

# Ph.D Defense

## Optimal Cross-Layer Resource Allocation for Real-Time Video Transmission over Packet Lossy Networks

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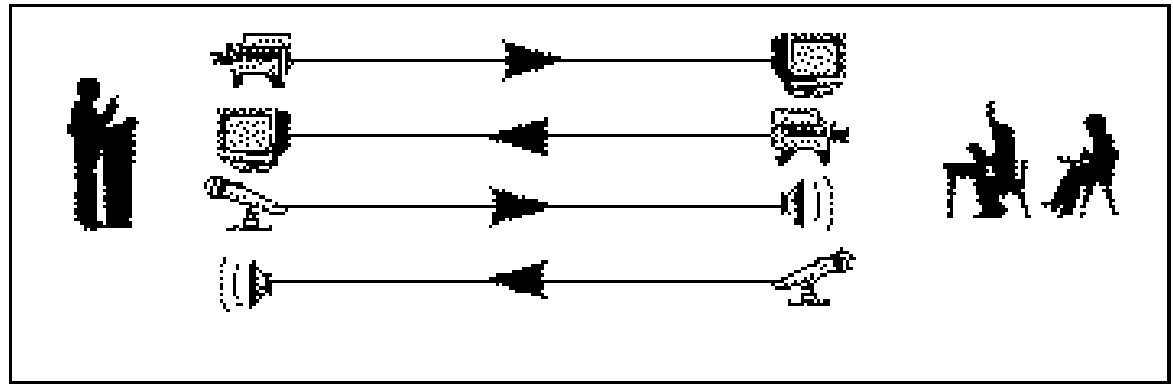
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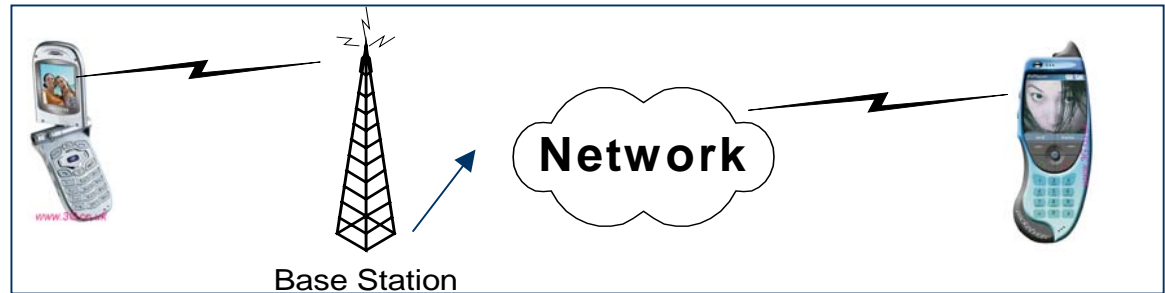
Feb. 9, 2004

# Current Streaming Video Applications

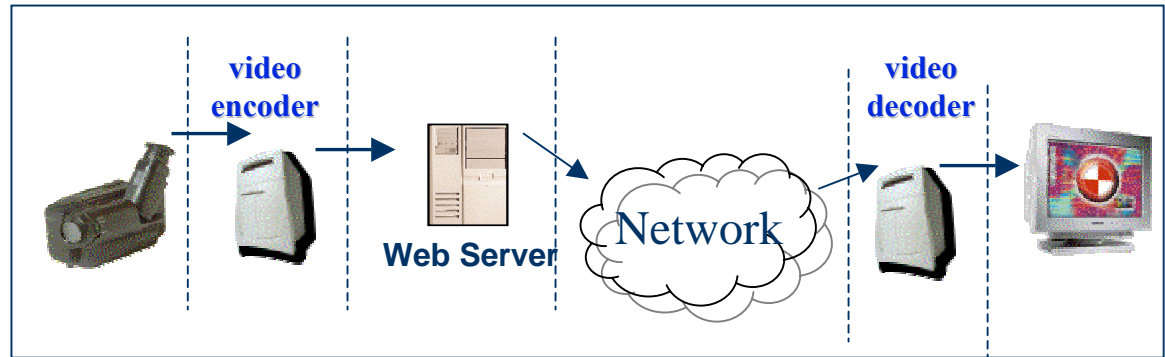
- Videoconferencing
- Distance learning



- Videophone

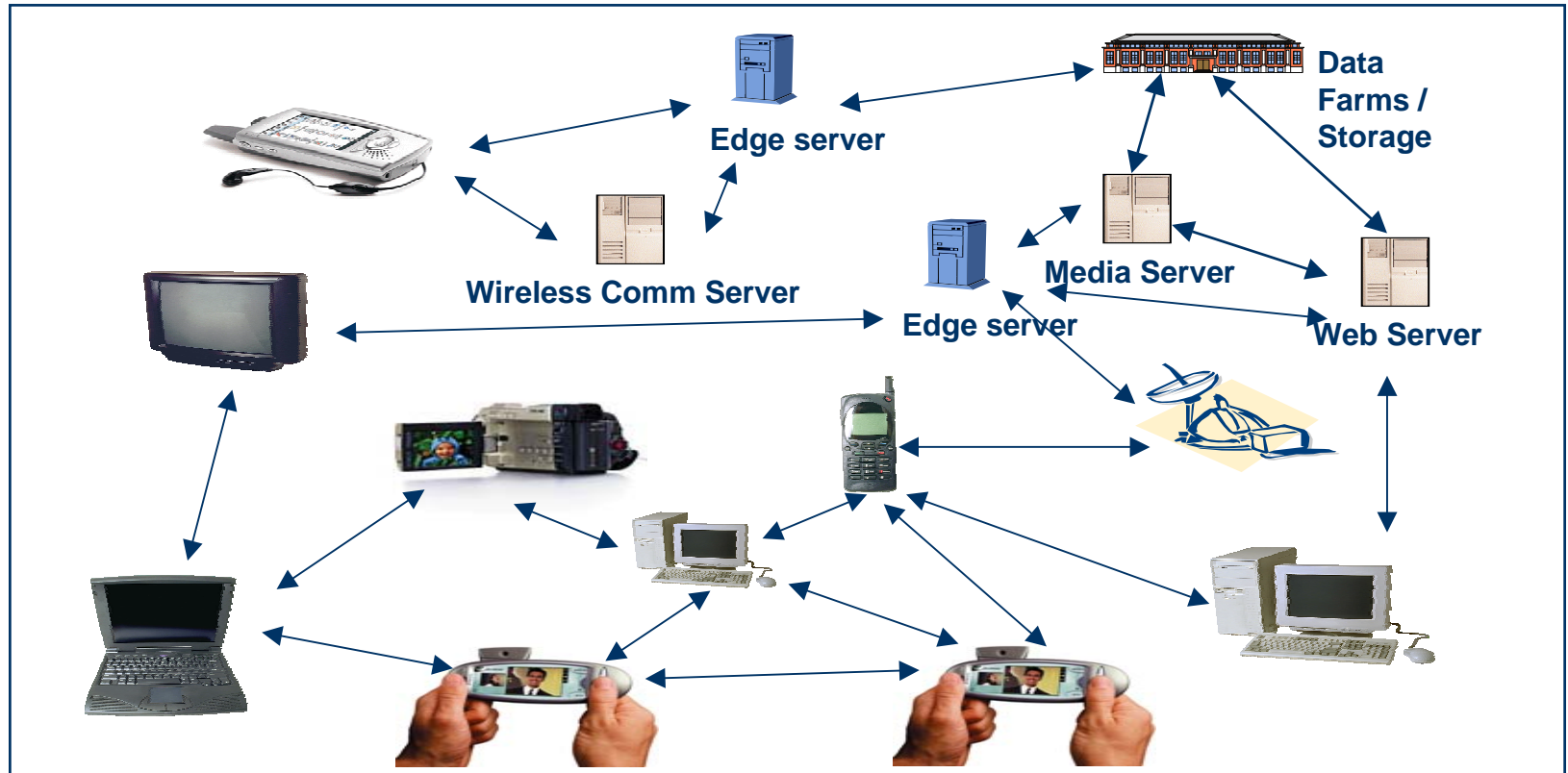


- On-demand streaming



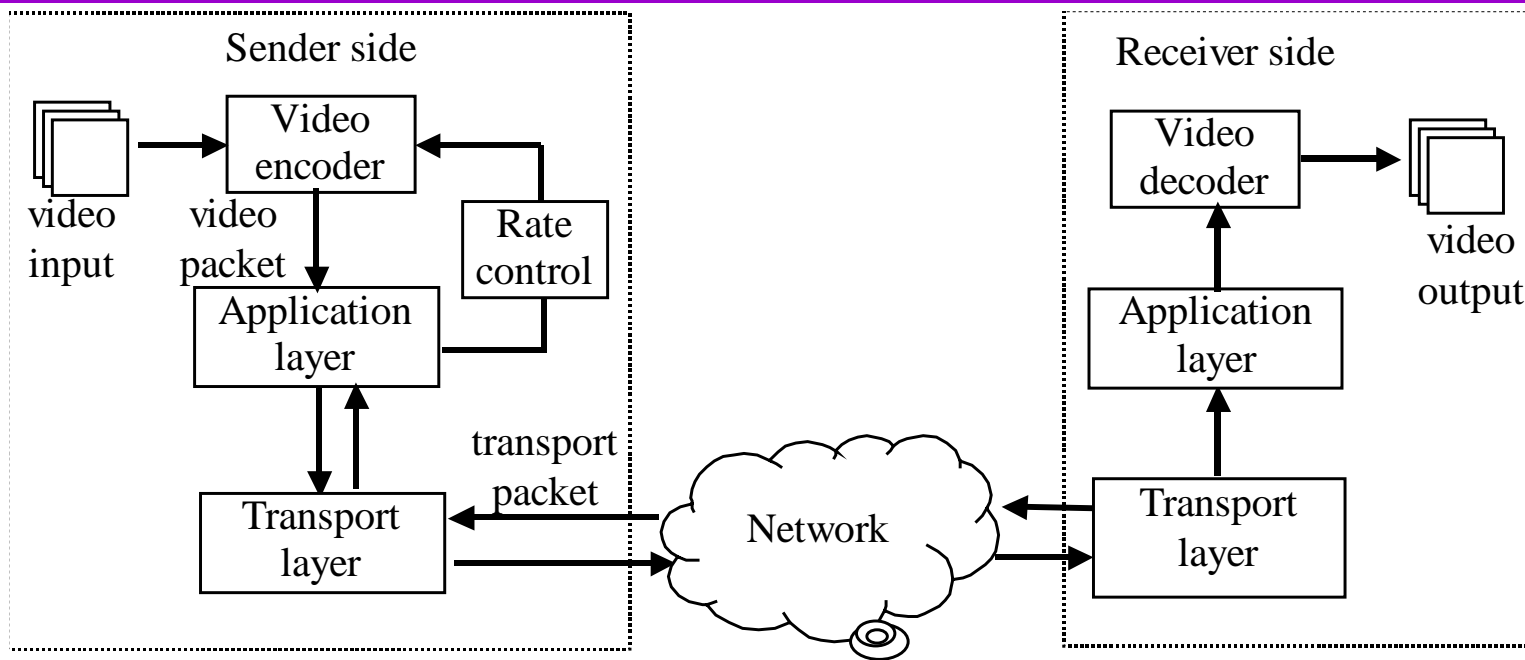
# Emerging Streaming Video Applications

*Anywhere, Anytime and Anyone*



- Cellular mobile networks: 2.5G, 3G, 4G systems
- Wireless LANs: IEEE 802.11, Hiper LAN 2, Bluetooth

# Video Transmission System



- Encoder: compression
- Rate control: to constraint bit rate based on estimated CSI
- Protocol stack: RTP/UDP/IP
- Network: packet loss and delay
- Decoder: display video in real-time; error concealment
- Max end-to-end delay (setup time): application-dependent

# Challenges and Approaches

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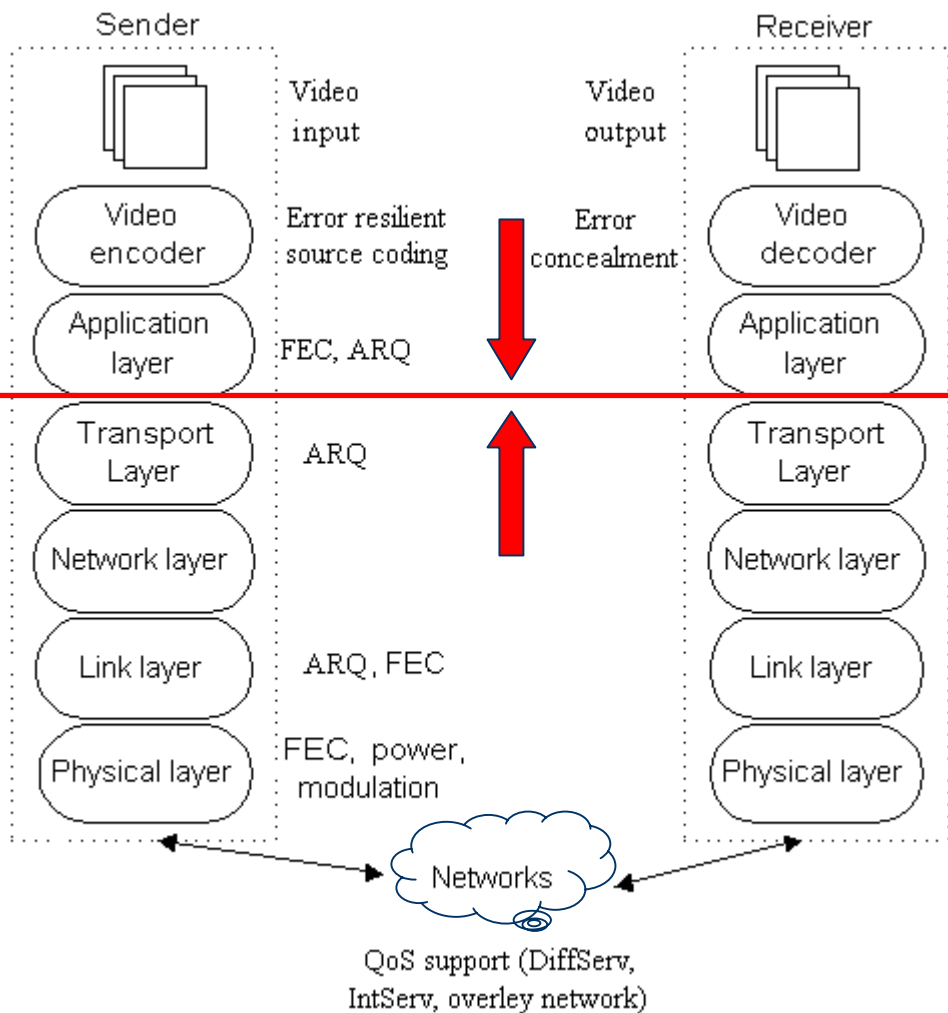
## Challenges

- Bandwidth (limited, time-varying)
  - Internet: congestion
  - Wireless: fading, shadowing
- QoS (packet loss, delay)
  - Distortion
  - Strict end-to end delay constraint
- Limited resources
  - Internet: bandwidth, buffer
  - Wireless: battery, bandwidth

## Approaches

- Network-adaptive
- Error control
  - Unequal error protection
- Cross-layer design

# Cross-Layer Design



## Error control Components

- Error resilient source coding
  - adapt source coding to channel
- FEC (forward error correction)
  - modify channel characteristics
- ARQ (delay constraint)
- Power control
  - modify channel characteristics
- Modulation, Rate adaptation
- QoS support
  - reserve resource to support QoS
- Error concealment
  - recover/conceal the error at decoder

# Related Work

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## ➤ Optimization on each error control component

- Error resilient source coding: R. Zhang, et al, '00
- FEC: B. Hong, et al '02
- ARQ: P. A. Chou, et al, '01
- Error concealment: Y. Wang et al, '00

## ➤ Cross-layer design

- Joint source-channel coding: M. Gallant, et al, *IEEE CSVT'01*
- Joint source coding and power control: Y. Eisenberg, et al, *CSVT'02*
- Joint source-channel coding and power control: Y.S. Chan, et al, *JSAC '02, sub*
- Joint source-channel coding and power control: S. Appadwedula, et al, *ICIP'98*

## ➤ Difference in our work

- The available error control components are considered in an integrated manner
- We consider error resilient source coding
- We consider error concealment

# Presentation Outline

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- Resource-distortion optimization framework
- Internet: **JSCC** -- Joint source-channel coding
- Wireless: **JSCCPA** -- Joint source-channel coding and power allocation
- DiffServ: **JSCPC** -- Joint source coding and packet classification
- Extensions to scalable video transmission
- Conclusions and future work



# Resource-Distortion Optimization

➤ General formulation

$$\min_{\{\boldsymbol{\mu}, \boldsymbol{\nu}\}} E [D (\boldsymbol{\mu}, \boldsymbol{\nu})]$$

End-to-End Distortion

$$\text{s.t. } C (\boldsymbol{\mu}, \boldsymbol{\nu}) \leq C_0$$

Cost Constraint

$$T (\boldsymbol{\mu}, \boldsymbol{\nu}) \leq T_0$$

Transmission Delay Constraint

- $\boldsymbol{\mu}$ : Source coding parameter vector
- $\boldsymbol{\nu}$ : Resource allocation parameter vector
- $C_0$  is explicitly determined by specific application
- $T_0$  is more implicitly determined by the application
  - obtained from higher-level rate controller, which is not incorporated into this work.

# End-to-End Distortion

- Consider end-to-end distortion



Original



Encoded



Decoded

- Distortion of the  $i$ -th pixel in the  $n$ -th frame

$$E[d_i^{(n)}] = E[(f_i^{(n)} - \tilde{f}_i^{(n)})^2] = (f_i^{(n)})^2 - 2f_i^{(n)} E[\tilde{f}_i^{(n)}] + E[(\tilde{f}_i^{(n)})^2]$$

- Distortion for the  $k$ -th packet that has  $M$  pixels

$$E[D_k] = \sum_{i=1}^M E[d_i^{(n)}] / M$$

# ROPE (Recursive Optimal Per-pixel Estimate)

➤ 1<sup>st</sup> order expected value (same fashion for 2<sup>nd</sup> order)

- Intra:  $E[\tilde{f}_i^{(n)}] = (1 - \rho_k) E[\tilde{f}_i^{(n)}] + \rho_k E[\tilde{f}_i^{(n-1)}]$

- Inter:  $E[\tilde{f}_i^{(n)}] = (1 - \rho_k)(\hat{e}_i^{(n)} + E[\tilde{f}_j^{(n-1)}]) + \rho_k E[\tilde{f}_i^{(n-1)}]$

➤ In order to calculate the distortion for one frame, only need the 1<sup>st</sup> order and 2<sup>nd</sup> order expected reconstructed pixel value in the previous frame

➤ This algorithm is per-pixel accurate and can fully capture error propagation.

