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# Coding for the Network: Scalable and Multiple description coding

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# Overview

- **Examples and motivations**
- **Scalable coding for network transmission**
- **Techniques for multiple description coding**



# Coding for the Network

- **Traditional compression techniques:**  
**Minimum size for a given quality**
  
- **Assumptions:**
  - Known bandwidth
  - Known decoding parameters (resolution, frame-rate)
  - No errors
  
- **Network video delivery**
  - Different links and users
  - Unreliable links



# Coding for the Network

- **Basic idea: from a monolithic representation to a structured representation**
- **Layered (i.e. hierarchical) sub-streams: scalable video coding (SVC)**
- **Equal importance sub-streams: multiple description coding (MDC)**

# Image transmission

1. Sequential coding
2. Progressive coding (scalable)
3. Multiple description (MDC)
  - 5 packets
  - 2 cases:
    - no error
    - packet 2/5 not received



# Examples: transmission without errors

Packet 1



Non-scalable

Scalable

Multiple descriptions

# Examples: transmission without errors

Packet 2



Non-scalable



Scalable



Multiple descriptions

# Examples: transmission without errors

Packet 3



Non-scalable

Scalable

Multiple descriptions

# Examples: transmission without errors

Packet 4



Non-scalable

Scalable

Multiple descriptions

# Examples: transmission without errors

Packet 5



Non-scalable

Scalable

Multiple descriptions

# Examples: transmission with errors

Packet 1



Non-scalable

Scalable

Multiple descriptions

# Examples: transmission with errors

Packet 2: **LOSS**



Non-scalable



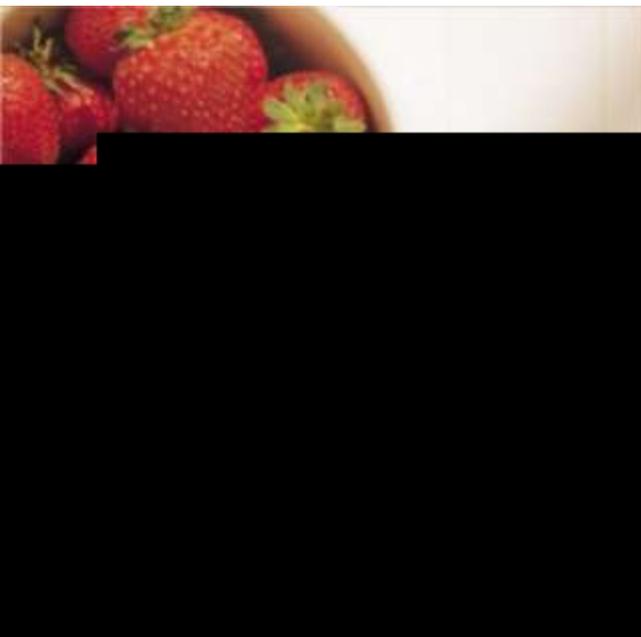
Scalable



Multiple descriptions

# Examples: transmission with errors

Packet 3



Non-scalable



Scalable



Multiple descriptions

# Examples: transmission with errors

Packet 4



Non-scalable



Scalable



Multiple descriptions

# Examples: transmission with errors

Packet 5

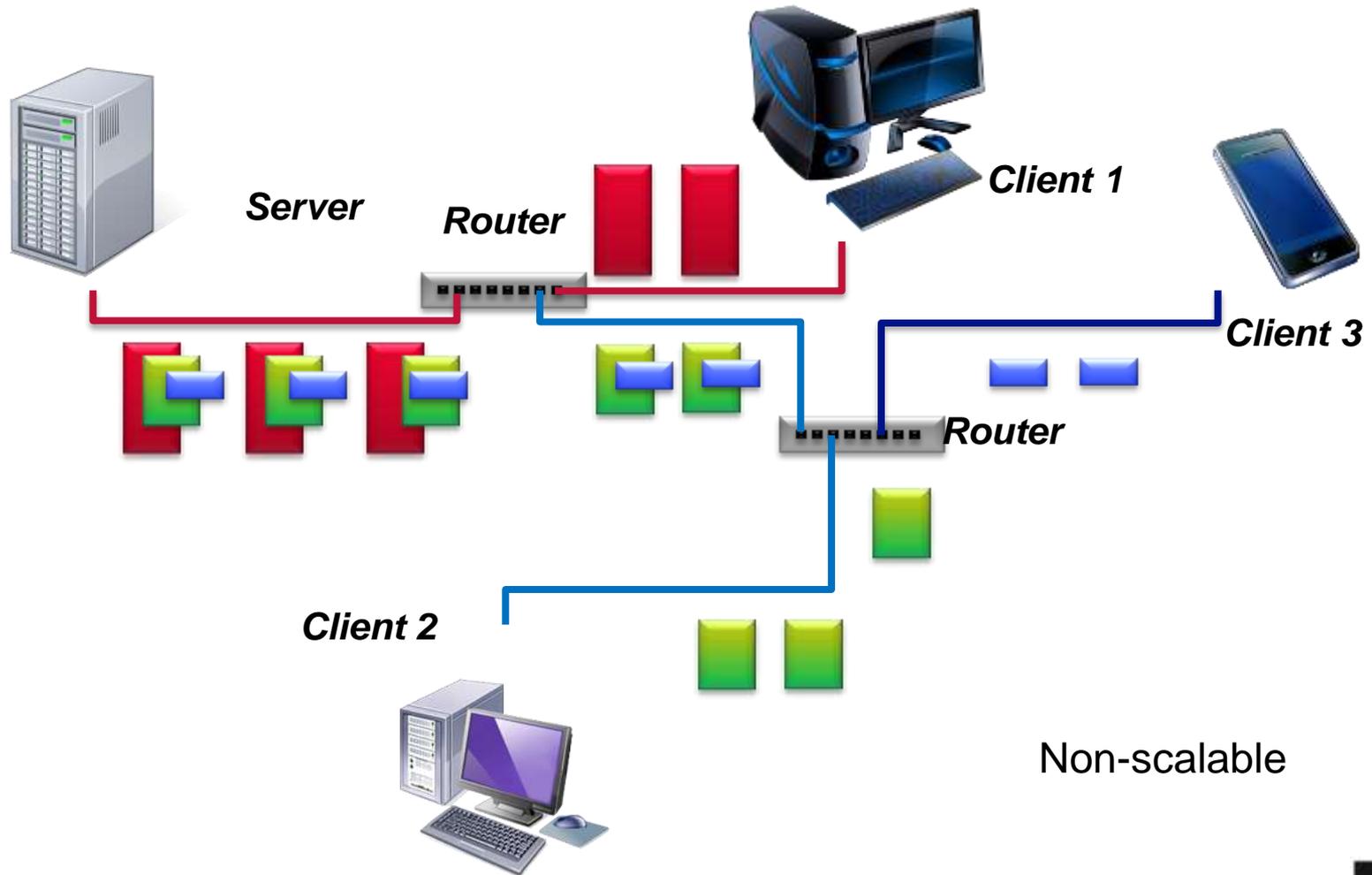


Non-scalable

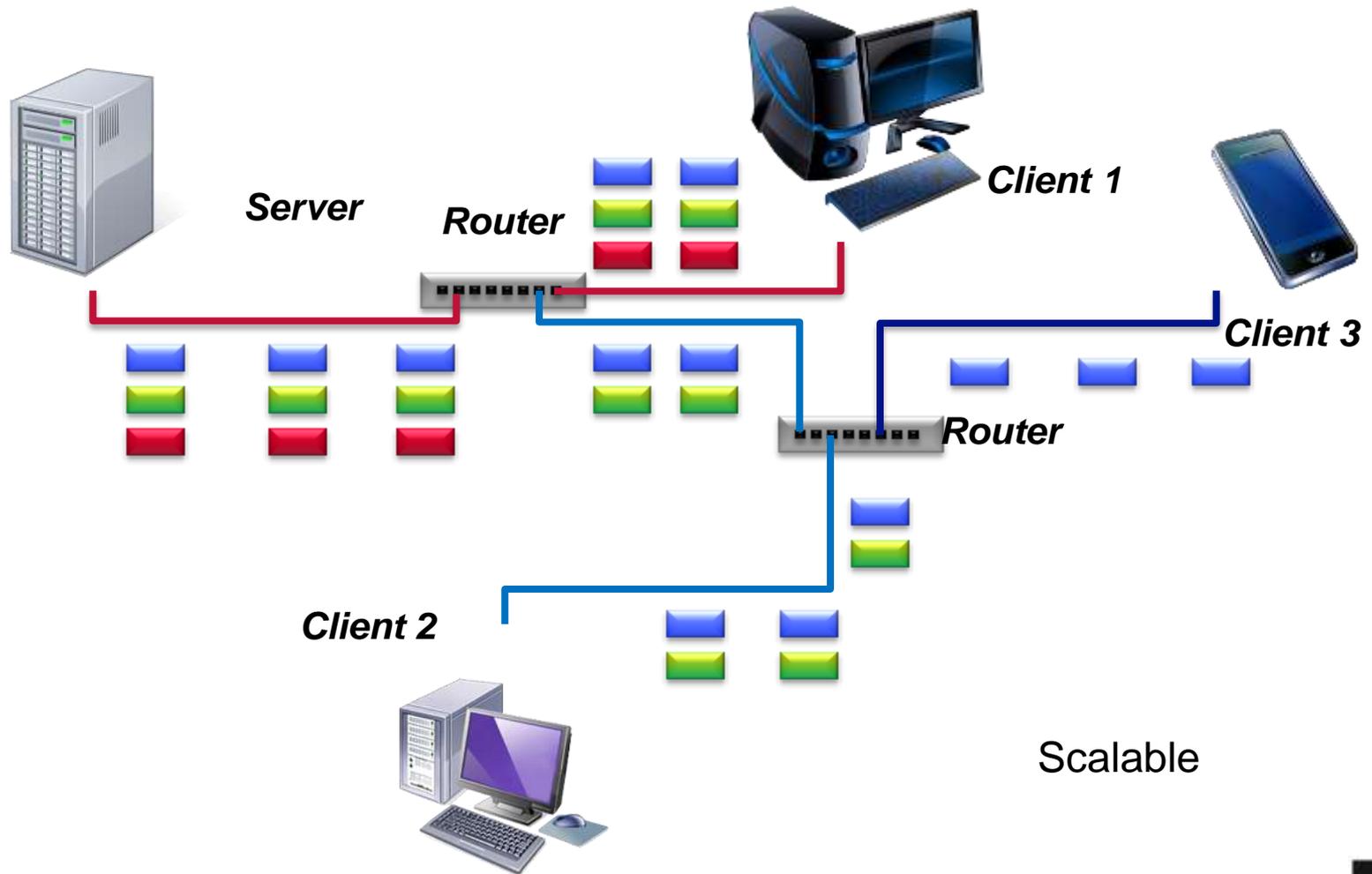
Scalable

Multiple descriptions

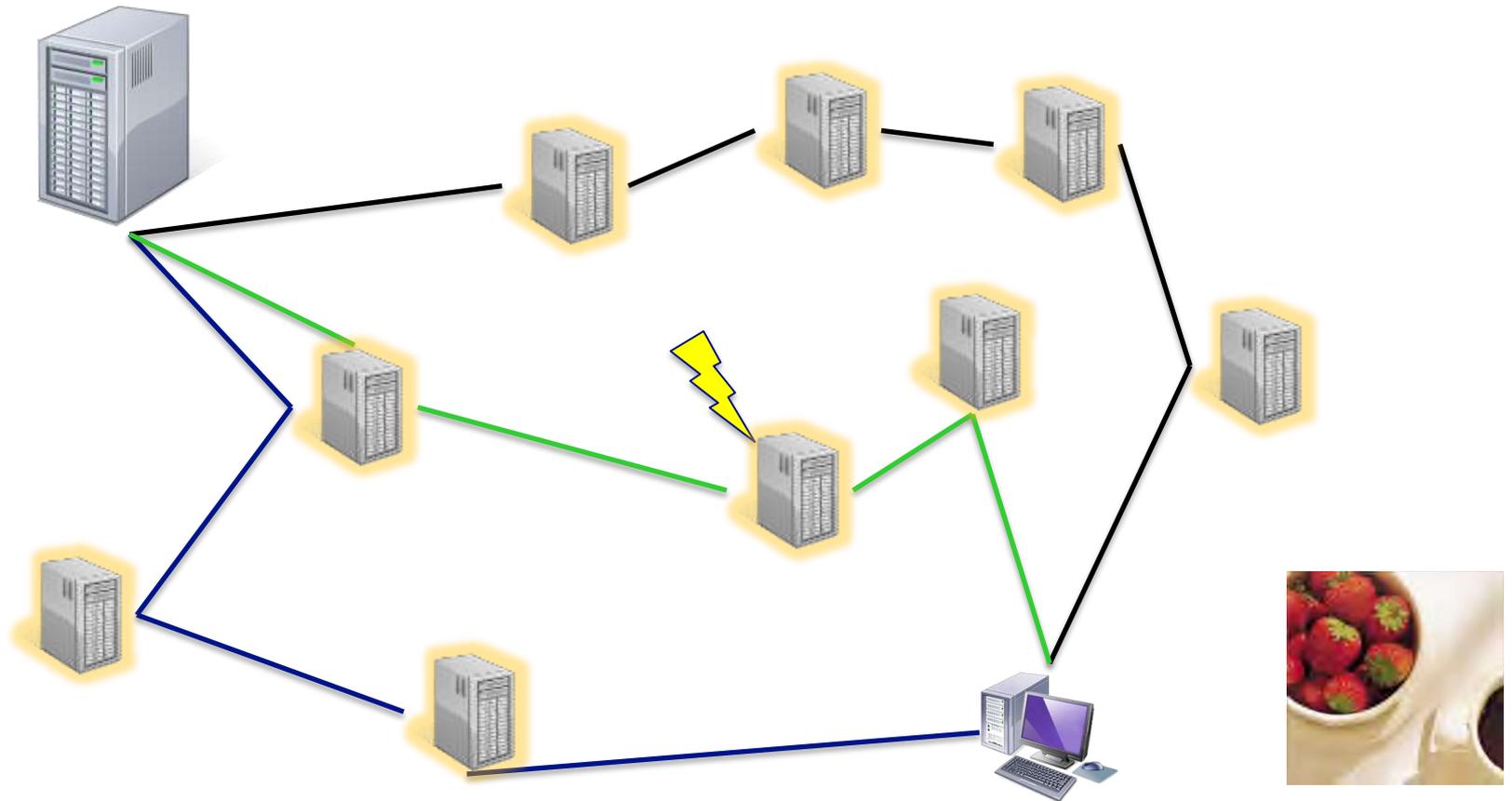
# Examples: heterogeneous users and networks



