Systems for Video Tech.*, vol. 6, pp.243-250, June 1996.
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- [6] B. A. Banister, B. Belzer, and T. R. Fisher, “Robust image transmission using JPEG2000 and turbo codes,” *IEEE Signal Process. Lett.*, vol. 9, no. 4, pp.117-119, Apr. 2002
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- [8] X. Pan, A. H. Banihashemi and A. Cuhadar, “Combined Source and Channel Coding With JPEG2000 and Rate-Compatible Low-Density Parity-Check Codes,” *IEEE Trans. on Signal Processing*, vol. 54, no 3, pp.1160 - 1164, March 2006.