- R. Zhang, S.L. Regunathan, and K. Rose, “Video coding with optimal inter/intra-mode switching for packet loss resilience,” *IEEE J. Select. Areas Commun.*, June 2000.

# Expected Distortion

- Expected distortion for the  $k$ -th packet

$$E[D_k] = (1 - \rho_k) E[D_{R,k}] + \rho_k E[D_{L,k}]$$

Probability of Reception

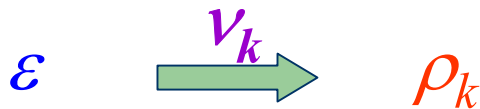
Expected distortion if the packet is received

Probability of Loss

Expected distortion if the packet is lost

- $E[D_{R,k}], E[D_{L,k}]$  : depends on source coding parameter  $\mu_k$
- $\rho_k(\varepsilon, \nu)$ : depends on CSI and resource allocation parameter

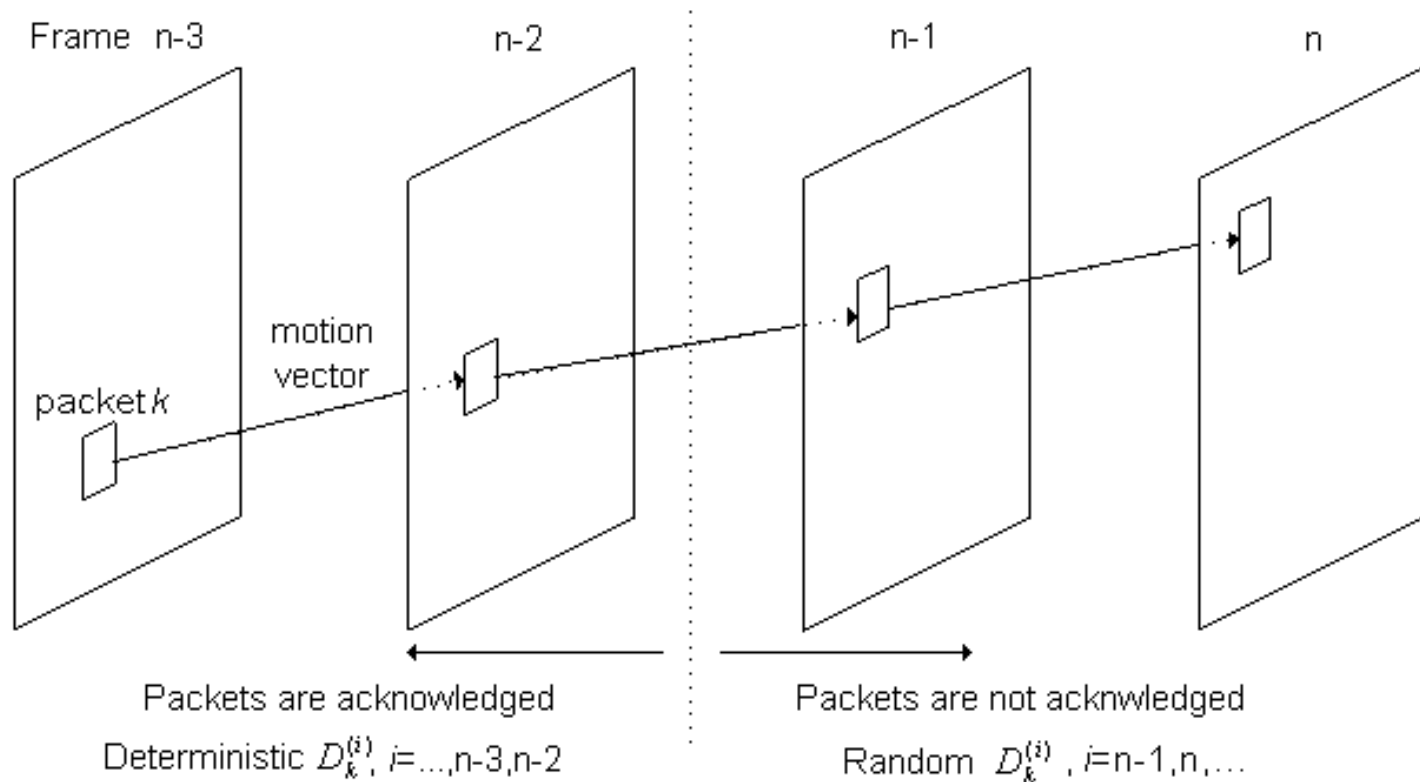
FEC parameter



Prob of loss for transport packet

Prob of loss for source/video packet

# Distortion Estimation Based on Feedbacks



- Expectations are taken with respect to the updated probability distribution of channel losses given the available feedback

# Transmission Delay and Cost

## ➤ Transmission Delay:

Number of packets in a frame

Packet index

$$T = \sum_{k=1}^M \frac{B_k(\mu_k)}{R_T}$$

Number of bits

Transmission rate

Source coding parameter

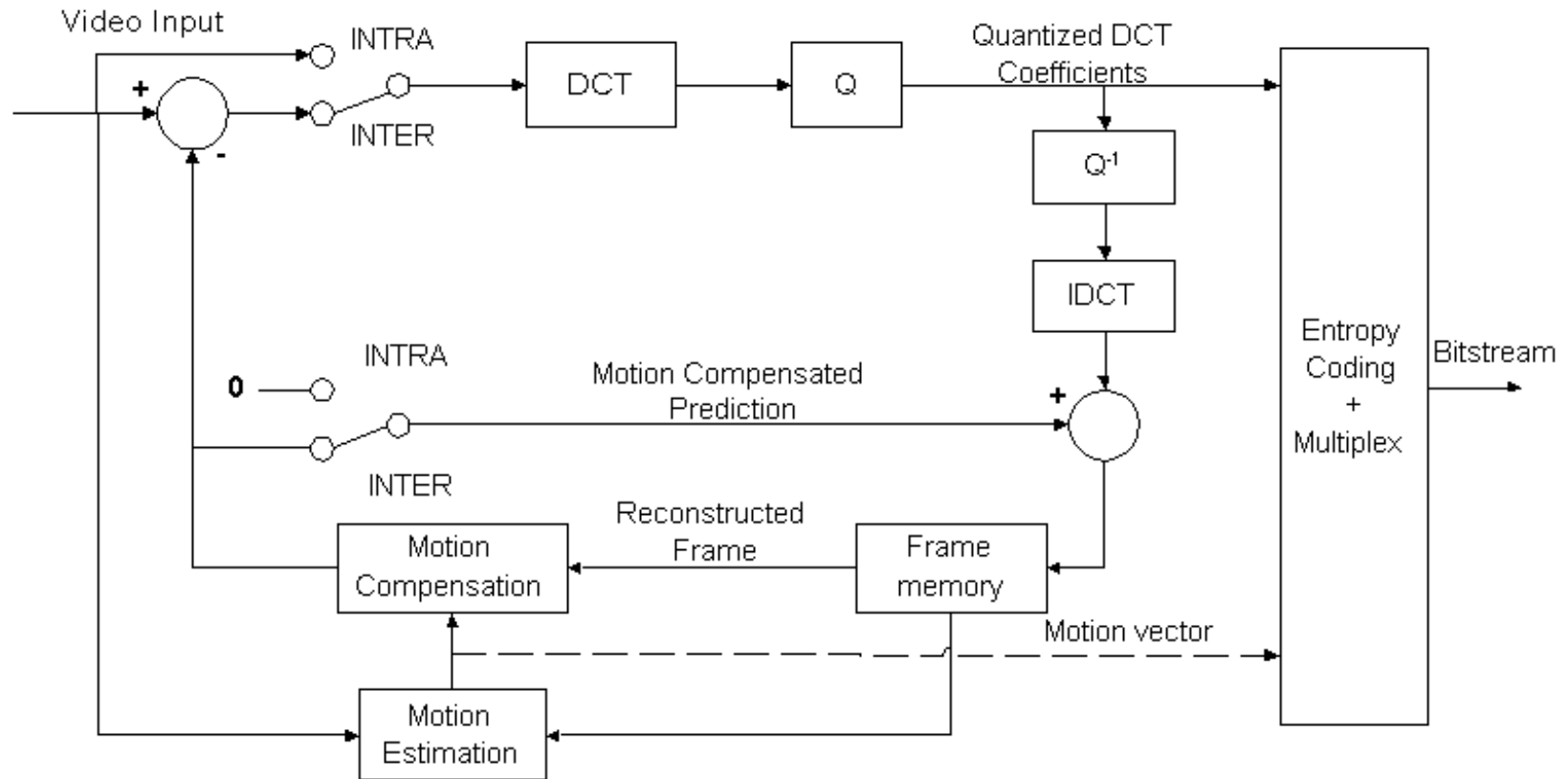
## ➤ Energy:

$$C = \sum_{k=1}^M \frac{B_k(\mu_k) P_k(\nu_k)}{R_T}$$

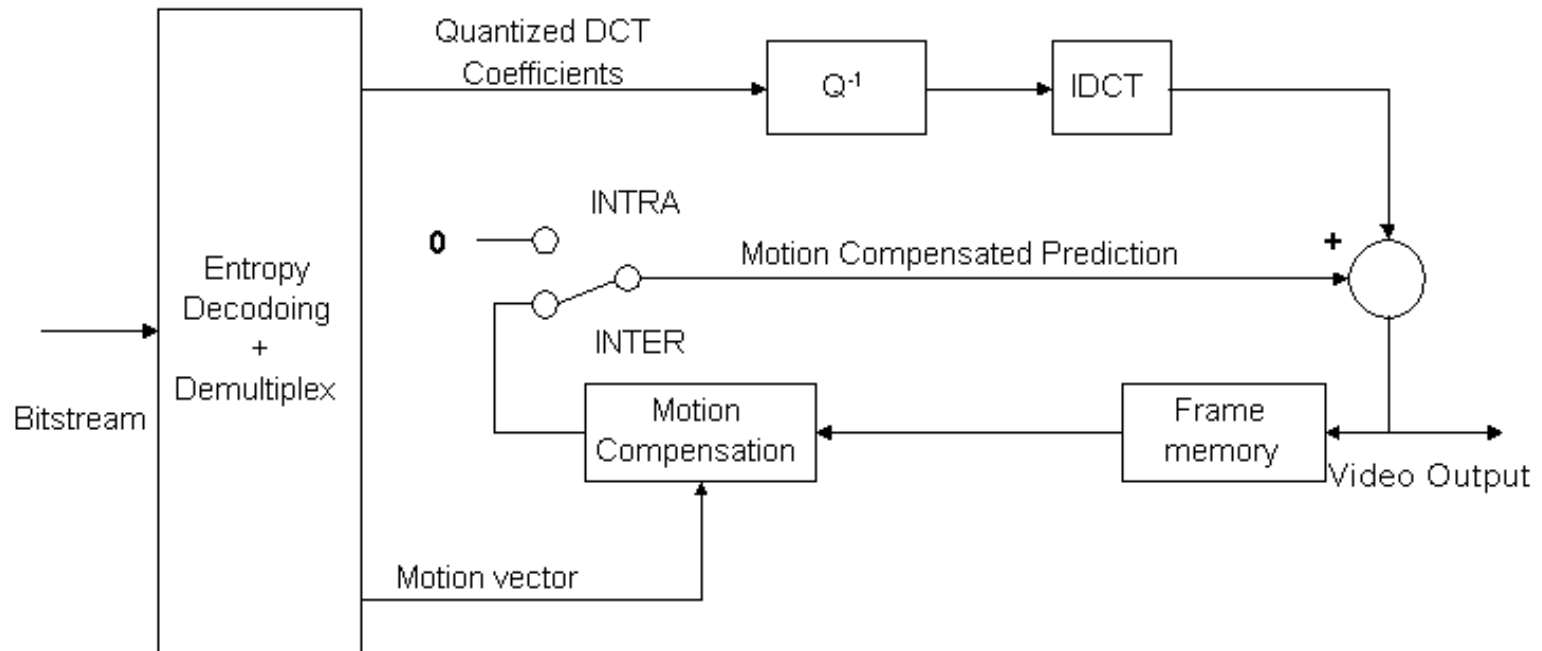
Power allocation parameter

Transmission power

# Motion-Compensated Video Encoder

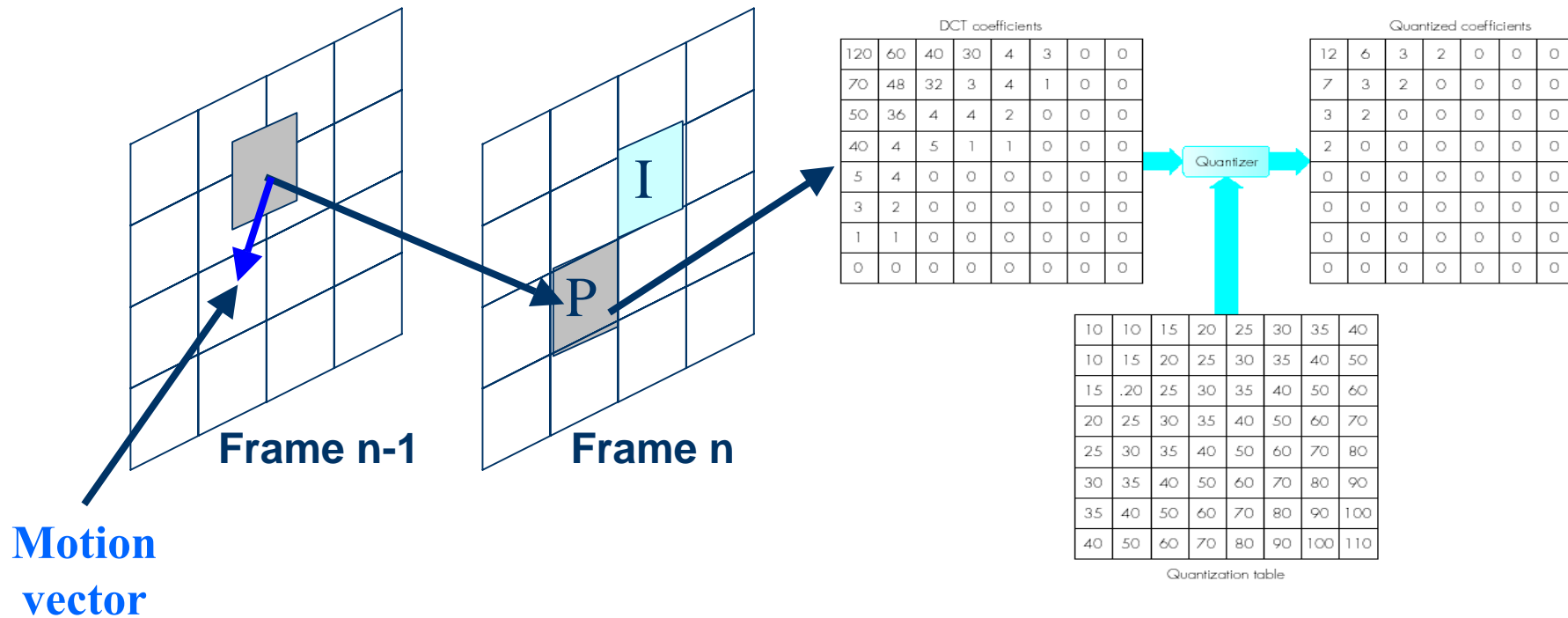


# Motion-compensated Video Decoder





# Motion-compensated Video Codec



- Source coding parameter  $\mu_k$ 
  - Prediction mode (Intra, Inter, Skip)
  - Quantization stepsize
- Different encoding modes result in different levels of coding efficiency and error robustness

# Channel Model

- Network model
  - Independent time-invariant packet erasure channel
- Packet loss
  - Internet:
    - Before FEC :  $\varepsilon$  -- Bernoulli process
    - After FEC :  $\rho_k = f(\varepsilon, v_k)$

- Wireless:

$$\rho_k = 1 - (1 - p_{e,k})^{B_k}$$

- i.i.d fading
- $B_k$  : Packet size
- $p_{e,k}$  : BER after channel coding

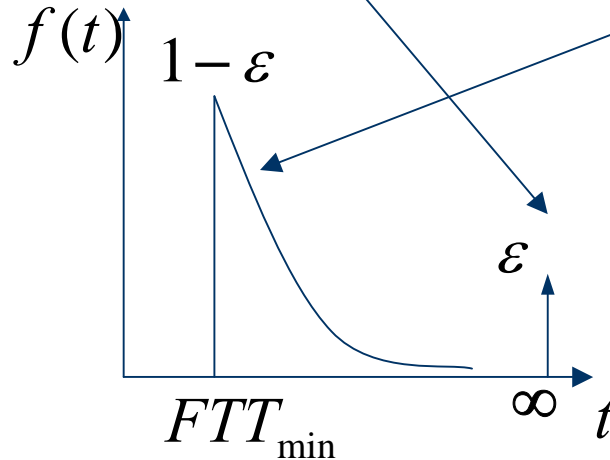
(depending FEC, channel SNR, and fading model)

# Channel Model—Internet

## ➤ Network model :

- Independent time-invariant packet erasure + random delays
- Receiver responds to a received/lost packet with a positive/negative acknowledgement.
- Perfect feedback channel: constant delay and no error

$$\rho^k = \varepsilon + (1 - \varepsilon) P\{\Delta T_n(k) > \tau\}$$



## ➤ Packet delay

- Exponential distribution: fast decaying tail
- **Gamma distribution**
- Pareto distribution: slowly decaying (heavy) tail

## ➤ Packet Loss

- **Bernoulli**
- 2-state Markov (Gilbert)
- High-order Markov

# Channel Model—Wireless

- Packet-erasure channel, from the point of view of the applications
- Channel BER (Bit Error Rate)

$$\text{BER} = \frac{1}{2} \left( 1 - \sqrt{\frac{\alpha E_b}{N_0 + \alpha E_b}} \right) \quad \text{Rayleigh fading}$$

$$\text{BER} = Q\left(\sqrt{\frac{2E_b}{N_0}}\right) \quad \text{AWGN}$$

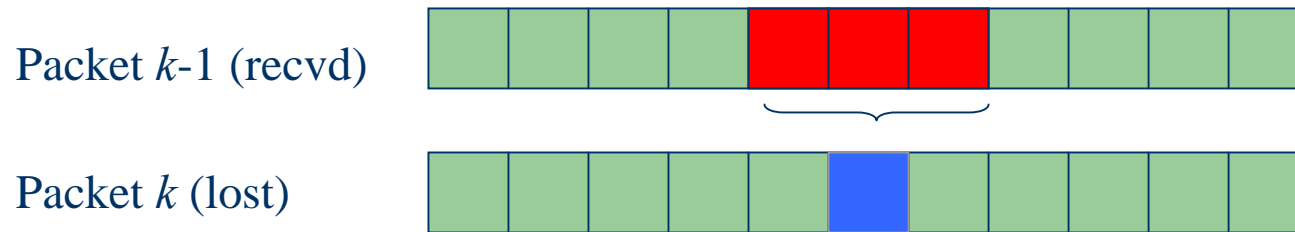
- Probability of Packet Loss

$$\rho_k = 1 - (1 - p_{e,k})^{B_k} \quad \begin{array}{l} p_{e,k} : \text{BER after channel coding} \\ B_k : \text{Packet size} \end{array}$$