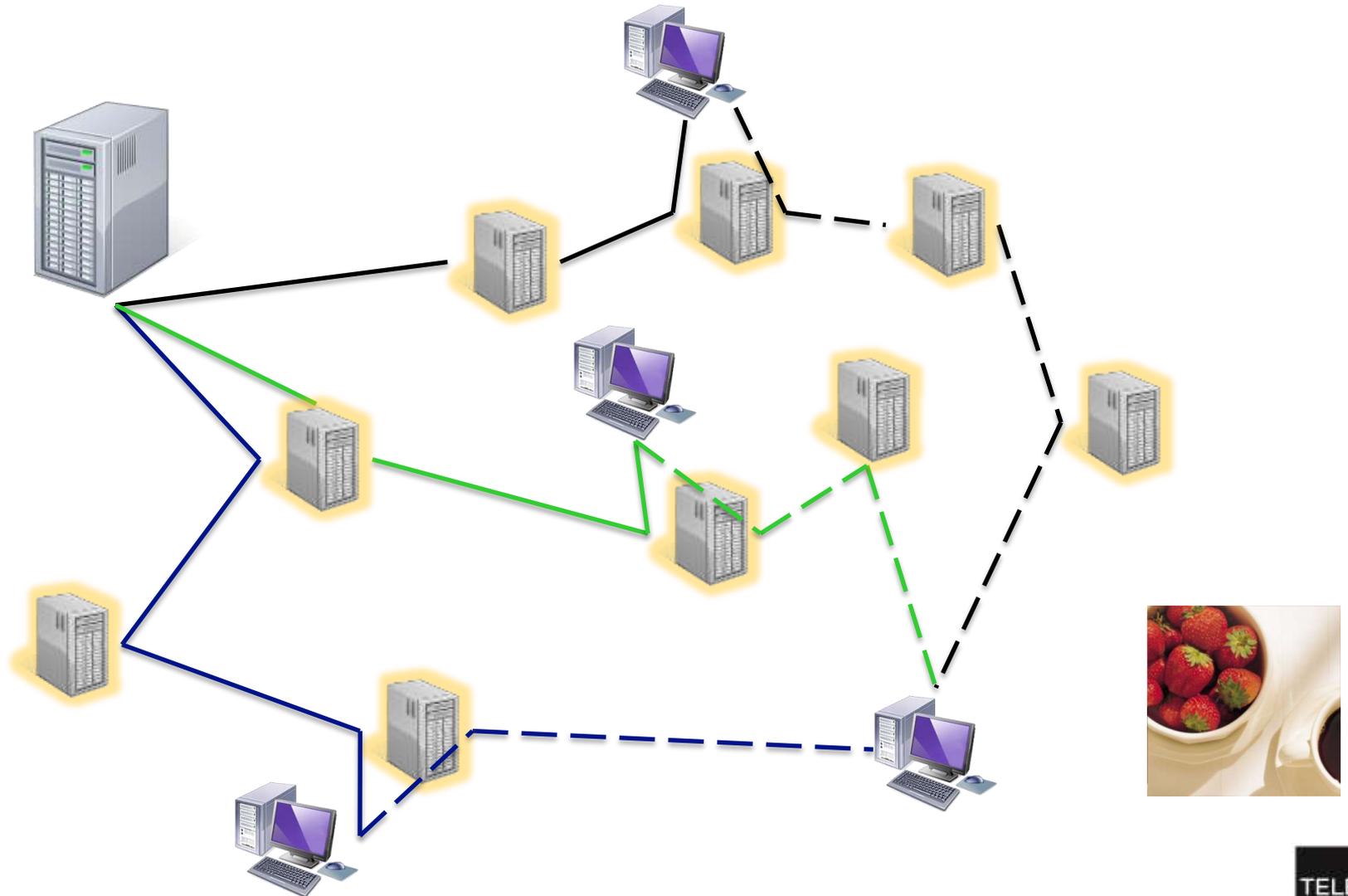
# Examples: efficient network use



# Multiple Description Coding



# Multiple Description Coding + P2P





# Overview

- **Examples and motivations**
- **Scalable coding for network transmission**
  - Principles and applications
- **Techniques for multiple description coding**



# SVC approach

## ■ Encode once

- Different versions obtained by combining sub-streams

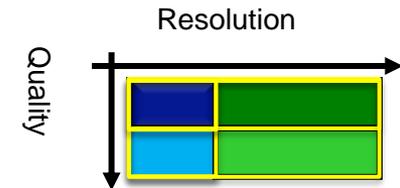
## ■ Intelligent network (multiple multicast groups)

- Packet replication only where needed
- Efficient network use

# Scalability in video coding

**Definition: A scalable video stream is a compressed representation of a video such that:**

- The representation is made up of *layers*
- Layers provide *incremental* refinement of the sequence
- The representation is *efficient* in term of reconstructed image quality for a given rate





# Kinds of scalability

- **SNR scalability**
  - aka quality, bit-rate scalability
- **Space scalability**
  - aka resolution scalability
- **Time scalability**
  - aka frame-rate scalability
- **Other (object scalability, ...)**



## Scalability: pros...

- **Encode once, decode many!**
- **Optimal use of the network**
- **Space saving on the server**
- **Each user is provided exactly with the demanded information**
- **Error resilience (but what if base layer is lost?)**
- **Dynamic decoding rate**



## Scalability: cons

- **Complexity increases**
- **Rate-distortion performances are degraded**
  - This goes worse and worse when more layers are added
  - Is it better to degrade base layer or details?
- **Fully useful only if the infrastructure is changed**
  - Multicast routers are needed



## Scalability: issues

- How can we obtain quality, space and time scalability?
- How can we achieve scalability with low impact on rate-distortion performance?
- How can we achieve scalability with low complexity increase?
- How can we trade-off between base layer and details degradation?