- Wireless channel: probability of packet loss depends on source coding parameter, channel coding parameter, and power level

# Implementation Issues

- Packetization: each row of blocks (GOB) is packetized into a packet
- Error concealment: temporal concealment, median motion vectors

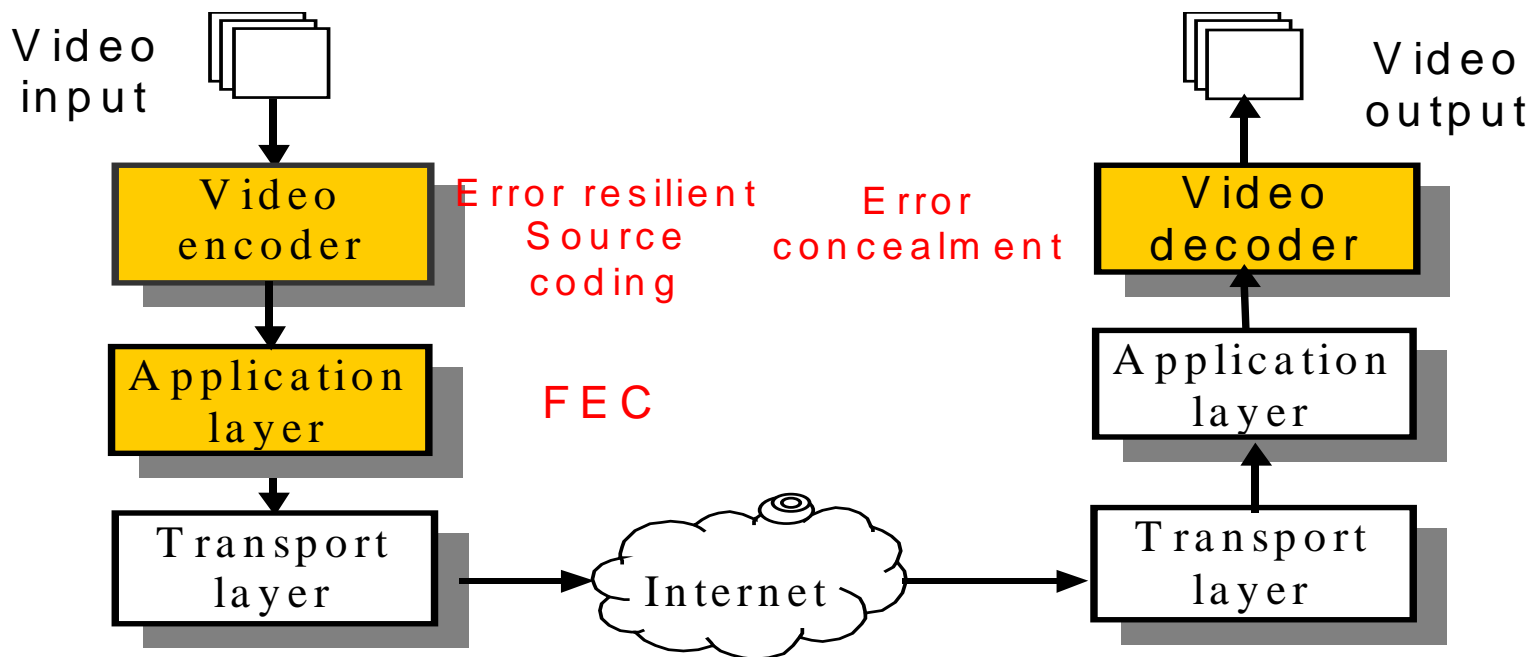


- Expected distortion

$$E[D_k] = (1 - \rho_k) E[D_{R,k}] + \rho_k E[D_{L,k}]$$

$$E[D_k] = (1 - \rho_k) E[D_{R,k}] + \rho_k (1 - \rho_{k-1}) E[D_{C,k}] + \rho_k \rho_{k-1} E[D_{Z,k}]$$

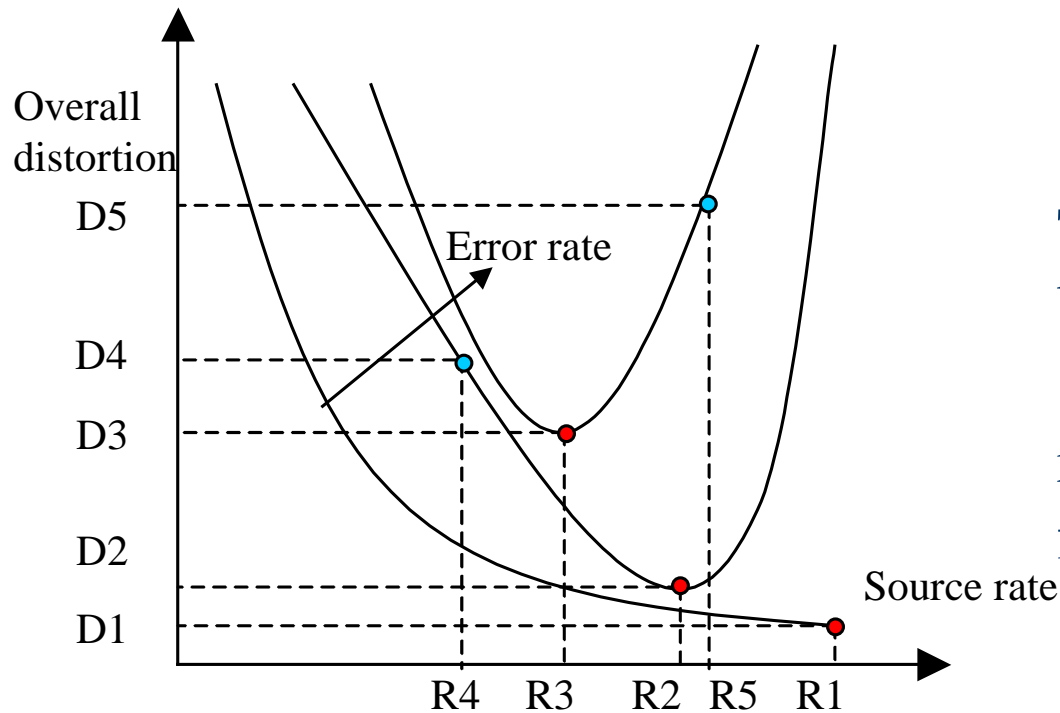
# Internet Video Transmission



- We jointly consider error resilient source coding, FEC, and error concealment

# Joint Source-Channel Coding

- Shannon's separation theorem does not strictly hold due to the delay and complexity constraints



The total bandwidth for source and channel rates is the same for the three curves

- Overall distortion = source distortion + channel distortion
- Optimal bit allocation depends on the channel characteristics

# Sequential Joint Source-Channel Coding

- Step 1: Bit allocation between source and channel

$$\begin{aligned} & \min_{\{\mathbf{v} \in R\}} E[D(\mathbf{v})] \\ & s.t. \quad T(\mathbf{v}) = B_s(\boldsymbol{\mu}(\mathbf{v})) / R_T + B_c(\mathbf{v}) / R_T \leq T_0 \end{aligned}$$

- Step 2: Source coding based on given bit budget

$$\begin{aligned} & \min_{\{\boldsymbol{\mu} \in Q\}} E[D(\boldsymbol{\mu})] \\ & s.t. \quad T_s(\boldsymbol{\mu}) = B_s(\boldsymbol{\mu}) / R_T \leq T_{s,0} \end{aligned}$$

$\boldsymbol{\mu}$ : source coding parameter	$B_s$ : source bits
$\mathbf{v}$ : channel coding parameter	$B_c$ : channel bits
$T_0$ : transmission delay constraint	$T_{s,0}$ : transmission delay constraint for the source



# Integrated Joint Source-Channel Coding

- Solve (1) Bit allocation between source and channel and (2) source coding and channel coding, in one step

$$\begin{aligned} & \min_{\{\boldsymbol{\mu} \in Q, \boldsymbol{v} \in R\}} E[D(\boldsymbol{\mu}, \boldsymbol{v})] \\ & s.t. \quad T(\boldsymbol{\mu}, \boldsymbol{v}) = B(\boldsymbol{\mu}, \boldsymbol{v}) / R_T \leq T_0 \end{aligned}$$

$B$ : total bits used for both source and channel coding

$R_T$  : transmission rate

$T_0$  : transmission delay constraint for this frame

# Experimental Results

- System 1: Integrated JSCC

$$\min_{\{\boldsymbol{\mu} \in Q, \mathbf{v} \in R\}} E[D(\boldsymbol{\mu}, \mathbf{v})]$$

- System 2: Fixed channel coding rate

$$\min_{\{\boldsymbol{\mu} \in Q\}} E[D(\boldsymbol{\mu}, \mathbf{v}_\theta)]$$

- System 3: Sequential JSCC

$$\min_{\{\mathbf{v} \in R\}} E[D(\mathbf{v})]$$

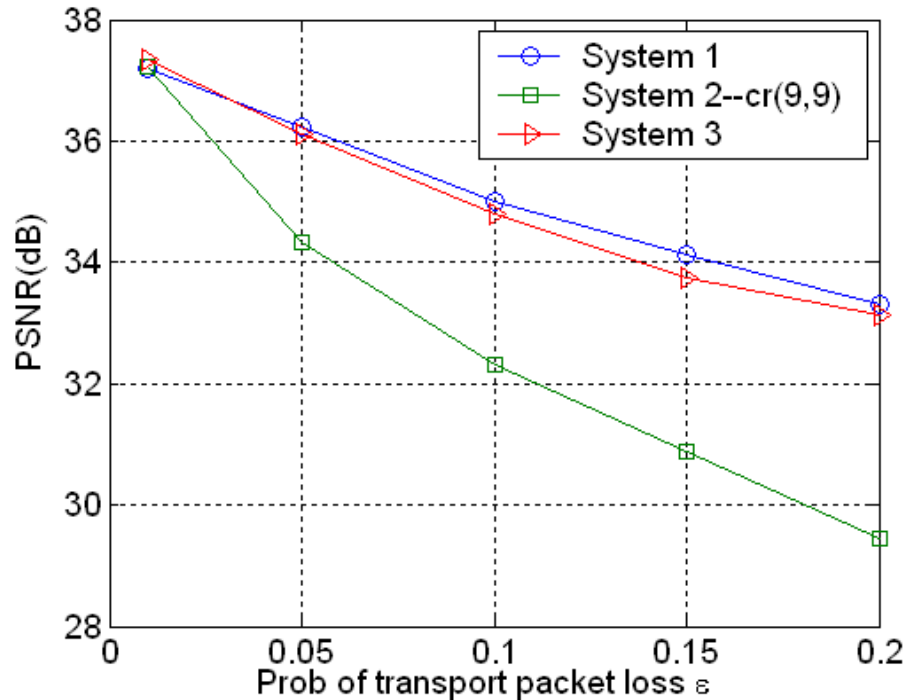
$$s.t. \quad T(\mathbf{v}) = B_s(\boldsymbol{\mu}(\mathbf{v})) / R_T + B_c(\mathbf{v}) / R_T \leq T_0$$

$$\min_{\{\boldsymbol{\mu} \in Q\}} E[D(\boldsymbol{\mu})]$$

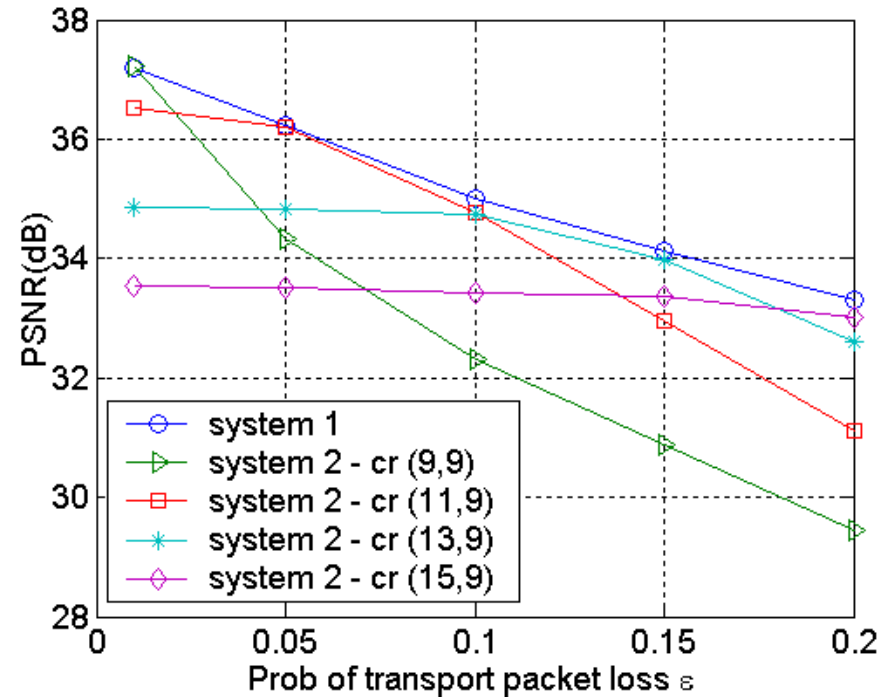
$$s.t. \quad T_s(\boldsymbol{\mu}) = B_s(\boldsymbol{\mu}) / R_T \leq T_{s,0}$$

# Experimental Results

## Average PSNR vs. transport packet loss probability



System 1 vs. System 2 and 3

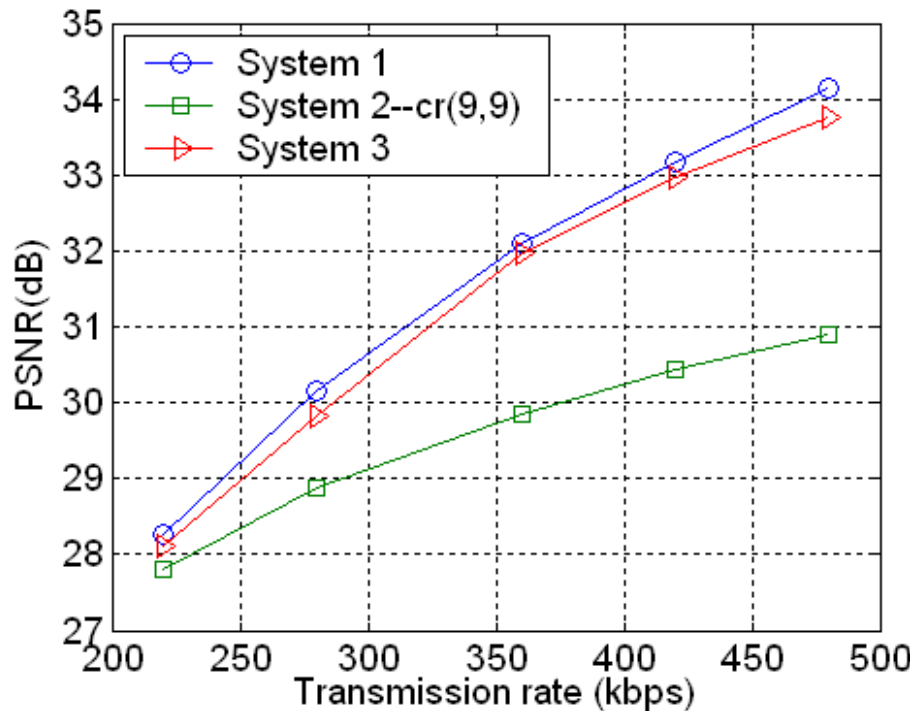


System 1 vs. System 2 with indicated channel rates

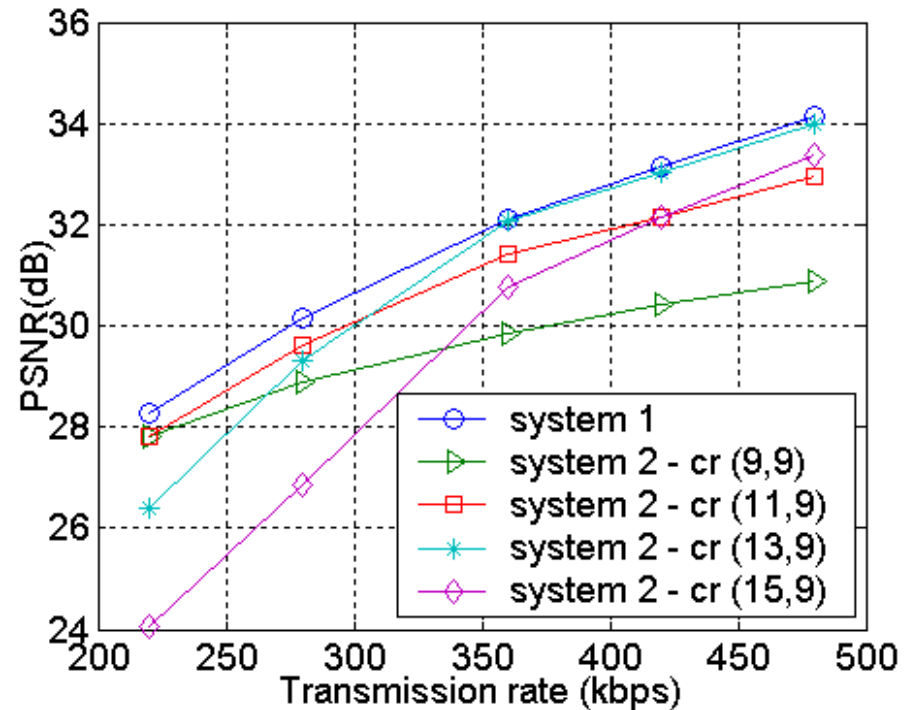
( $R_T=480$  kbps,  $F=15$  fps, cr in the legend denotes channel rates).

# Experimental Results

## Average PSNR vs. transmission rate



System 1 vs. System 2 and 3

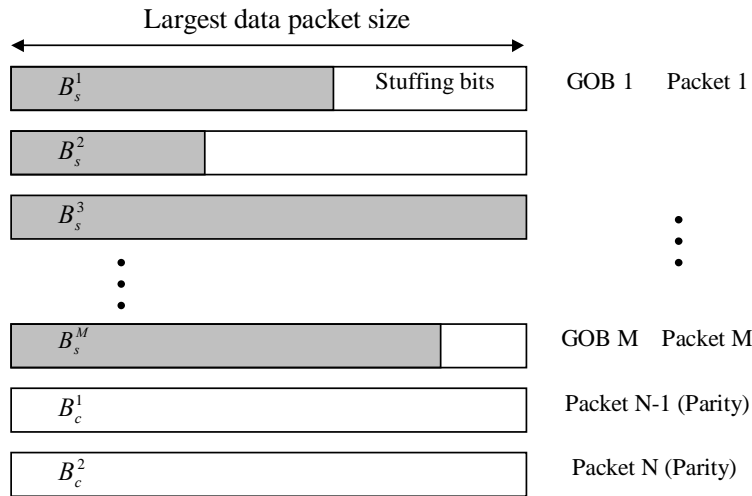


System 1 vs. System 2 with indicated channel rates

( $\varepsilon=0.15$ ,  $F=15$  fps, cr in the legend denotes channel rates).

# JSCC—Packetization

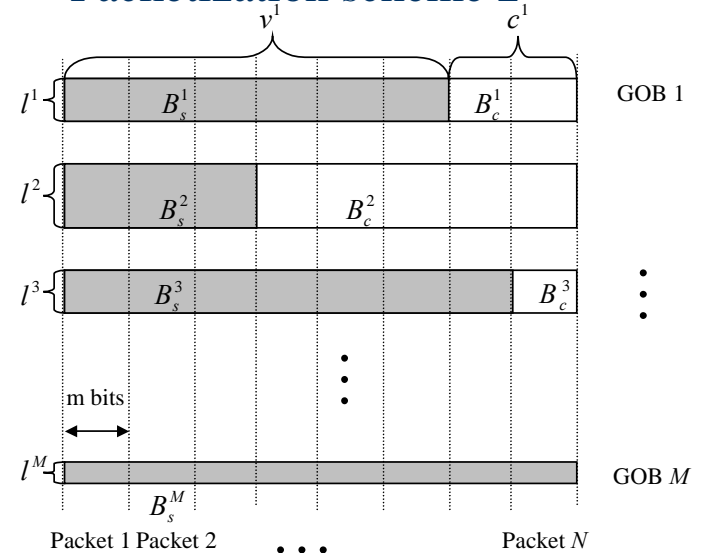
Packetization scheme 1



$RS(N, M)$

$$\begin{aligned} \rho &= \varepsilon P_b(N-1, k) \\ &= \varepsilon \left( 1 - \sum_{i=0}^{N-1-k} \binom{N-1}{i} \varepsilon^i (1-\varepsilon)^{N-1-i} \right) \end{aligned}$$

Packetization scheme 2



$RS(N, v^k)$

$$\begin{aligned} \rho^k &= P_b(N, v^k) \\ &= 1 - \sum_{i=0}^{N-v^k} \binom{N}{i} \varepsilon^i (1-\varepsilon)^{N-i} \end{aligned}$$

# JSCC — Solution Algorithm

- Lagrangian relaxation and DP

$$\min_{\{\mu, \nu\}} \sum_{k=1}^M J_k = \sum_{k=1}^M E[D_k(\mu, \nu)] + \lambda B_k(\mu_k, \nu) / R_T$$

- Packetization 1

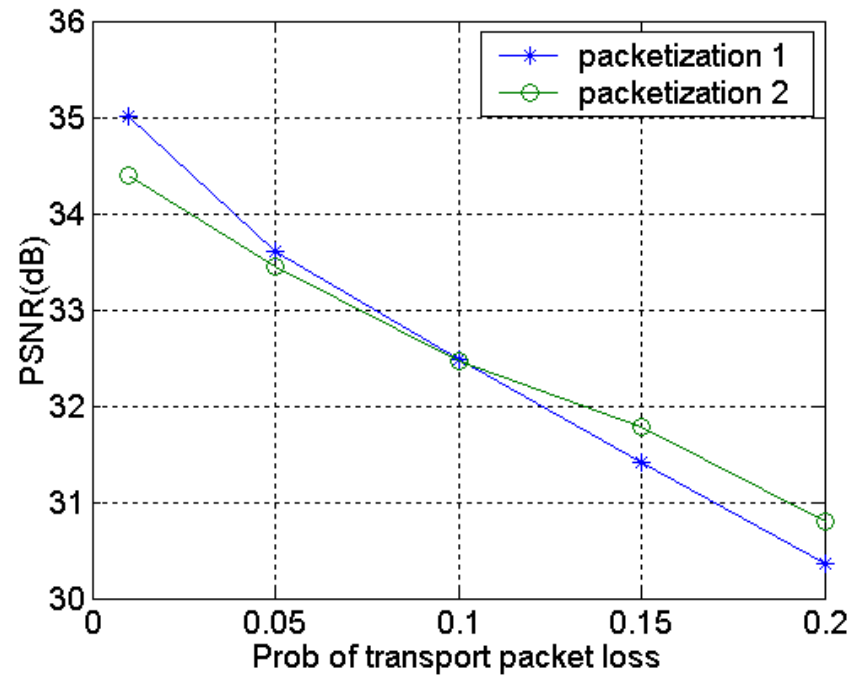
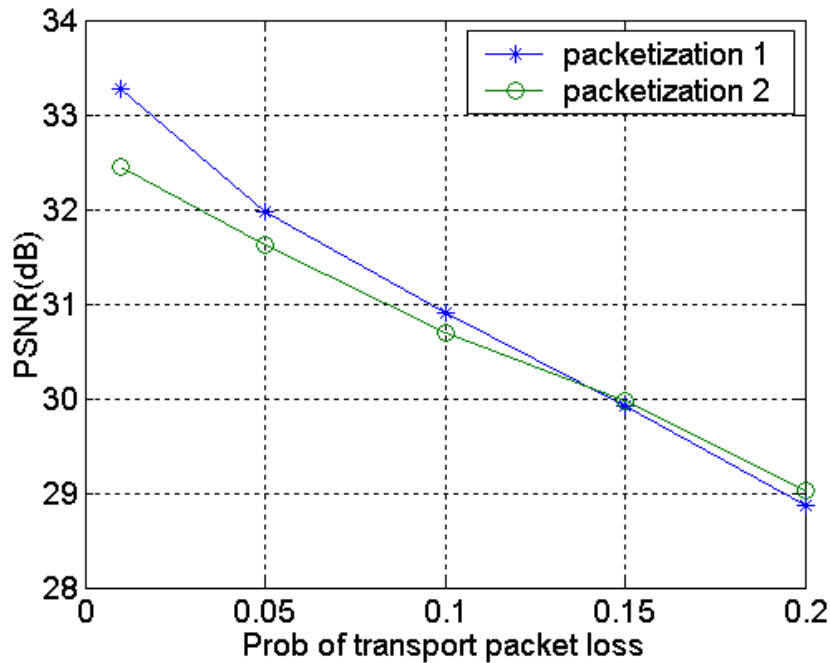
$$\min_{\{\nu\}} \sum_{k=1}^M J_k(\mu^* | \nu) = \min_{\{\nu\}} \left\{ \min_{\{\mu\}} \sum_{k=1}^M J_k(\mu, \nu) \right\}$$

- Packetization 2

$$\min_{\{\mu, \nu\}} \sum_{k=1}^M J_k = \min_{\{\mu, \nu\}} \sum_{k=1}^M J_k(\mu_{k-1}, \mu_k, \nu_{k-1}, \nu_k)$$

# Experimental Results

## Packetization scheme 1 vs 2



RTP/UDP/IP header = 40bytes

# FEC vs. ARQ

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## ➤ FEC:

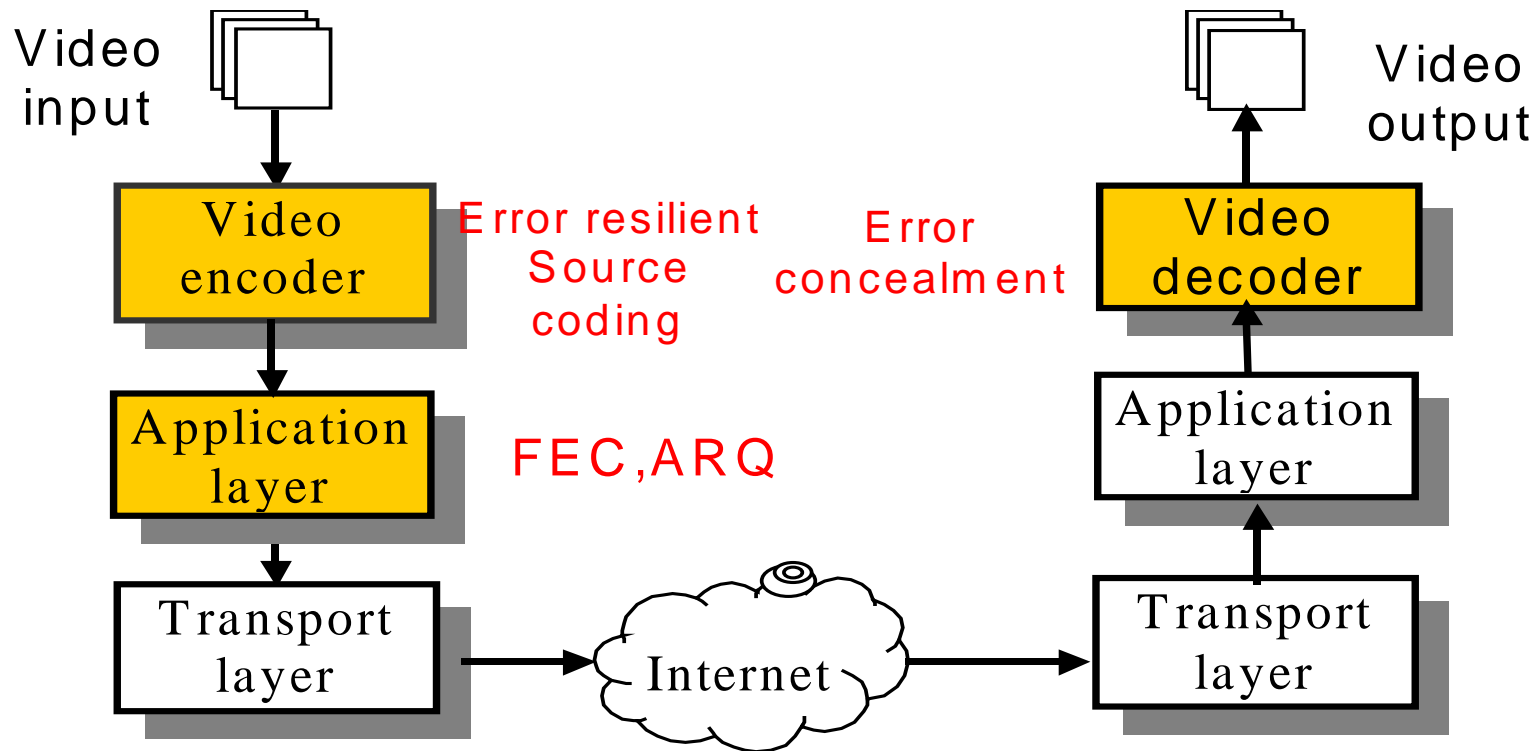
- Usually preferred for real-time video applications
- Cannot completely avoid packet loss
- Incur constant overhead even when the channel has no loss
- Depend on the accurate CSI estimation

## ➤ ARQ:

- Can automatically adapt to the channel loss characteristics
- Longer delay
- Useful for long end-to-end delay applications, e.g., on-demand video streaming
- Useful for short RTT situations, e.g., LAN

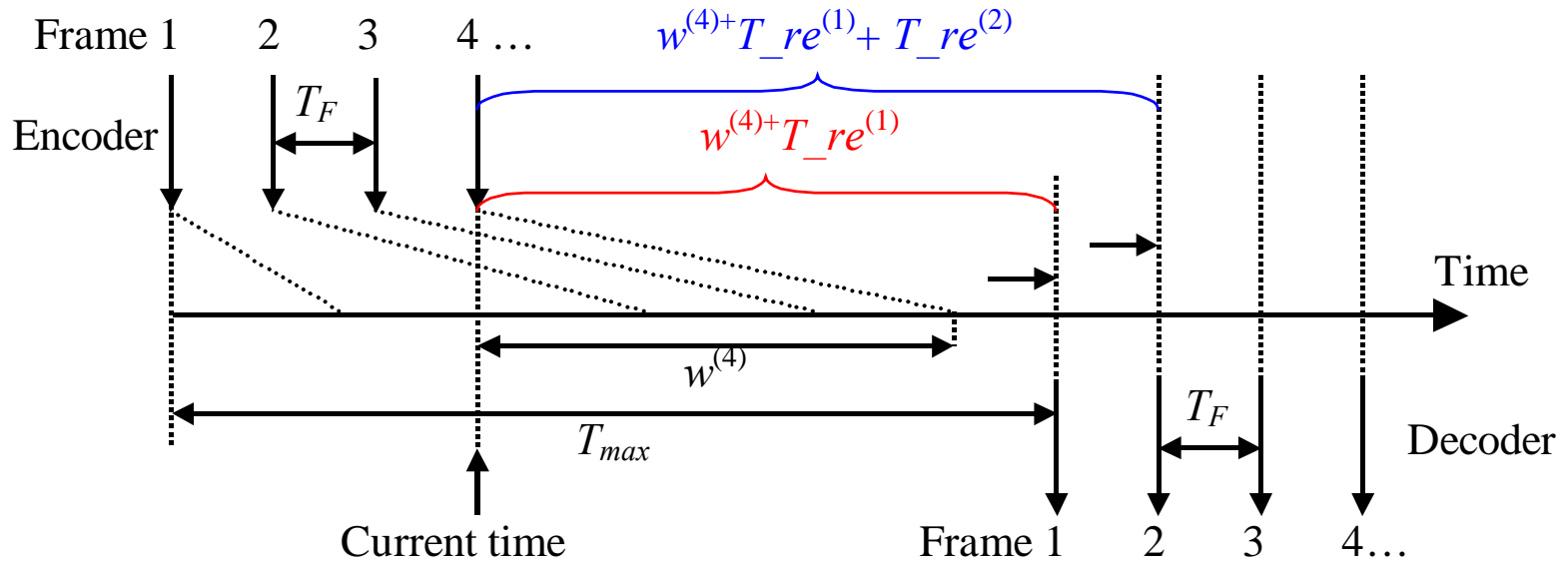


# JSCC—Hybrid FEC/Retransmission



- We jointly consider error resilient source coding, FEC, application-layer ARQ, and error concealment

# Delay-Constraint Retransmission



- $T_{max}$ : Maximum end-to-end delay (imposed by application)
- $w^{(n)}$ : waiting time for the  $n$ -th frame in the encoder buffer
- $T_{re}^{(n)}$ : transmission delay for retransmitting packets in the  $n$ -th frame
- $T_F$ : One frame's time ( $=1/T_F$ )

# Problem Formulation

$$\left\{ \begin{array}{l} \min_{\{\mu, \gamma, \sigma\}} \sum_{i=0}^A E[D^{(n-i)}] = \sum_{i=1}^A E[D^{(n-i)}(\sigma^{(n-i)})] + \sum_{k=1}^M E[D_k^{(n)}(\mu, \gamma)] \\ \text{s.t. } w^{(n)} + \sum_{i=1}^j \sum_{k=1}^M \sigma_k^{(n-i)} T_k^{(n-i)} \leq T_{\max} - (A+1-j)T_F \quad j = 1 \dots A \\ w^{(n)} + \sum_{i=1}^A \sum_{k=1}^M \sigma_k^{(n-i)} T_k^{(n-i)} + \sum_{k=1}^M T_k^{(n)} \leq T_{\max} \end{array} \right.$$

- $A$ : # of frames that retransmission is eligible
- $\sigma^{(n)} = \{\sigma_1^{(n)}, \dots, \sigma_M^{(n)}\} \in \{0, 1\}$ : Retrans parameter vector for frame  $n$
- $\gamma$ : FEC parameter (using RS code)
- $T_k^{(n)}$ : Transmission delay for packet  $k$  in frame  $n$ .
- $T_{\max}$  can be replaced by  $T_{\max} - \Delta T^{(n)}$ , where  $\Delta T^{(n)}$  is decided by a rate controller.