# Overview

- **Examples and motivations**
- **Scalable coding for network transmission**
- **Techniques for multiple description coding**
  - Principles
  - MDC techniques

# MDC vs. SVC

- **Scalable coding: base + enhancement**
  - If base is lost, enhancements are useless
  - TCP-based transmission: all packets arrive...
  - ... but retransmission delay is usually much higher than packet interarrival
- **In conclusion, SVC relies too much on transport layer, which could fail for several reasons**
- **A different transmission model could help when transmission is not reliable and delay is a constraint**

# Multiple description coding: principles

- **The encoder creates 2 description of the same importance**
  - Trivial example: odd and even frames
- **Descriptions sent over different (logical) channels**
- **One description is enough for acceptable quality**
- **Receiving 2 descriptions provides the best quality**
- **Generalization:  $N$  descriptions**

# MDC: pros and cons

## Pros

### ■ Good quality without retransmission

- Real-time and interactive applications
- Simplification of network design: no feedback channel needed

### ■ Traffic dispersion

- Better than scalable coding since all streams have the same importance
- Application to P2P distribution using multiple paths

## Cons

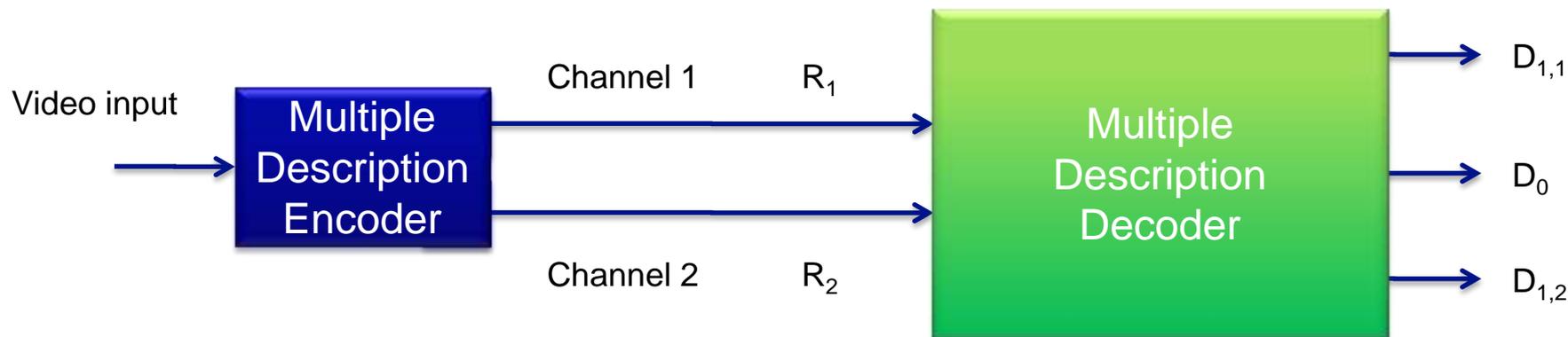
### ■ Excess rate: for the same quality, MDC demands more bits wrt single description coding

# MDC: conceptual scheme

Usually a balanced scheme  
is considered:

$$R_1 = R_2 = R/2$$

$$D_{1,1} = D_{1,2} = D_1$$



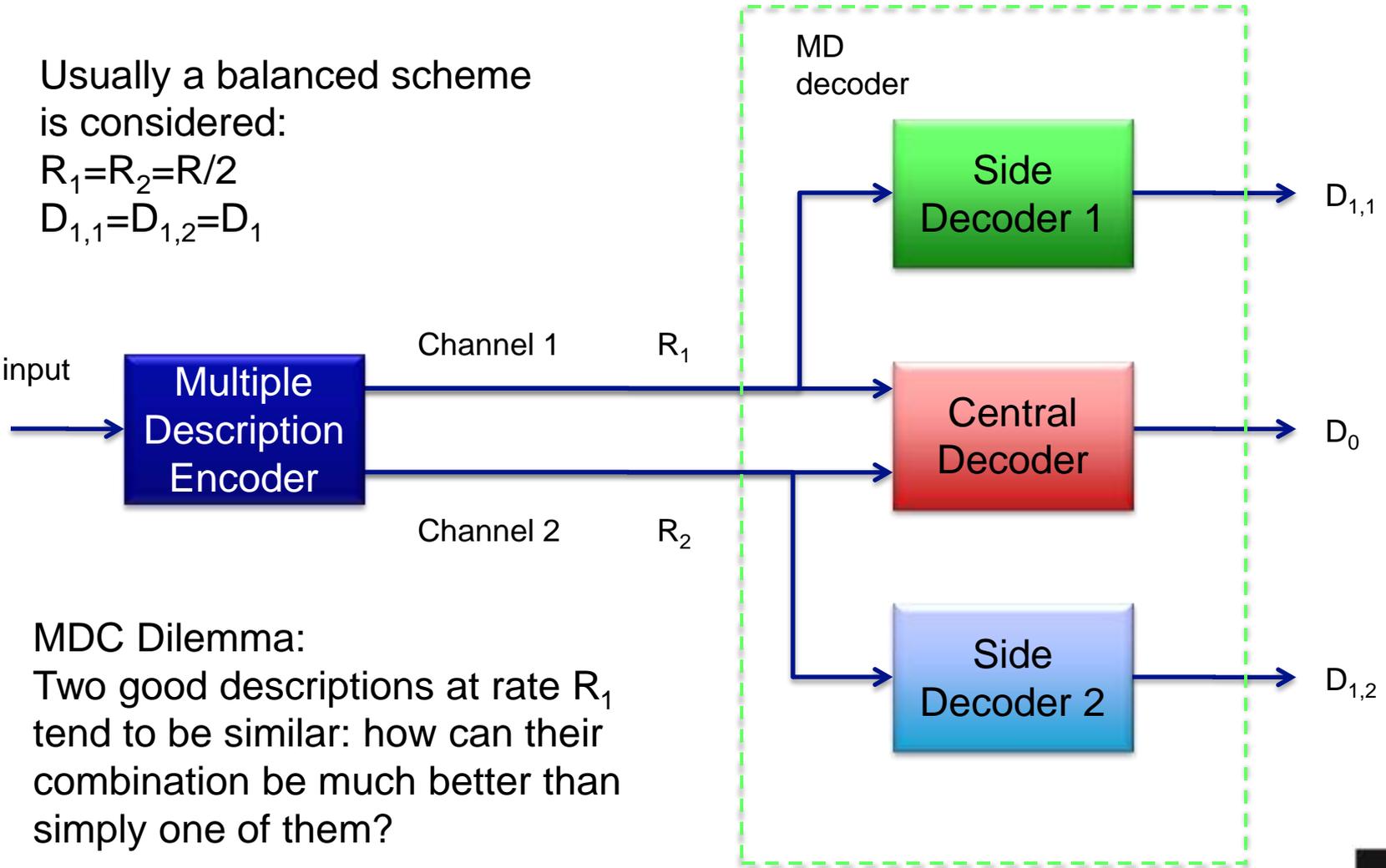
# MDC: conceptual scheme

Usually a balanced scheme is considered:

$$R_1 = R_2 = R/2$$

$$D_{1,1} = D_{1,2} = D_1$$

Video input



MDC Dilemma:

Two good descriptions at rate  $R_1$  tend to be similar: how can their combination be much better than simply one of them?