# Problem Formulation

$$\left\{ \begin{array}{l} \min_{\{\mu, \gamma, \sigma\}} \sum_{i=0}^A E[D^{(n-i)}] = \sum_{i=1}^A E[D^{(n-i)}(\sigma^{(n-i)})] + \sum_{k=1}^M E[D_k^{(n)}(\mu, \gamma)] \\ \text{s.t.} \quad \sum_{i=1}^A \sum_{k=1}^M \sigma_k^{(n-i)} T_k^{(n-i)} + \sum_{k=1}^M T_k^{(n)} \leq T_0^{(n)} \end{array} \right.$$

- We assume  $T_0^{(n)}$  is given from higher-level rate controller
- Optimization with a sliding window of  $A+1$  frames
- Optimization decisions: retransmission policy for the first  $A$  frames, and source coding and FEC for the current frame
- Based on the updated probabilities of packet loss, recalculate the expected distortion of all packets in the window
- Optimization window shifts at the frame level

# Probability of Packet Loss for the Current Frame

- Probability of packet loss

$$\rho_k^{(n)} = \rho_{k,FEC}^{(n)} \rho_{k,RET}^{(n)}$$

Due to FEC Due to retransmission

- The probability of loss in future retransmissions can only be estimated since the acknowledgement information and retransmission decisions are not available in the encoding of the current frame
- Approximation formula

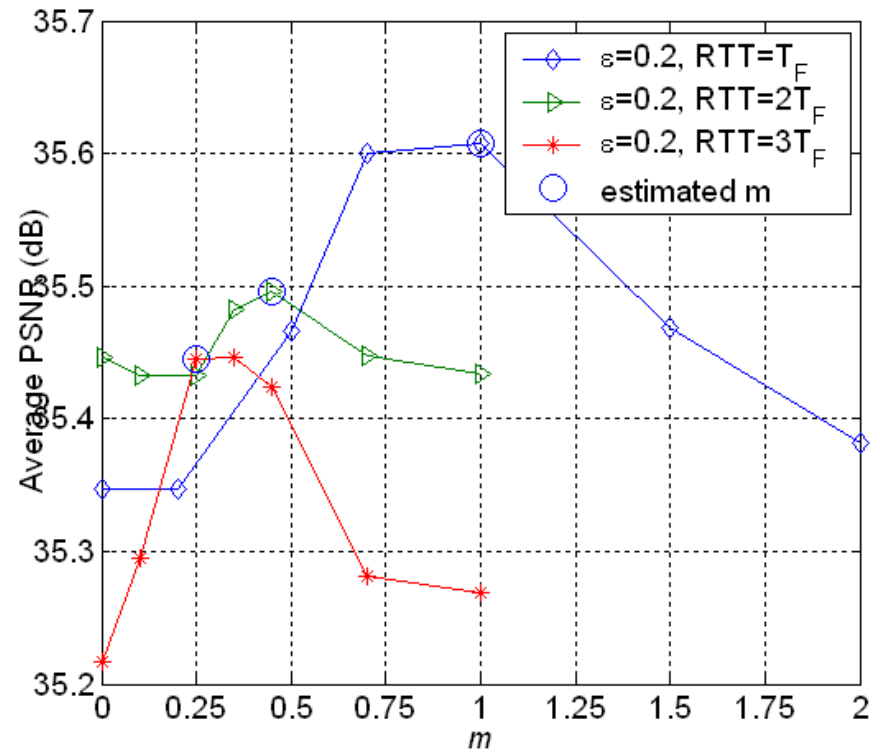
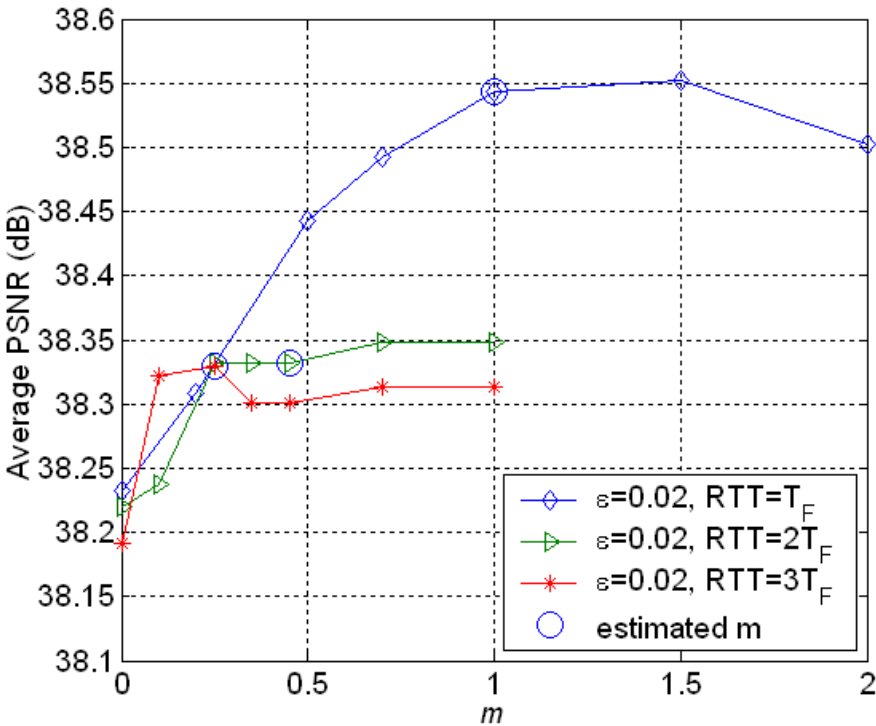
$$\rho_{k,RET}^{(n)} = \varepsilon^m$$

- The estimate of the total number of retransmissions

$$\tilde{m} = \frac{A}{(1 + RTT)^2}$$

# Estimating $\tilde{m}$

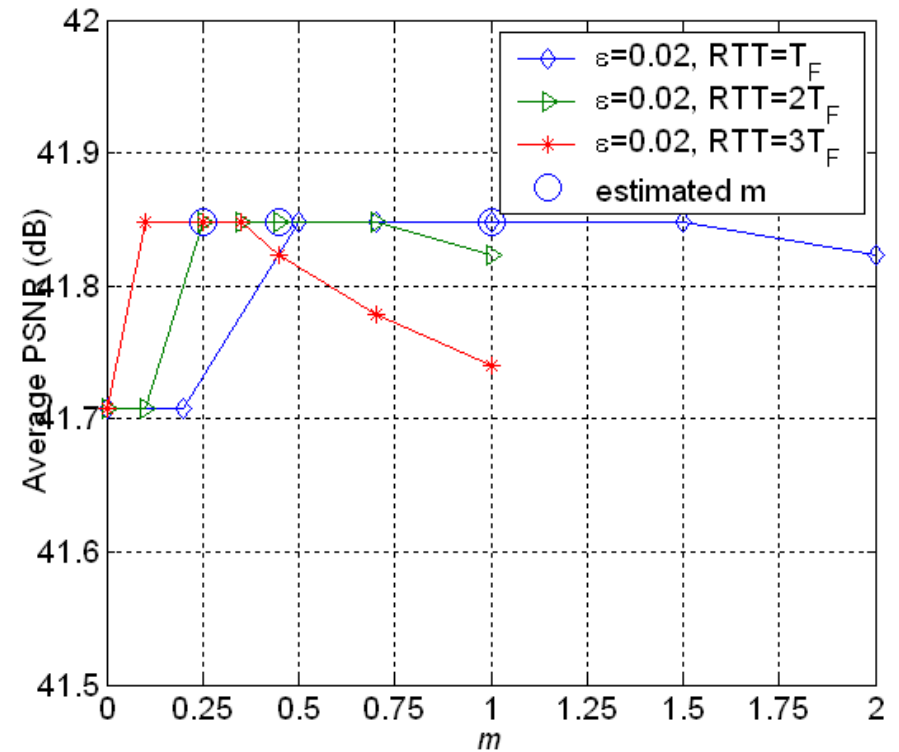
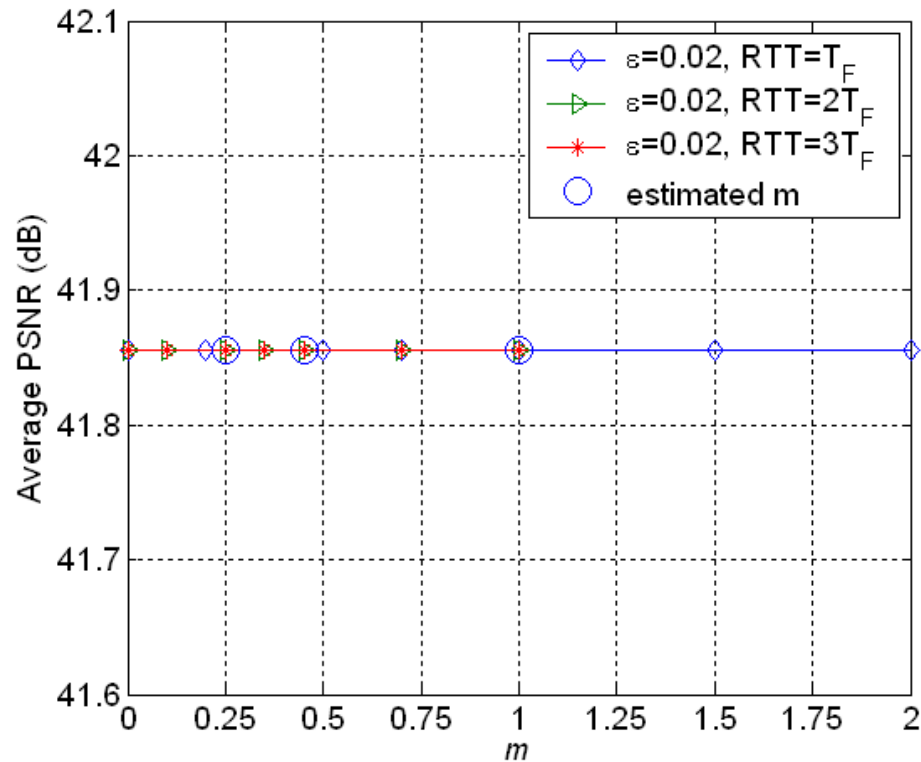
Average PSNR vs.  $\tilde{m}$



QCIF Foreman sequence at  $F=15$  fps,  $R_T=480$  kbps and  $A=4$

# Estimating $\tilde{m}$

## Average PSNR vs $\tilde{m}$



QCIF Akiyo sequence at  $F=15$  fps,  $R_T=360$  kbps and  $A=4$

# Probability of Packet Loss for the Past Frame

- Probability of packet loss

$$\rho_k^{(n)} = \rho_{k,UPD}^{(n)} \rho_{k,RET}^{(n)}$$

Updated loss probability  
based on feedback

Due to  
retransmission

- $\rho_{k,UPD}^{(n)}$  can be accurately calculated
- $\rho_{k,RET}^{(n)}$  can be calculated using the similar way as before



# Probability of Packet Loss for the Past Frame

➤ Assume protected by RS( $N, M$ ),

- $L$ : # packets lost
- $V$ : # retransmitted packets
- $J=L+M-N$

➤ If  $V < J$   $\rho_{k,RET}^{(n-i)} = \varepsilon^{\sigma_k^{(n-i)}}$

➤ If  $V = J$   $\rho_{k,RET}^{(n-i)} = \begin{cases} \varepsilon & \text{if } \sigma_k^{(n-i)} = 0 \\ 1 - (1 - \varepsilon)^J & \text{if } \sigma_k^{(n-i)} = 1 \end{cases}$

➤ If  $V > J$   $\rho_{k,RET}^{(n-i)} = \begin{cases} \sum_{j=V-J+1}^V \frac{j}{V} \binom{V}{j} \varepsilon^j (1 - \varepsilon)^{V-j} & \text{if } \sigma_k^{(n-i)} = 0 \\ \sum_{j=V-J+1}^V \binom{V}{j} \varepsilon^j (1 - \varepsilon)^{V-j} & \text{if } \sigma_k^{(n-i)} = 1 \end{cases}$

# Solution Algorithm

## ➤ Lagrangian Relaxation

$$\min_{\{\mu, \gamma, \sigma\}} \sum_{i=0}^A J^{(n-i)} = \sum_{i=1}^A E[D^{(n-i)}(\sigma^{(n-i)})] + \sum_{k=1}^M E[D_k^{(n)}(\mu, \gamma)]$$

$$+ \lambda \left\{ \sum_{i=1}^A \sum_{k=1}^M \sigma_k^{(n-i)} T_k^{(n-i)} + \sum_{k=1}^M T_k^{(n)} \right\}$$

## ➤ Minimization of the Lagrangian

$$\min_{\{\mu, \gamma, \sigma\}} \sum_{i=0}^A J^{(n-i)} = \min_{\{\sigma\}} \sum_{i=1}^A J^{(n-i)}(\sigma^{(n-i)}) + \min_{\{\gamma\}} \left\{ \min_{\{\mu\}} \sum_{k=1}^M J_k^{(n)}(\mu, \gamma) \right\}$$

Retransmission  
Exhaustive  
search

FEC  
Exhaustive  
search

Source coding  
DP

# Simulation – Hybrid FEC/Retransmission

---

- NFNR– Neither FEC Nor Retransmission

$$\min_{\{\mu\}} E[D(\mu, \gamma_0, \sigma_0)]$$

- Pure Retransmission

$$\min_{\{\mu, \sigma\}} E[D(\mu, \gamma_0, \sigma)]$$

- Pure FEC

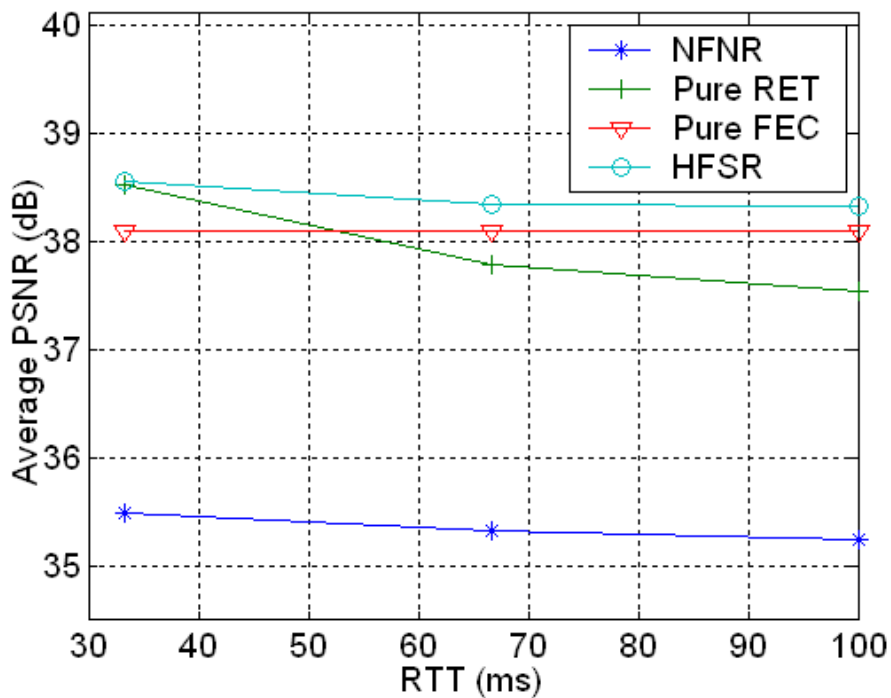
$$\min_{\{\mu, \gamma\}} E[D(\mu, \gamma, \sigma_0)]$$

- HFSR (Hybrid FEC and Selective Retransmission)

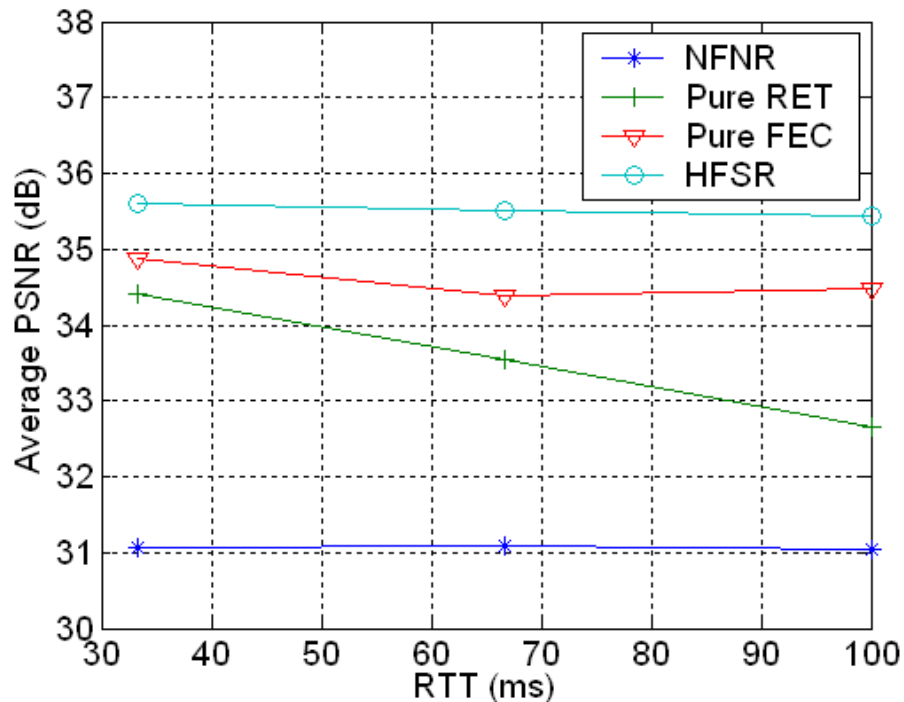
$$\min_{\{\mu, \gamma, \sigma\}} E[D(\mu, \gamma, \sigma)]$$

# Sensitivity to RTT

## Average PSNR vs. RTT



$\varepsilon=0.02$

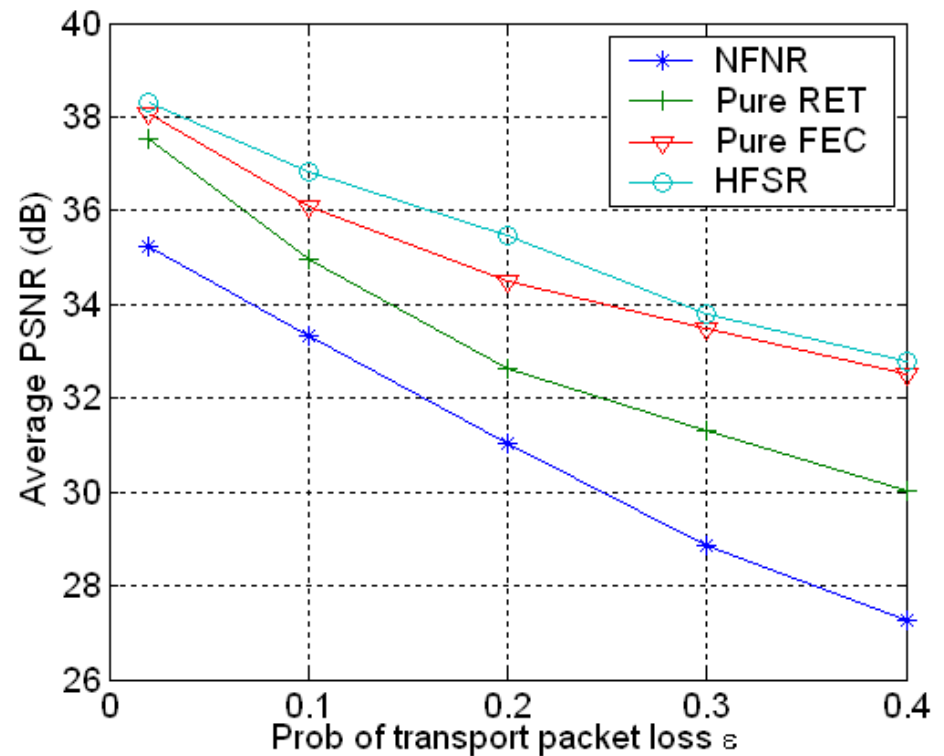
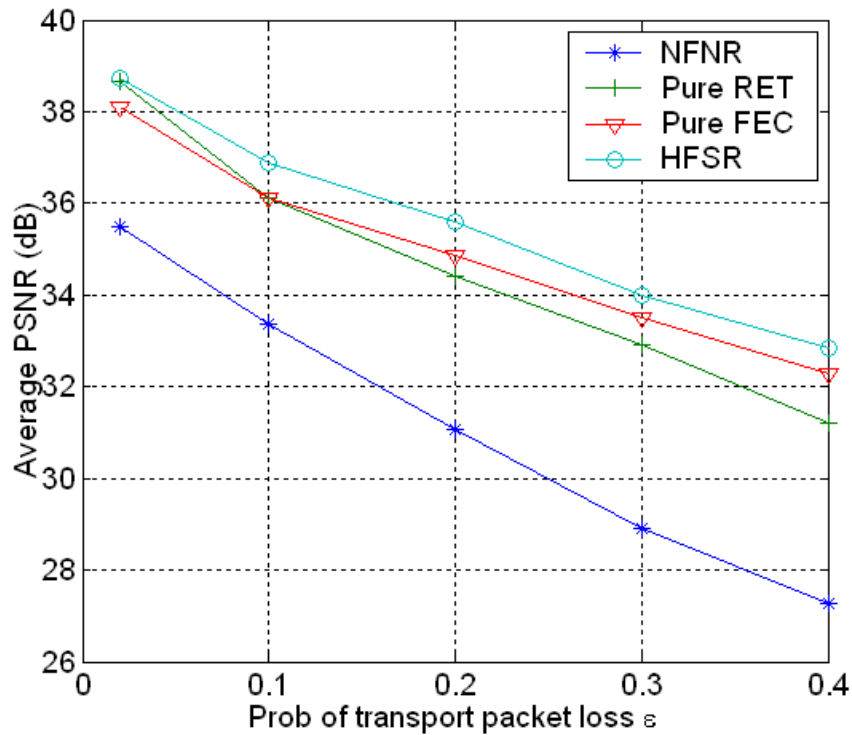


$\varepsilon=0.2$

$R_T=480$  kbps,  $F=15$  fps, QCIF Foreman

# Sensitivity to Packet Loss Rate

Average PSNR vs. probability of transport packet loss  $\varepsilon$



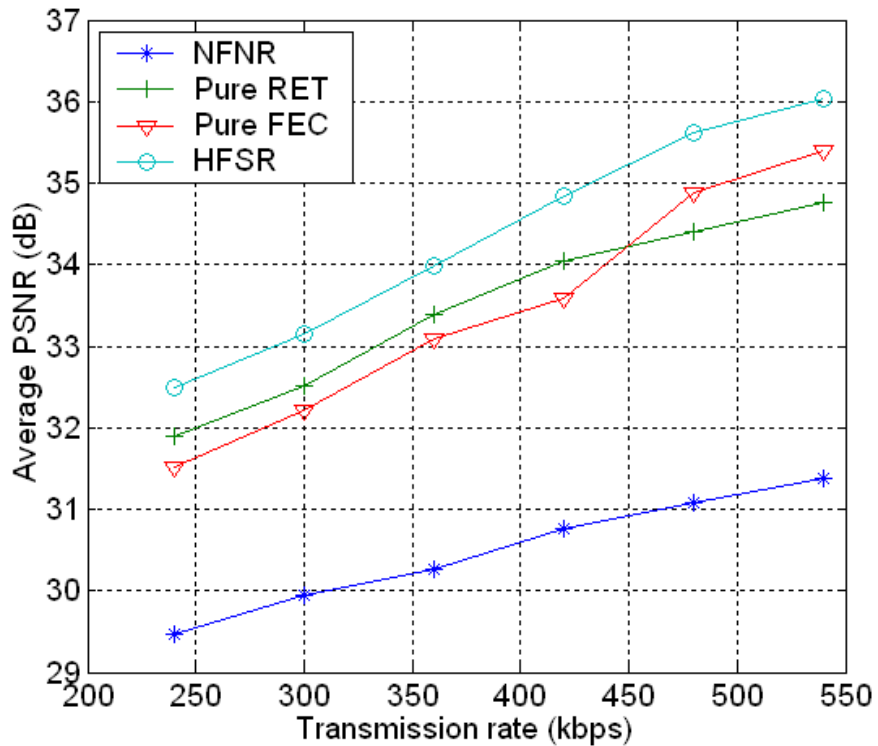
$$RTT=T_F$$

$$RTT=3T_F$$

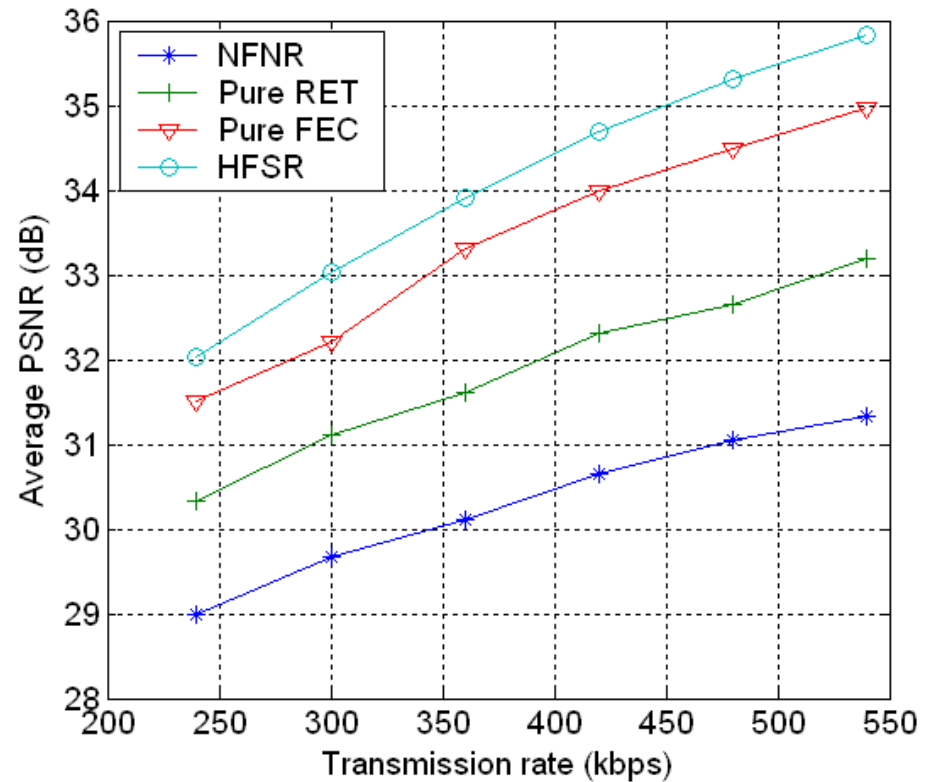
$R_T=480$  kbps,  $F=15$  fps , QCIF Foreman

# Sensitivity to Transmission Rate

Average PSNR vs. channel transmission rate  $R_T$



$$RTT = T_F$$



$$RTT = 3T_F$$

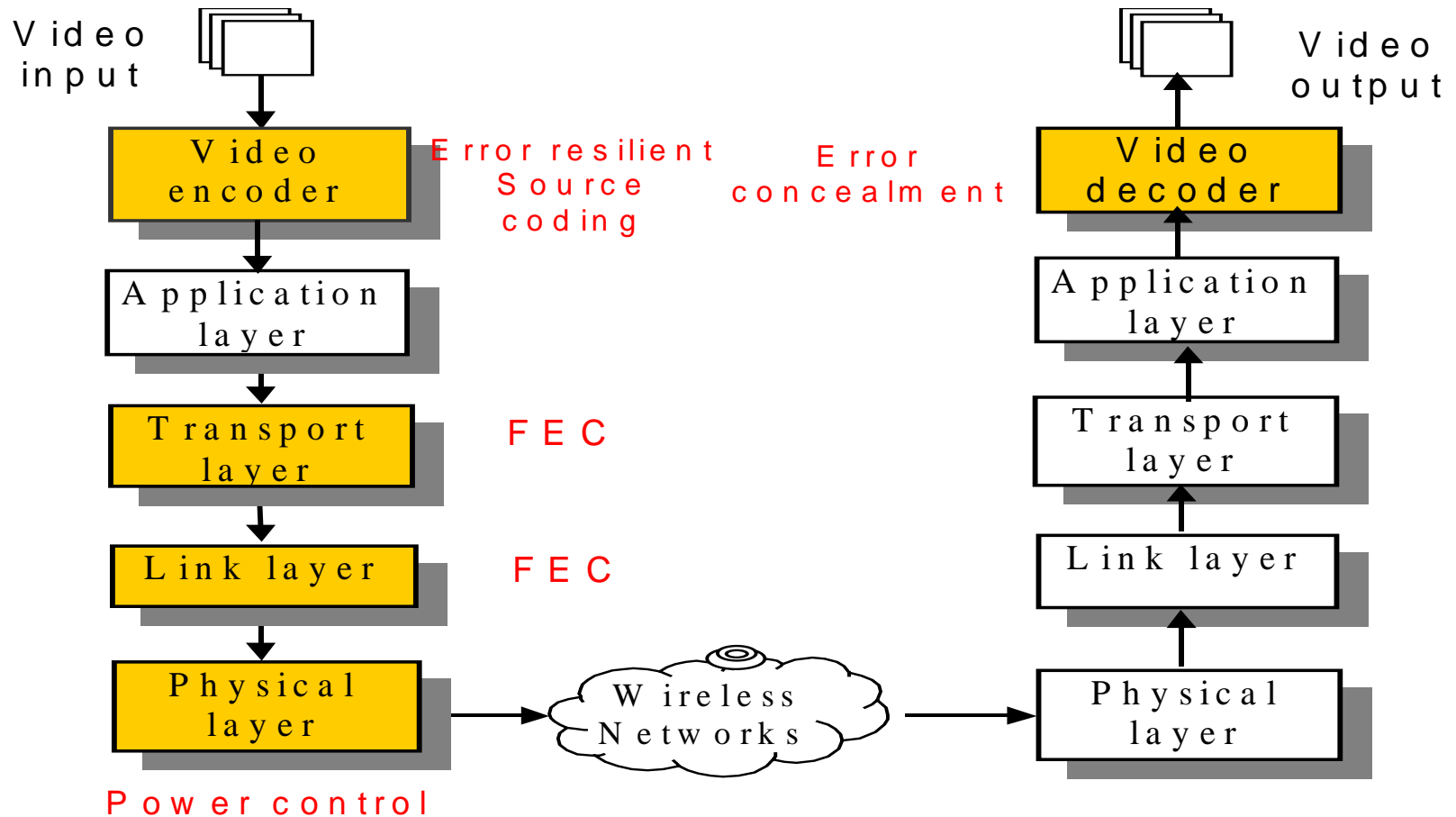
$\varepsilon=0.2$ ,  $F=15$  fps, QCIF Foreman

# JSCC– Conclusions

---

- Retransmission is suitable for short network RTT, low probability of packet loss, and low transmission rate.
- FEC is more suitable otherwise.
- The proposed hybrid FEC/selective retransmission scheme outperforms both (average gain: 0.7 dB in PSNR).
- In our simulations, we assume the CSI is accurately estimated, which favors FEC, because ARQ does not require accurate CSI.

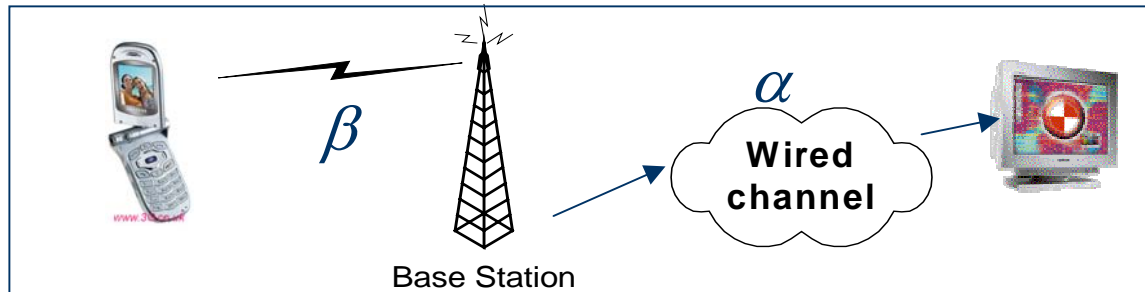
# Wireless Video Transmission



## ➤ JSCCPA– Jointly Source-Channel Coding and Power Allocation



# Hybrid Wireless Network



- At the IP level, the hybrid network can be modeled as the combination of two independent packet erasure channels: the wired part with loss rate  $\alpha$  and the wireless part with loss rate  $\beta$
- Overall probability of loss for a transport packet

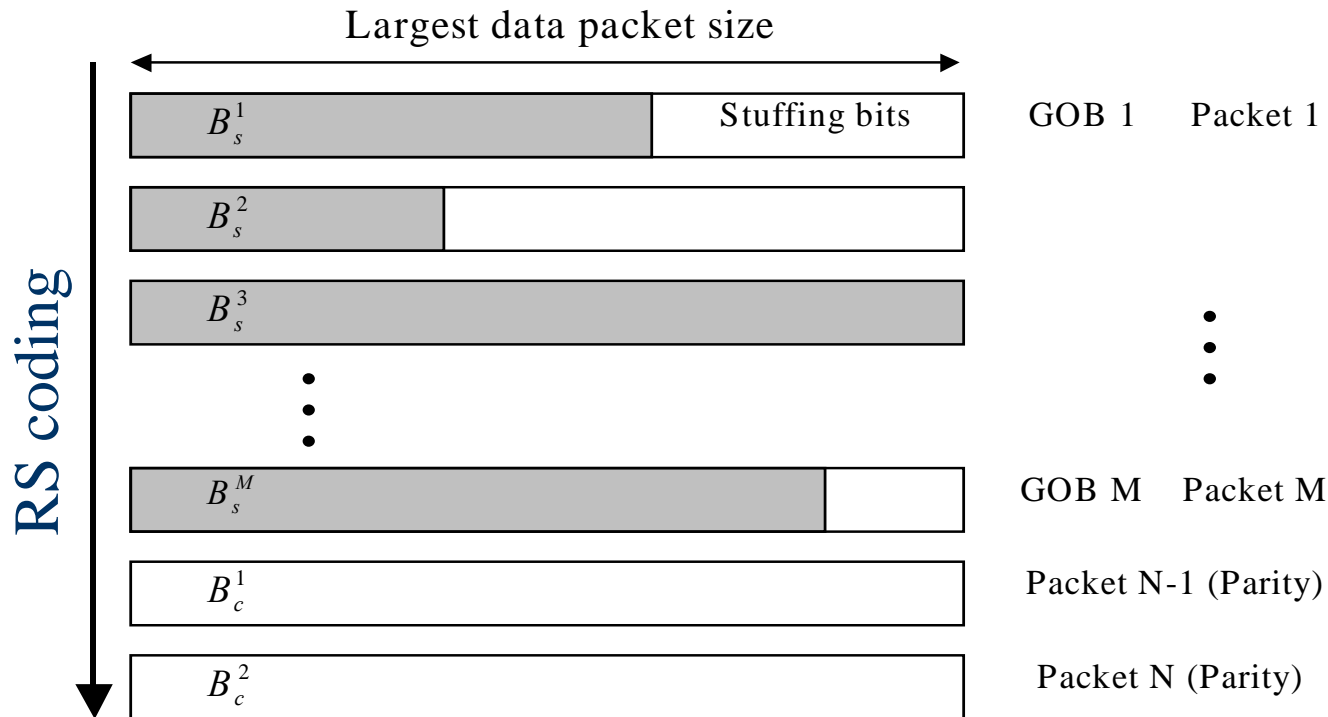
$$\varepsilon_k = \alpha + (1 - \alpha)\beta_k$$

- Probability of loss for a transport packet in the wireless channel

$$\beta_k(\mu_k, v_k, \eta_k) = 1 - (1 - p_b)^{B_k}$$

- We need Product FEC to combat different types of channel error

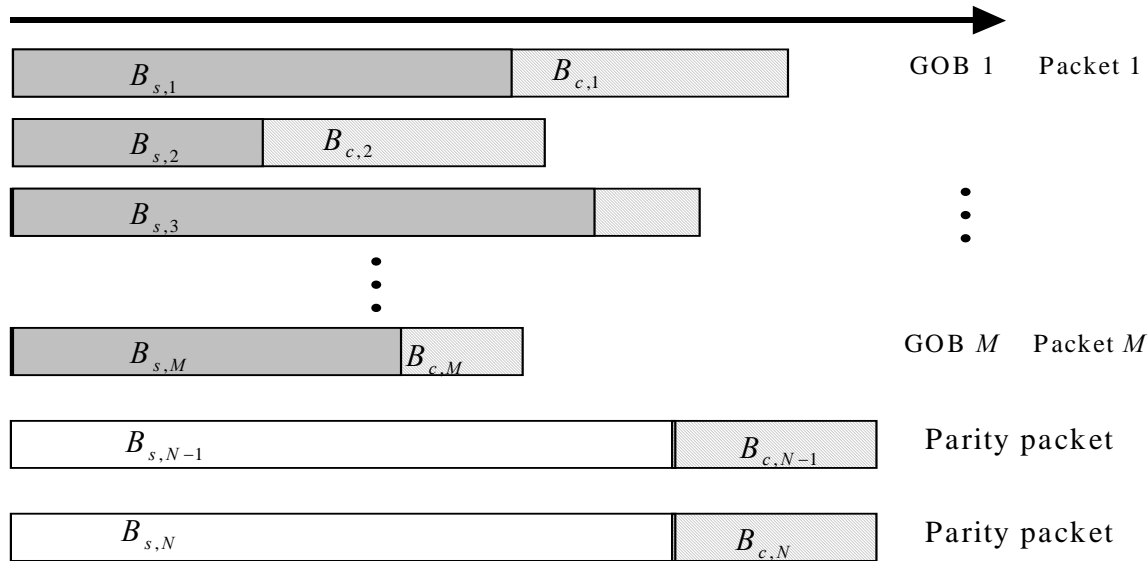
# Transport-Layer Packetization



- One row corresponds to one source packet
- Inter-packet FEC
- RS coding  $RS(N, M)$ —transport-layer FEC
- Transport-layer FEC parameter:  $\gamma$

# Link-Layer Packetization

## RCPC coding



- One row corresponds to one source/transport packet
- Intra-packet FEC
- RCPC coding—link-layer FEC
- Link-layer FEC parameter vector  $\mathbf{v} = \{v_1, \dots, v_N\}$

# Problem Formulation – Hybrid Network

$$\begin{aligned} & \min_{\{\boldsymbol{\mu}, \boldsymbol{\gamma}, \boldsymbol{v}, \boldsymbol{\eta}\}} E[D(\boldsymbol{\mu}, \boldsymbol{\gamma}, \boldsymbol{v}, \boldsymbol{\eta})] \\ \text{s.t.} \quad & C = \sum_{k=1}^{N(\boldsymbol{\gamma})} B_k P_k / R_T \leq C_0 \\ & T = \sum_{k=1}^{N(\boldsymbol{\gamma})} B_k / R_T \leq T_0 \end{aligned}$$