## MDC problems

- For SDC, one has to minimize  $D_0$  for a given  $R$
- For MDC, we consider the redundancy  $\rho(D_0, D_1)$ 
  - $\rho(D_0, D_1) = R - R^*$
  - $R = R_1 + R_2$ : rate for MDC to assure distortion  $D_0$  for the central decoder and  $D_1$  for the side decoders
  - $R^*$ : rate for the SDC to assure distortion  $D_0$

# MDC problems

## ■ Distortion bounds

- If  $\rho \rightarrow 0^+$ , then  $\frac{\partial D_1}{\partial \rho} \rightarrow +\infty$
- If the distortion bound is achieved for central decoder, use the additional rate for side decoder

## ■ Another approach uses the error probability $p$ :

$$J = (1 - p)^2 D_0 + 2p(1 - p)D_1 + \lambda R$$

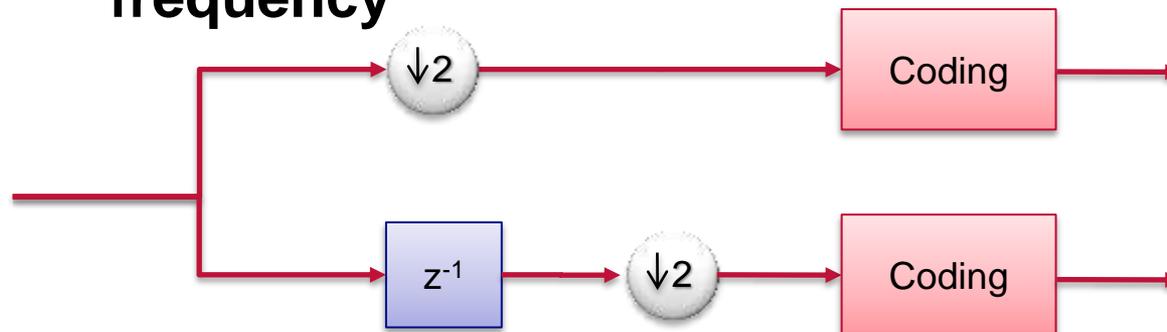


# Overview

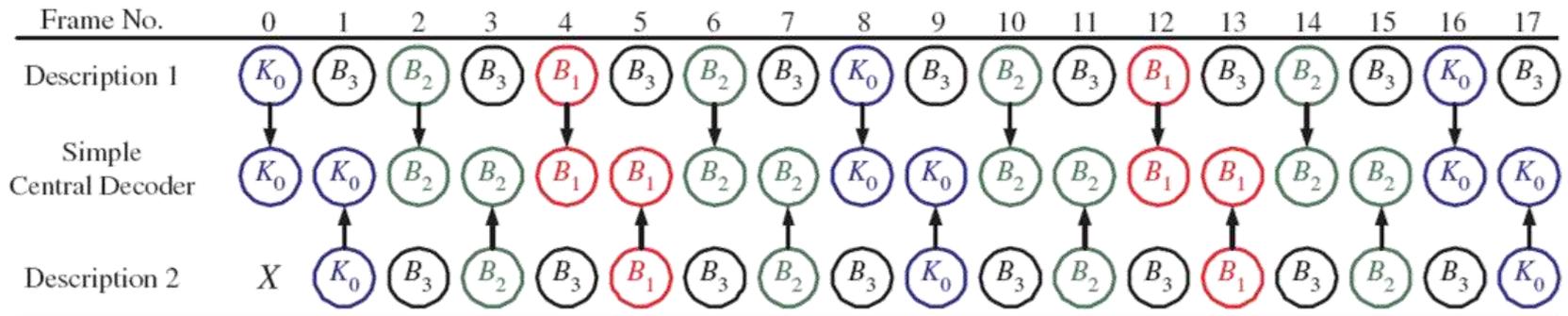
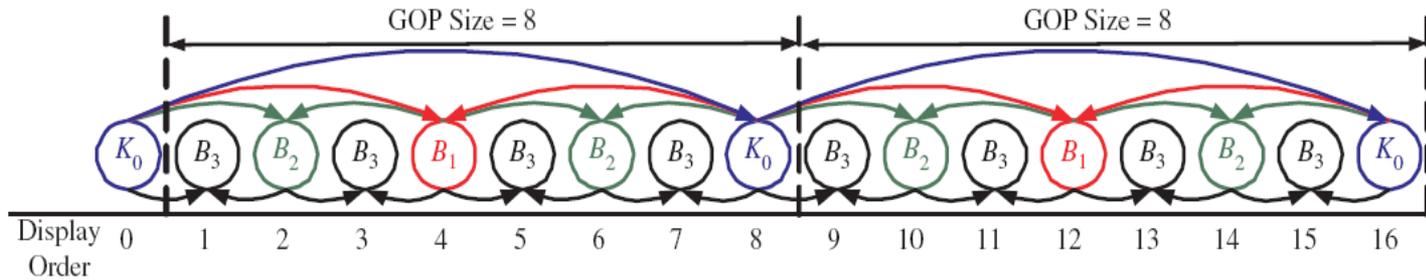
- **Examples and motivations**
- **Techniques for scalable coding**
- **Techniques for multiple description coding**
  - Principles
  - MDC techniques

# MDC techniques: Channel splitting

- Video data are split into 2 (or more channels) and processed separately
- Channel splitting relies only on channel redundancy
- The splitting can be performed in space, time, frequency