- $\boldsymbol{\mu} = \{\mu_1, \dots, \mu_M\}$  : Source coding parameter vector
- $\boldsymbol{\gamma} \in \{(N_1, M), \dots, (N_q, M)\}$  : Transport-layer FEC parameter
- $\boldsymbol{v} = \{v_1, \dots, v_N\}$  : Link-layer FEC parameter vector
- $\boldsymbol{\eta} = \{\eta_1, \dots, \eta_N\}$  : Power level parameter vector

# Solution Algorithm

## ➤ Lagrangian relaxation

$$L(\boldsymbol{\mu}, \boldsymbol{\gamma}, \boldsymbol{v}, \boldsymbol{\eta}, \lambda_1, \lambda_2) = E[D] + \lambda_1(C - C_0) + \lambda_2(T - T_0)$$

- Have proposed an iterative search algorithm to solve the above problem (Allerton'03)

## ➤ Minimization of the Lagrangian

$$g(\lambda_1, \lambda_2) = \min_{\{\boldsymbol{\mu}, \boldsymbol{\gamma}, \boldsymbol{v}, \boldsymbol{\eta}\}} L(\boldsymbol{\mu}, \boldsymbol{\gamma}, \boldsymbol{v}, \boldsymbol{\eta}, \lambda_1, \lambda_2)$$

# Solution Algorithm

- Lagrangian relaxation

$$L(\boldsymbol{\mu}, \boldsymbol{\gamma}, \boldsymbol{v}, \boldsymbol{\eta}, \lambda_1, \lambda_2) = E[D] + \lambda_1(C - C_0) + \lambda_2(T - T_0)$$

- Minimization of Lagrangian

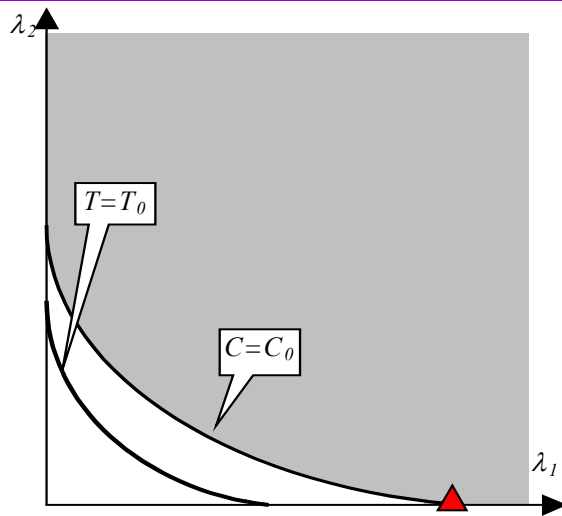
$$g(\lambda_1, \lambda_2) = \min_{\{\boldsymbol{\mu}, \boldsymbol{\gamma}, \boldsymbol{v}, \boldsymbol{\eta}\}} L(\boldsymbol{\mu}, \boldsymbol{\gamma}, \boldsymbol{v}, \boldsymbol{\eta}, \lambda_1, \lambda_2)$$

- Dual Problem

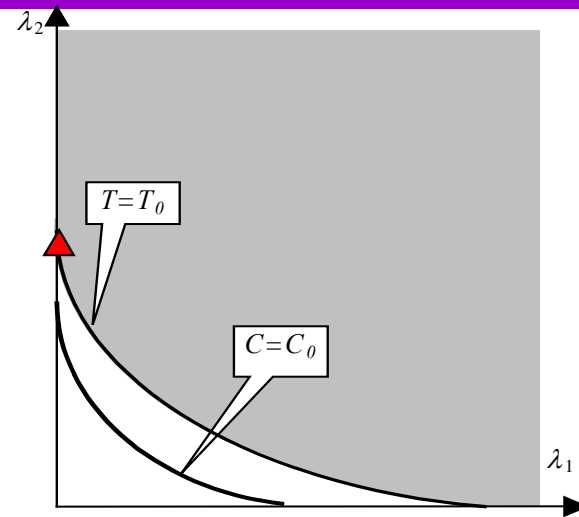
$$\max_{\{\lambda_1 \geq 0, \lambda_2 \geq 0\}} g(\lambda_1, \lambda_2)$$

- Goal: find the convex-hull solution
  - The dual problem is concave
  - complementary slackness applies

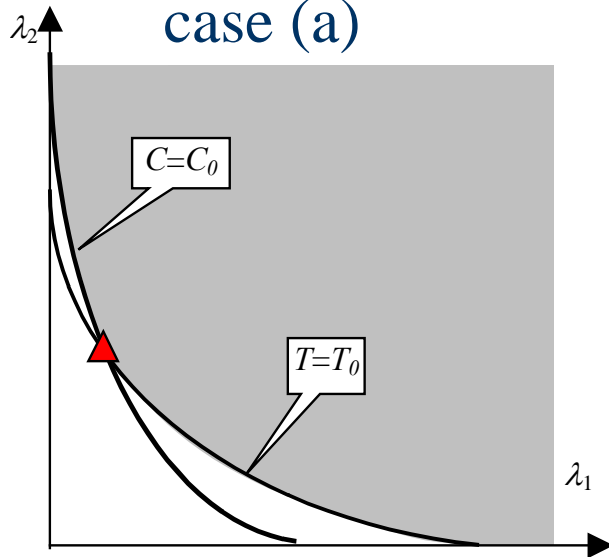
# Algorithm– Lagrangian relaxation



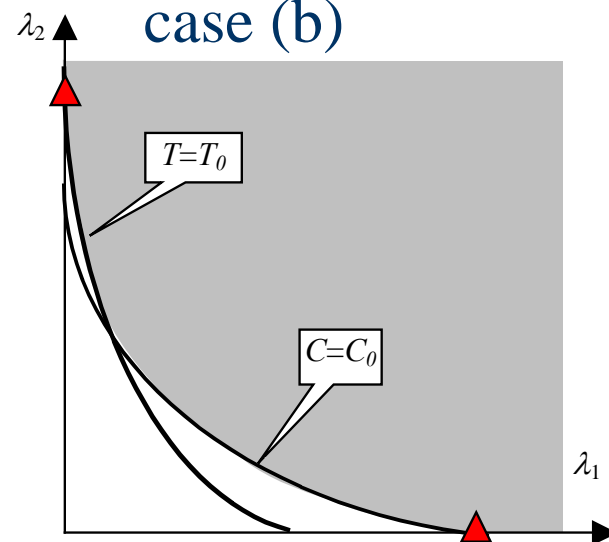
case (a)



case (b)



case (c)



case (d)

# Algorithm– Lagrangian Relaxation

**Step 1 (case a, d):** Let  $\lambda_2 = 0$ , find  $\lambda_1^*$  to satisfy  $C(H(\lambda_1^*, 0)) \leq C_0$ .

If  $T(H(\lambda_1^*, 0)) \leq T_0$ ,  $H(\lambda_1^*, 0)$  corresponds to the optimal solution. Otherwise,

**Step 2: (case b, d):** Let  $\lambda_1 = 0$ , find  $\lambda_2^*$  to satisfy  $T(H(0, \lambda_2^*)) \leq T_0$ .

If  $C(H(\lambda_1^*, 0)) \leq C_0$ ,  $H(0, \lambda_1^*)$  corresponds to the optimal solution. Otherwise,

**Step 3 (case c):**

- i. Let  $\lambda_1^l = 0$ ,  $\lambda_1^r = \lambda_1^*$ ,  $\lambda_2^b = 0$ ,  $\lambda_2^t = \lambda_2^*$ .
- ii.  $\lambda_1^m = (\lambda_1^l + \lambda_1^r) / 2$ , find  $\lambda_2^*$  within  $[\lambda_2^b, \lambda_2^t]$  to satisfy  $T(H(\lambda_1^m, \lambda_2^*)) \leq T_0$ .
- iii. If  $C(H(\lambda_1^m, \lambda_2^*)) > C_0$ , then let  $\lambda_1^l = \lambda_1^m$ ,  $\lambda_2^t = \lambda_2^*$ , and go to step 3ii. Otherwise,
- iv. If  $C(H(\lambda_1^m, \lambda_2^*)) < C_0 - \varepsilon$  ( $\varepsilon$  is a relatively small number), then let  $\lambda_1^r = \lambda_1^m$ ,  $\lambda_2^b = \lambda_2^*$ , and go to step 3ii. Otherwise,
- v. The optimal solution corresponds to  $H(\lambda_1^m, \lambda_2^*)$ .



# Calculation of Probability of Packet Loss

$$\rho_k(\boldsymbol{\mu}, \boldsymbol{\gamma}, \boldsymbol{\nu}, \boldsymbol{\eta}) = \sum_{j=N(\boldsymbol{\gamma})-M+1}^{N(\boldsymbol{\gamma})} P_b(N, j) = \sum_{j=N-M+1}^N \sum_{Q_j^t \in I_j(N, k)} \left( \prod_{i \in Q_j^t} \varepsilon_i \prod_{l \in \bar{Q}_j^t} (1 - \varepsilon_l) \right)$$

$Q_j^t(N)$ ,  $j = 1, \dots, N$ ,  $t = 1, \dots, \binom{N}{j}$  the  $t$ -th subset with  $j$  elements of  $Q(N)$

$$I_j(N, k) = \{Q_j^t \in Q(N) \mid k \in Q_j^t(N), |Q_j^t| = j\}$$

- Loss probability of a source packet depends on the parameters chosen for all the other packets in the frame
  - $\varepsilon_k$  differs from packet to packet
  - Inter-packet dependency introduced by inter-packet FEC

# Calculation of Probability of Packet Loss

$$\begin{aligned}\rho_k(\boldsymbol{\mu}, \boldsymbol{\gamma}, \boldsymbol{v}, \boldsymbol{\eta}) &= \sum_{j=N(\boldsymbol{\gamma})-M+1}^{N(\boldsymbol{\gamma})} P_b(N, j) \\ &= \sum_{j=N-M+1}^N \sum_{Q_j^t \in I_j(N, k)} \left( \prod_{i \in Q_j^t} \varepsilon_i \prod_{l \in \overline{Q}_j^t} (1 - \varepsilon_l) \right)\end{aligned}$$

$P_b(N, j)$  Probability that the  $k$ -th packet is correctly decoded by the RCPC decoder and the total number transport packets that are not correctly received from the group of  $N$  packets is  $j$

$$Q(3) = \{1, 2, 3\}, Q_1^1(3) = \{1\}, Q_1^2(3) = \{2\}, Q_1^3(3) = \{3\}$$

$$Q_2^1(3) = \{1, 2\}, Q_2^2(3) = \{1, 3\}, Q_2^3(3) = \{2, 3\}, Q_3^1(3) = \{1, 2, 3\},$$

# Minimization of the Lagrangian– Hybrid

- Iterative descent algorithm (alternating variables)

Step 1: Set  $\mathbf{x}^{(0)}$  corresponds to any initial state

Step 2: let  $t_n = (t \bmod N)$  (round-robin style)

if  $i \neq t_n$ , let  $x_i^{(n)} = x_i^{(n-1)}$  ; otherwise, for  $i = t_n$

$$x_i^{(t)} = \arg \min_{x_i^{(t)}} L(x_1^{(t)}, \dots, x_{i-1}^{(t)}, x_i, x_{i+1}^{(t)}, \dots, x_N^{(t)})$$

Step 3: If not convergence, go back to Step 2

$x = \{x_1, x_2, \dots, x_N\}$     Parameter selected for the  $N$  packets

$x^{(t)} = \{x_1^{(t)}, x_2^{(t)}, \dots, x_N^{(t)}\}$  for  $t = 0, 1, \dots$  at step  $t$

# Experimental Setup

---

## ➤ “Real Time” Applications

- Short-delay applications, where ARQ is not applicable
- Delay constraint equals one frame time

## ➤ Channel Model

- Flat Rayleigh fading channel with AWGN
- i.i.d. channel fading
- $RTT = 2T_F$ , which preclude retransmission

# Experimental Setup

Performance of RCPC (in BER) over a Rayleigh fading channel with interleaving

SNR(dB)	2	6	10	14	18
Cr = 1/2	$1.4 \cdot 10^{-3}$	$2.2 \cdot 10^{-5}$	$2.1 \cdot 10^{-6}$	$2.4 \cdot 10^{-7}$	$6.4 \cdot 10^{-8}$
Cr = 4/7	$1.1 \cdot 10^{-1}$	$5.3 \cdot 10^{-4}$	$4.1 \cdot 10^{-5}$	$1.2 \cdot 10^{-5}$	$3.8 \cdot 10^{-6}$
Cr = 2/3	$3.2 \cdot 10^{-1}$	$7.4 \cdot 10^{-3}$	$1.7 \cdot 10^{-4}$	$3.5 \cdot 10^{-5}$	$1.2 \cdot 10^{-5}$
Cr = 4/5	$4.2 \cdot 10^{-1}$	$4.0 \cdot 10^{-2}$	$6.6 \cdot 10^{-4}$	$1.1 \cdot 10^{-4}$	$3.6 \cdot 10^{-5}$

- Generator polynomial (133, 171), mother code rate  $\frac{1}{2}$ , and puncturing rate 4
- Soft Viterbi decoding in conjunction with BPSK

# PFEC vs. Link-Layer FEC

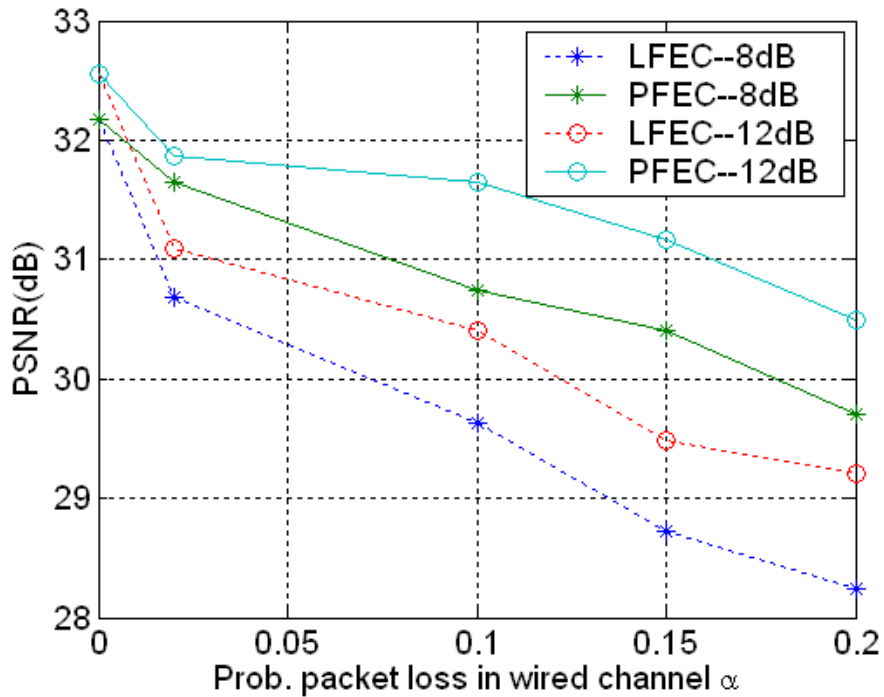
- Approach 1 – Link Layer FEC (LFEC)

$$\min_{\{\mu, \nu\}} E[D(\mu, \gamma_0, \nu, \eta_0)]$$

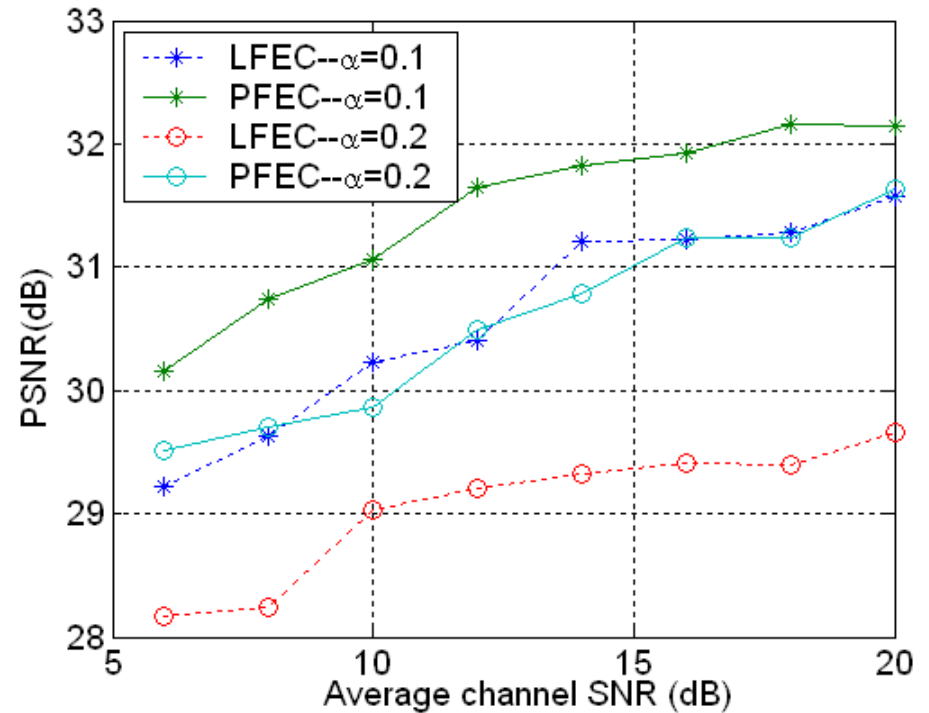
- Approach 2 – Product FEC (PFEC)

$$\min_{\{\mu, \gamma, \nu\}} E[D(\mu, \gamma, \nu, \eta_0)]$$

# PFEC vs. LFEC



Average PSNR vs.  $\alpha$



Average PSNR vs. channel SNR

QCIF Foreman, 30 fps

# UEP vs. EEP

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- Approach 1 – EEP-PFEC

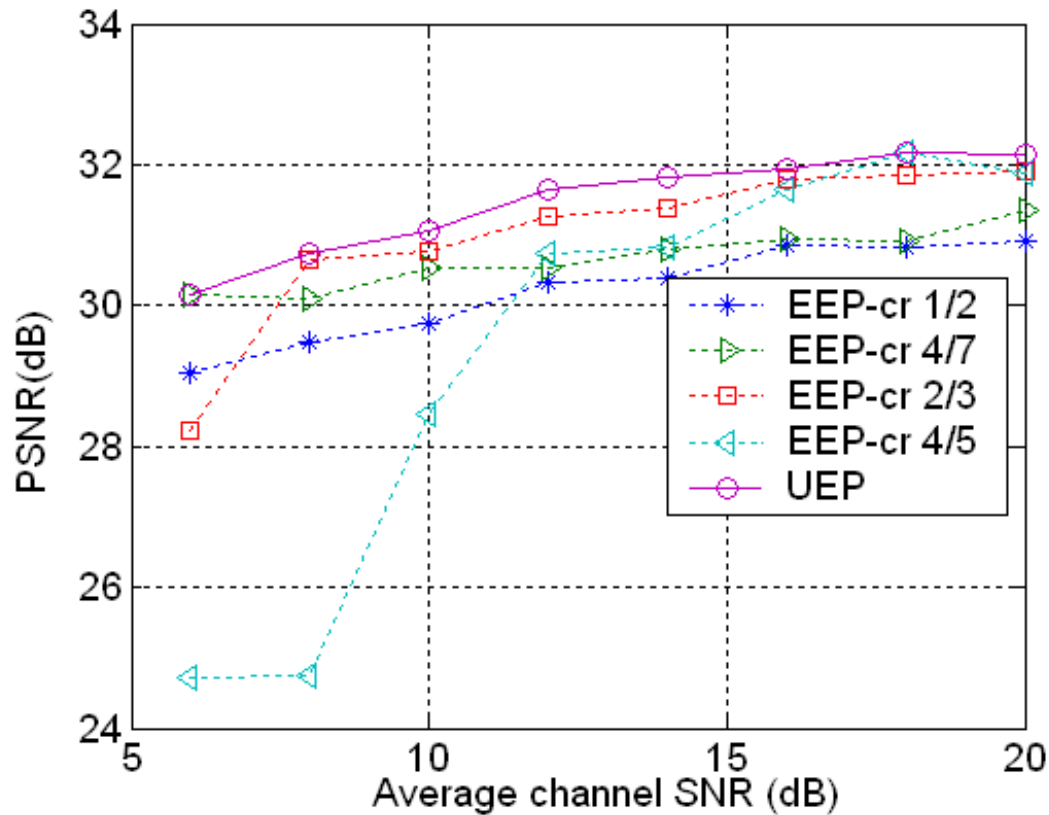
$$\min_{\{\mu, \gamma\}} E[D(\mu, \gamma, \mathbf{v}_0, \boldsymbol{\eta}_0)]$$