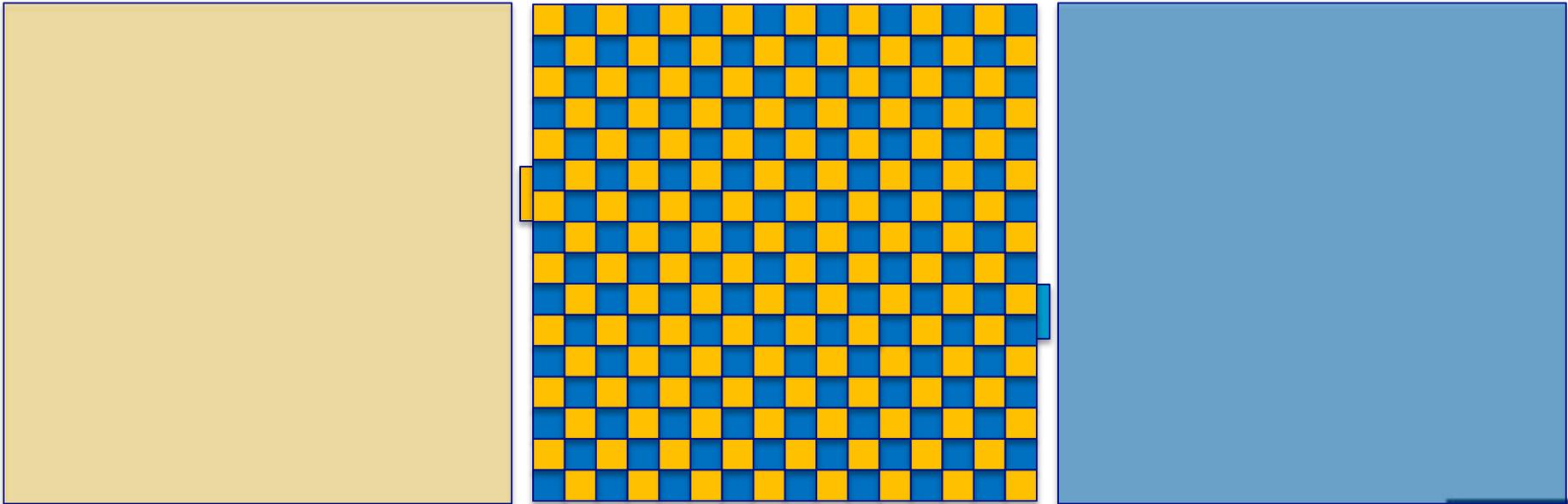


# Temporal Channel splitting: An example



## Spatial channel splitting: FMO in H.264

- Images divided into slices
- Slices coded independently
- FMO allows to define arbitrarily shaped slices
- Slices correspond to descriptions



## Channel splitting: example

- AR model for vocal signal

$$X_n = aX_{n-1} + Y_n$$

$$Y_n = N(0, \sigma) \text{i. i. d.}, a \in ]0, 1[$$

- Filter characterization ?

## Channel splitting: example

- AR model for vocal signal

$$X_n = aX_{n-1} + Y_n$$

$$Y_n = N(0, \sigma) \text{ i. i. d. , } a \in ]0, 1[$$

- Filter characterization ?

$$X(z) = az^{-1}X(z) + Y(z)$$

$$X_n = \sum_{m=0}^{+\infty} a^m Y_{n-m}$$

- Statistical characterization of  $X_n$  ?

## Channel splitting: example

- AR model for vocal signal

$$X_n = aX_{n-1} + Y_n$$

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- Filter characterization ?

$$X(z) = az^{-1}X(z) + Y(z)$$

$$X_n = \sum_{m=0}^{+\infty} a^m Y_{n-m}$$

- Statistical characterization of  $X_n$ ? Zero-mean gaussian

$$r_{xx}(n) = \frac{\sigma^2 a^{|n|}}{1 - a^2}$$

For simplicity, we set  $\sigma^2 = 1 - a^2 \rightarrow r_{xx}(k) = a^{|k|}$

## Channel splitting: example

### ■ AR model for vocal signal

$$X_n = aX_{n-1} + Y_n$$

$$Y_n = N(0, \sigma) \text{ i. i. d.}, a \in ]0, 1[$$

$$X_n = N(0, 1), \quad r_{xx}(k) = a^{|k|}$$

### ■ The larger $a$ , the more correlated the samples of $X$

### ■ Now we want to encode $X$

- Single description : DPCM coding
- Multiple description: Channel splitting with DPCM on each channel and linear interpolation at side decoders

## Channel splitting: example

DPCM coding for AR Gaussian signal with correlation  $a$

$$\tilde{x}_n = aX_{n-1}$$

$$E_n = Y_n$$

$$D_{\text{PCM}} = \sigma^2 2^{-2R} = (1 - a^2) 2^{-2R}$$

- For simplicity, we suppose  $a$  perfectly known, and we ignore the quantization error of the predictor (high-rate hypothesis)
- The formula for the distortion holds for the single description case, but also for the encoding of any AR Gaussian signal with correlation  $a$

# Channel splitting: example

## ■ Channel splitting

$$X_e(n) = X(2n)$$

$$X_o(n) = X(2n + 1)$$

$$r_{x_e x_e}(k) = ?$$

# Channel splitting: example

- Channel splitting

$$X_e(n) = X(2n)$$

$$X_o(n) = X(2n + 1)$$

$$r_{x_e x_e}(k) = E[X(2n)X(2n + 2k)] = r_{xx}(2k) = a^{2|k|}$$

- On each channel we have an AR Gaussian signal with correlation  $a^2$
- We can use the previous formula for distortion

## Channel splitting: example

### Central decoder

- On both channel we receive an AR Gaussian signal

$$D_o = D_e = (1 - a^4)2^{-2R}$$

- The distortion (called full-rate distortion) is the average of the distortions on the two channels

$$D_F = \frac{1}{2}D_o + \frac{1}{2}D_e = (1 - a^4)2^{-2R}$$

- The central decoder has a constant factor increase in distortion (at high rate):

$$\frac{D_F}{D_{PCM}} = (1 + a^2) > 1$$

- The more correlated are the samples, the larger losses will be caused by channel splitting

## Channel splitting: example

- Side decoder (half-rate decoder)
- We only receive even samples, therefore:  
 $D_e = (1 - a^4)2^{-2R}$
- For odd samples: the quantization distortion (modulated by a term depending on quantization noise correlation,  $\omega \in ]0, 1[$ ) *plus* the interpolation error

$$D_o = \omega D_e + (1 - a)^2 + \frac{1}{2}(1 - a^2)$$

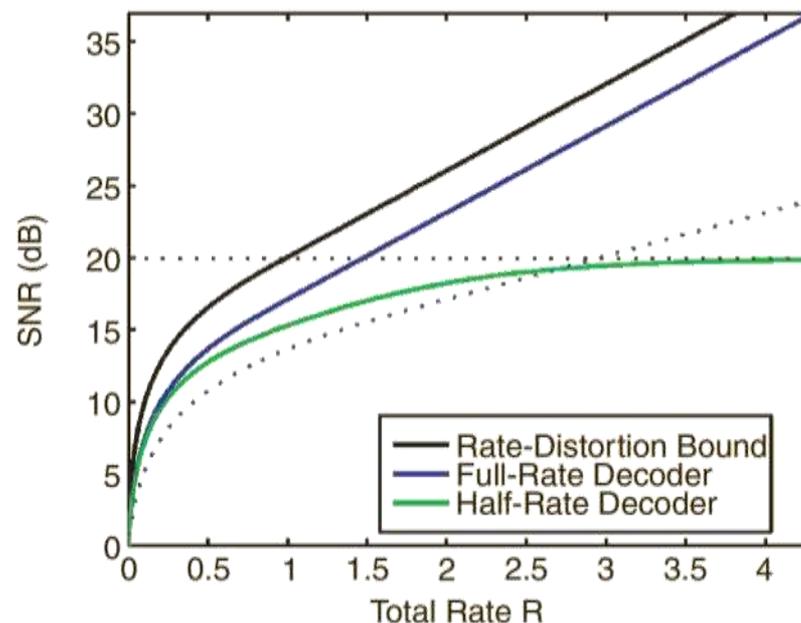
$$D_H = \frac{1}{2}D_o + \frac{1}{2}D_e = \frac{1}{2} \left[ (1 - a)^2 + \frac{1}{2}(1 - a^2) \right] + \frac{1+\omega}{2} D_F$$

- In the worst case ( $\omega = 1$ ),  $D_H = D_F + k_a$
- Actually at low bit rate we can have  $D_H < D_F$

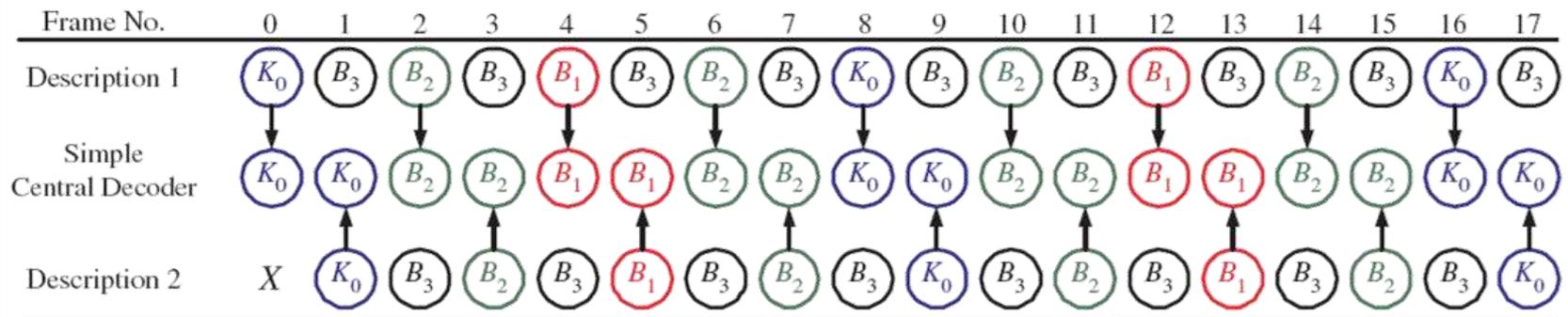
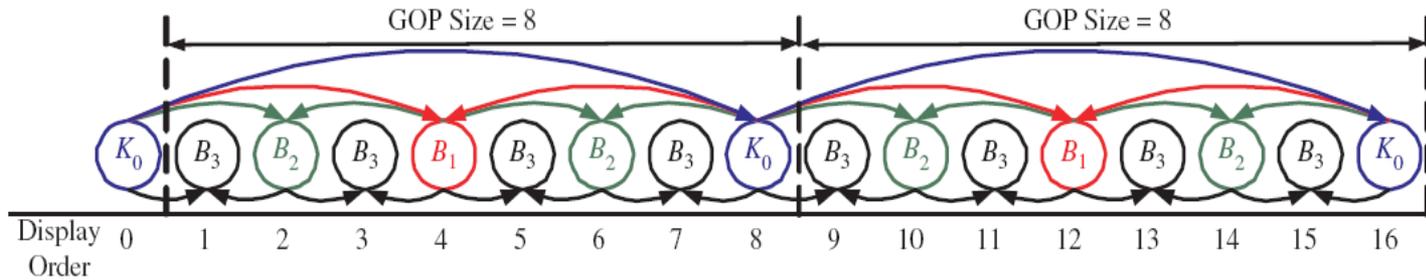
# Channel splitting: example

## Summary:

- Central decoder: constant loss ( in dB) wrt single description coding
- The side decoder performance saturates since it cannot get rid of interpolation error, no matter how high the rate is
- At low bit-rate, the side decoder can outperform the central one: our approximation do not hold longer; it could be better to use the few bits we have to encode pretty one only the even samples and interpolate the odd ones
- Results are coherent with experimentation in audio coding (see figure)

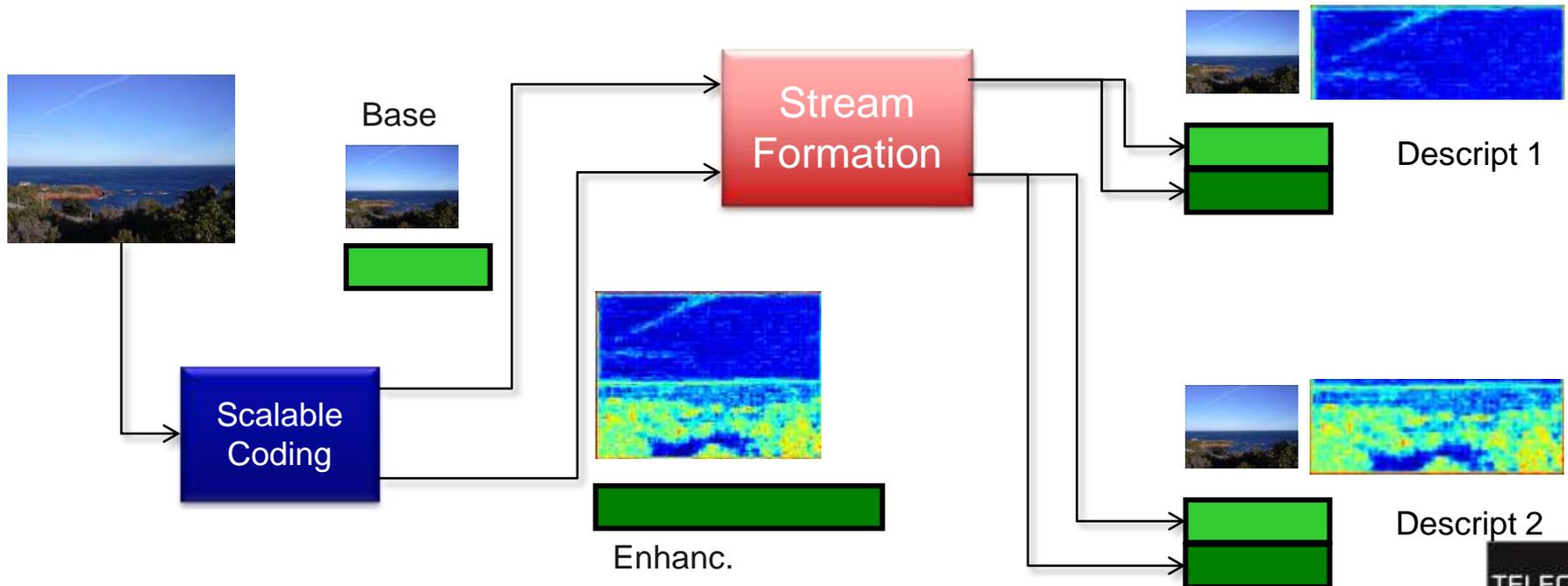


# Channel splitting: an example for video coding



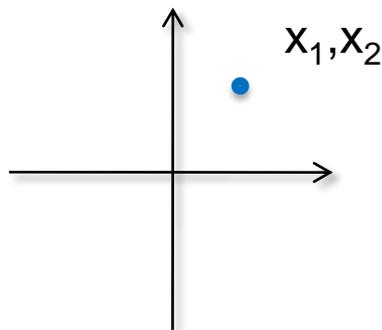
# MDC techniques: Unequal Error Protection

- Produce a two-layers progressive representation of data
- Insert the base layer in both descriptions
- Split the enhancement layer in the two descriptions

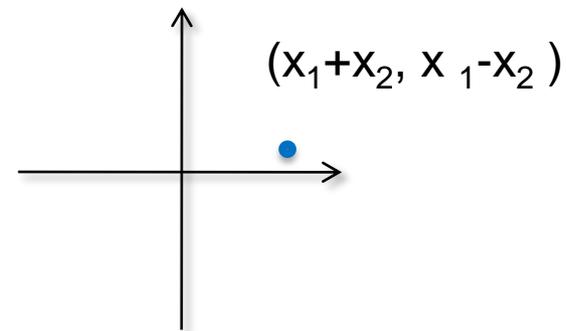


# MDC techniques: Correlating transforms

- Keep some correlation among coefficients
- Statistical dependencies are used to estimate transform coefficients that are in a lost description



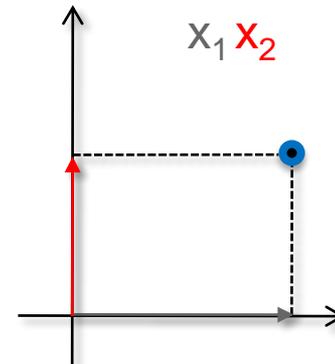
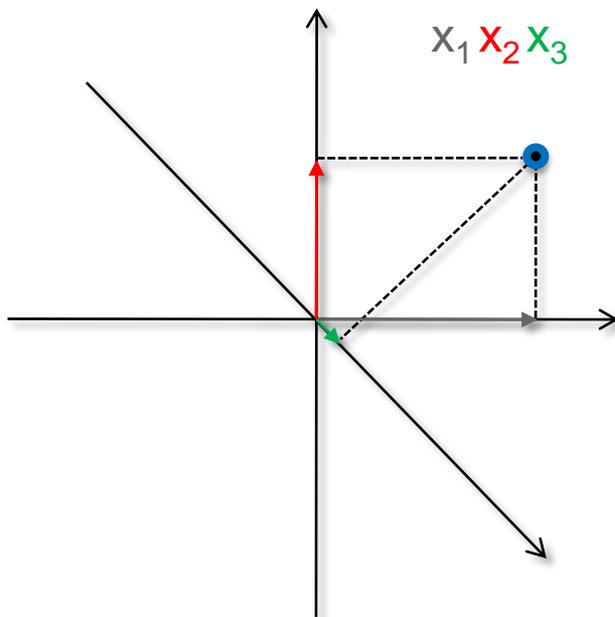
$X_1, X_2$ : Independent  
Gaussian variables



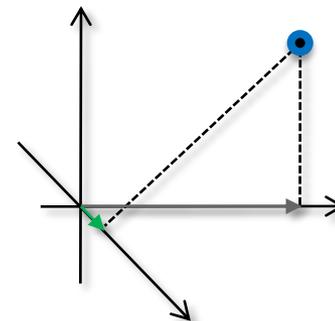
$Y_1, Y_2$ : Correlated  
Gaussian variables

# MDC techniques: Redundant transforms

- Project the input vector into a redundant set
- Use subsets of coefficients to form descriptions



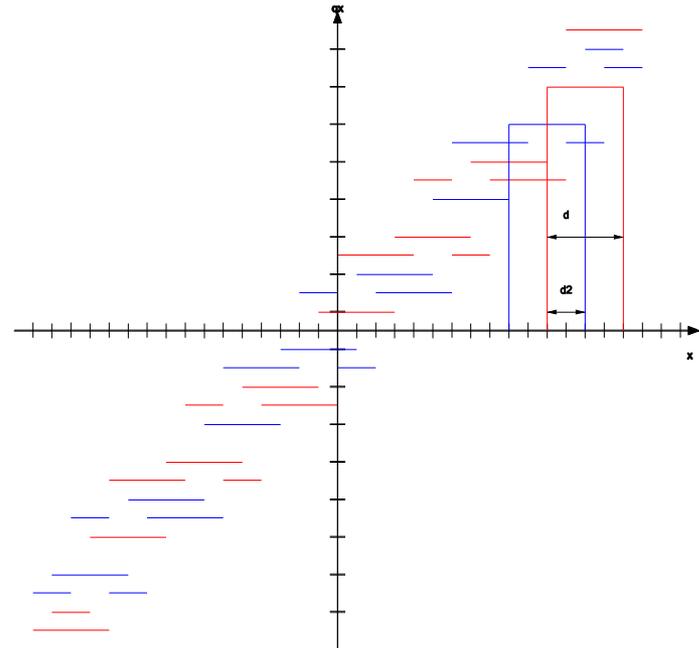
Description 1



Description 2

# Other MDC approaches

- **Shifted intervals**
- **Non-convex quantizers**
  - One index corresponds to disjoint intervals
- **Oversampled filter banks**
  - Related to wavelets
  - Scalability+MDC





## Mismatch for video MDC

- **Mismatch (i.e. drift) if prediction is made with information not available at the decoder**
- **Mismatch is probable because of packet loss**
- **No mismatch: improves side decoders**
- **Mismatch: improves central decoder**



# Conclusions

- **Traditional approaches unfit to video delivery over networks**
- **Basic idea: split the stream into substreams**
  - Scalable coding: good RD performances, good use of resources (memory, complexity, bandwidth), some error resilience
  - MDC: acceptable RD performances, quite good use of resources, good error resilience
- **Cons:**
  - Compression capability
  - Complexity

# Bibliography

## Scalable video coding

- [1] W. Li, “Overview of Fine Granularity Scalability in MPEG-4 Video Standard”. IEEE Trans. Circ. and Syst. for Video Tech., 11(3) Mar. 2001.
- [2] J.-R. Ohm, “Advances in Scalable Video Coding”. Proceedings of the IEEE, 93(1), Jan. 2005.
- [3] H. Schwarz, D. Marpe, T. Wiegand, “Overview of the Scalable Video Coding Extension of the H.264/AVC Standard”. IEEE Trans. Circ. and Syst. for Video Tech., 17(9), Sept. 2007

## Multiple Description Coding

- [4] V. Goyal, “Multiple Description Coding: Compression Meets the Network”. IEEE Sign. Proc. Magazine, Sept. 2001
- [5] Y. Wang, A. R. Reibman, S. Lin, “Multiple Description Coding for Video Delivery”. Proceedings of the IEEE, 93(1), Jan. 2005.