- Approach 2 – UEP-PFEC

$$\min_{\{\mu, \gamma, \mathbf{v}\}} E[D(\mu, \gamma, \mathbf{v}, \boldsymbol{\eta}_0)]$$



# UEP vs. EEP



PSNR vs. channel SNR,  $\alpha=0.1$

QCIF Foreman, 30 fps

# Problem Formulations—Wireless Channel

---

$$\min_{\{\boldsymbol{\mu}, \mathbf{v}, \boldsymbol{\eta}\}} E[D(\boldsymbol{\mu}, \mathbf{v}, \boldsymbol{\eta})]$$

$$\text{s.t.} \quad C = \sum_{k=1}^M B_k P_k / R_T \leq C_0$$

$$T = \sum_{k=1}^M B_k / R_T \leq T_0$$

# Solution Algorithm— Wireless

---

- Lagrangian relaxation

$$L(\boldsymbol{\mu}, \boldsymbol{v}, \boldsymbol{\eta}) = \sum_{k=1}^M J_k = \sum_{k=1}^M E[D_k] + \lambda_1 C_k + \lambda_2 T_k$$

- Dynamic Programming

$$J_k = J_k(\boldsymbol{\mu}_{k-1}, \boldsymbol{v}_{k-1}, \boldsymbol{\eta}_{k-1}, \boldsymbol{\mu}_k, \boldsymbol{v}_k, \boldsymbol{\eta}_k)$$

# Simulation 1– Wireless Channel

---

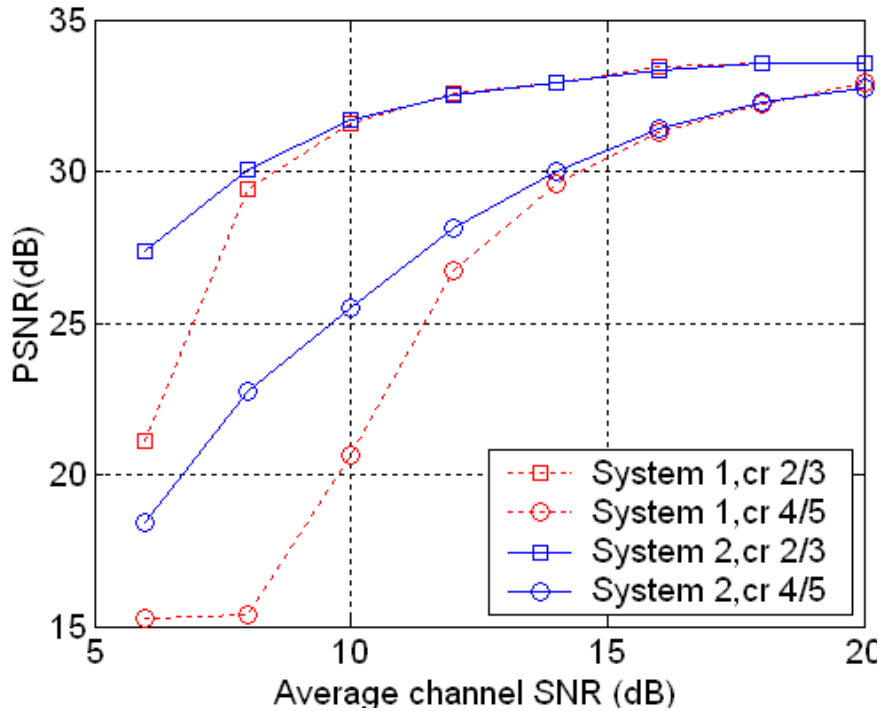
- Goal: To show the benefit by adjusting power levels
- System 1 -- Optimal error-resilient source coding

$$\min_{\{\mu\}} E[D(\mu, \mathbf{v}_0, \boldsymbol{\eta}_0)]$$

- System 2 -- Joint source coding and power allocation

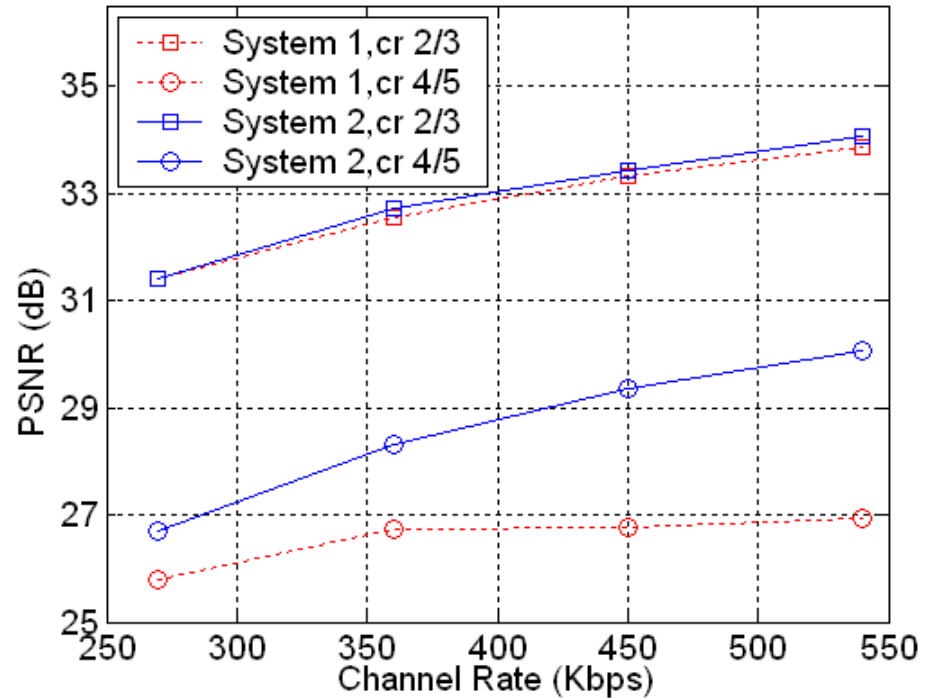
$$\min_{\{\mu, \boldsymbol{\eta}\}} E[D(\mu, \mathbf{v}_0, \boldsymbol{\eta})]$$

# Simulation 1



PSNR vs. channel SNR

(Transmission rate = 360 kbps)



PSNR vs. transmission rate

(reference SNR = 12 dB)

QCIF Foreman, 30 fps

# Simulation 1

Allocation of power level (1,2,3,4,5) in percentage in JSCPA system

Ref SNR(dB)	6	12	20
Cr = 1/2	(2.4,18.5,73.9,5.1,0)	(12.6,32.4,34,19.6,1.4)	(62.3,0,12.9,0.24,8)
Cr = 4/7	(18,0,14.3,66.1,1.6)	(2.3,29.9,56.4,11,0.3)	(10,35,39.2,13.4,2.3)
Cr = 2/3	(40,0,0,13,47)	(0.7,14,66,18.7,0.6)	(11.6,10.8,69.1,6.9,1.6)
Cr = 4/5	(45.8,0,0,0,54.2)	(2,4,41.8,47.3,5)	(8.2,31.5,43.8,15.3,1.3)

- The reference power level 3
- Each value here denotes the percentage of packets using the corresponding transmission power level.
- The transmission power is proportional to the power level parameter

# Simulation 2

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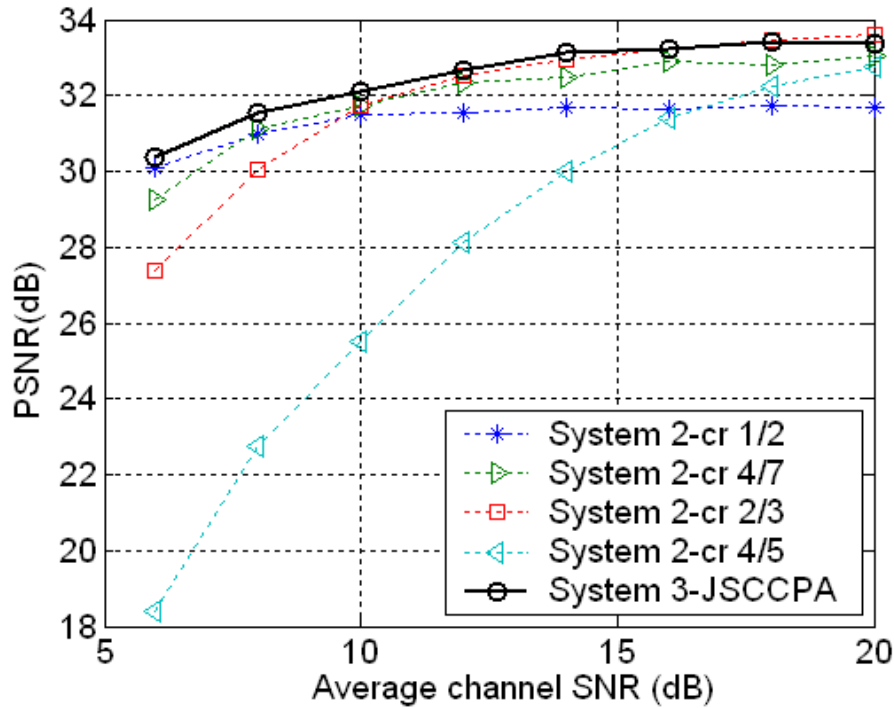
- Goal: To show the benefit by adjusting channel rates
- System 2 -- Joint source coding and power allocation

$$\min_{\{\mu, \eta\}} E[D(\mu, \mathbf{v}_0, \eta)]$$

- System 3 -- JSCC and power allocation (JSCCPA)

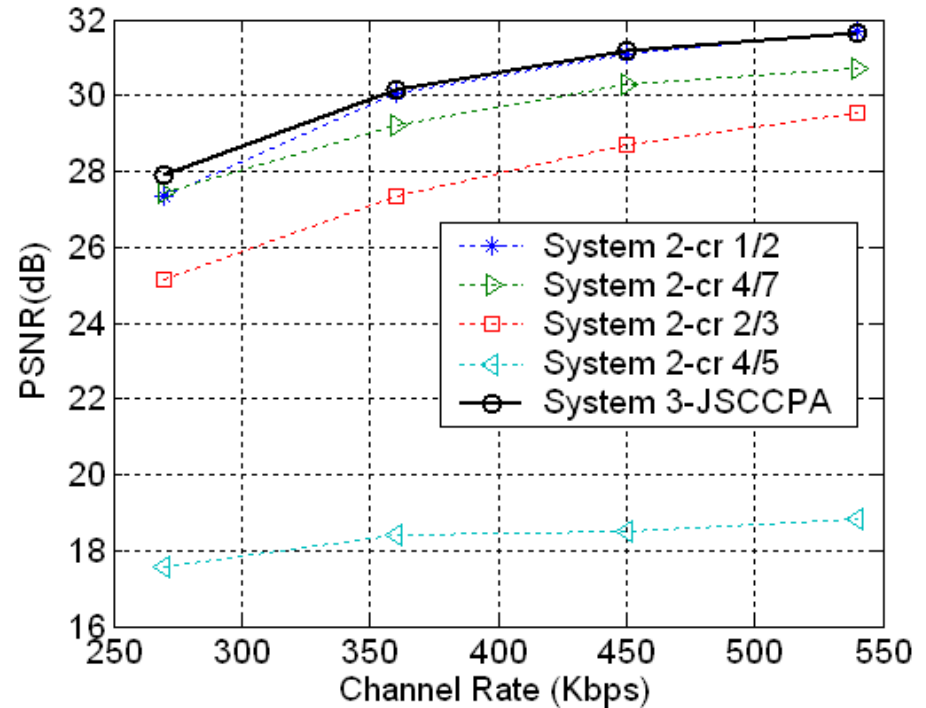
$$\min_{\{\mu, \mathbf{v}, \eta\}} E[D(\mu, \mathbf{v}, \eta)]$$

# Simulation 2



PSNR vs. channel SNR

(transmission rate = 360 kbps)



PSNR vs. transmission rate

(reference SNR = 12 dB)

QCIF Foreman, 30 fps



# Simulation 2

Channel Coding rates in percentages in JSCCPA system

Ref SNR(dB)	6	8	10	12	14	16	18	20
Cr = 1/2	96.2	67.7	41.2	19.6	6.7	4.7	1.0	1.6
Cr = 4/7	3.8	31.9	57.3	69.6	61.3	35.0	17.8	5.6
Cr = 2/3	0	0.4	1.5	10.8	31.3	57.7	73.9	69.6
Cr = 4/5	0	0	0	0	0.7	2.6	7.3	23.2

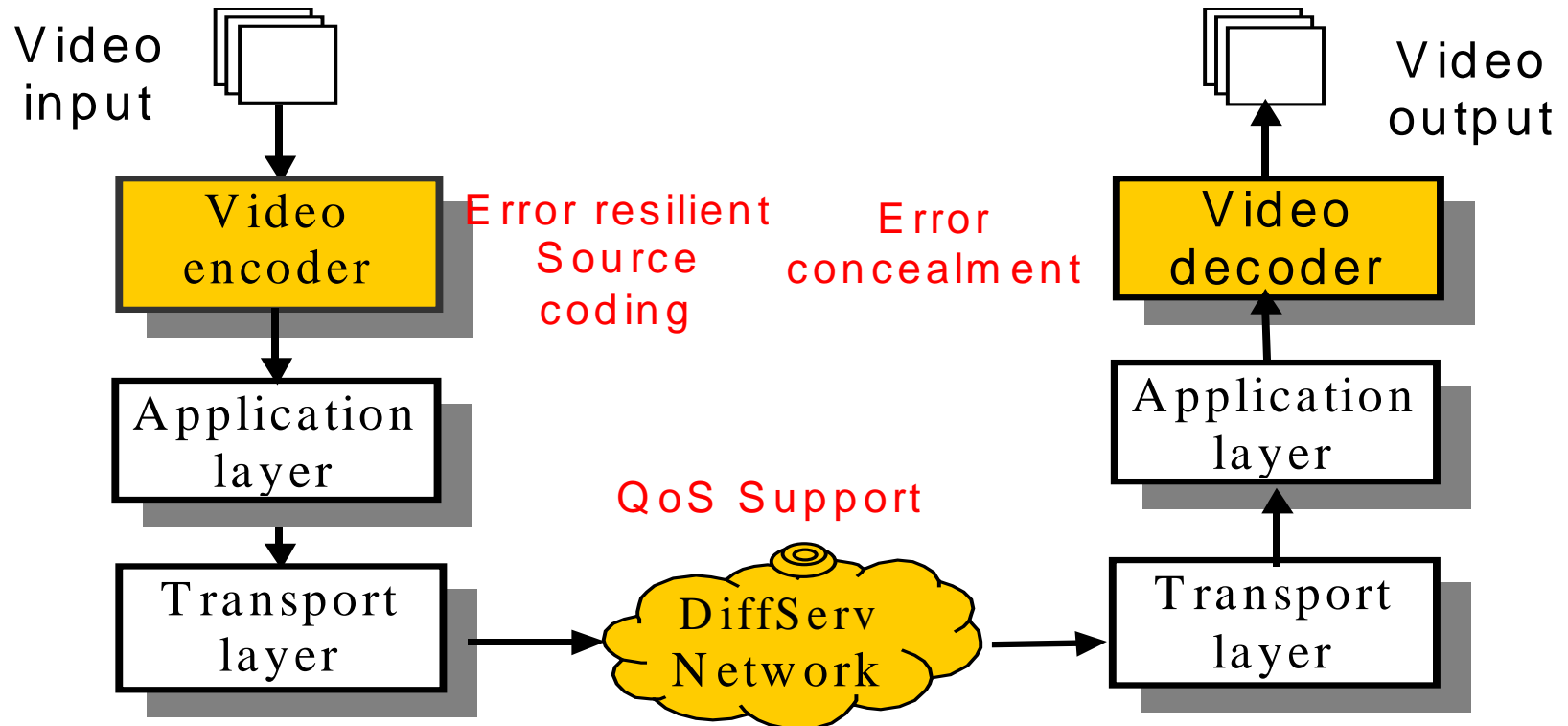
- The better channel, the less channel coding protection

# JSCCPA– Conclusions

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- Optimal cross-layer resource allocation to provide UEP for wireless video transmission
  - Error resilient source coding
  - Transport-layer FEC
  - Link-layer FEC
  - Physical-layer power allocation
  - Error concealment
- Transport-layer FEC (inter-packet FEC) is not necessary if the wired link has no error.
- Adapting either power or link-layer FEC (instead both) may be adequate to achieve the near-optimal result in some situations.

# DiffServ Video Transmission



## ➤ JSCPC– Jointly Source Coding and Packet Classification

# Differentiated Services Networks

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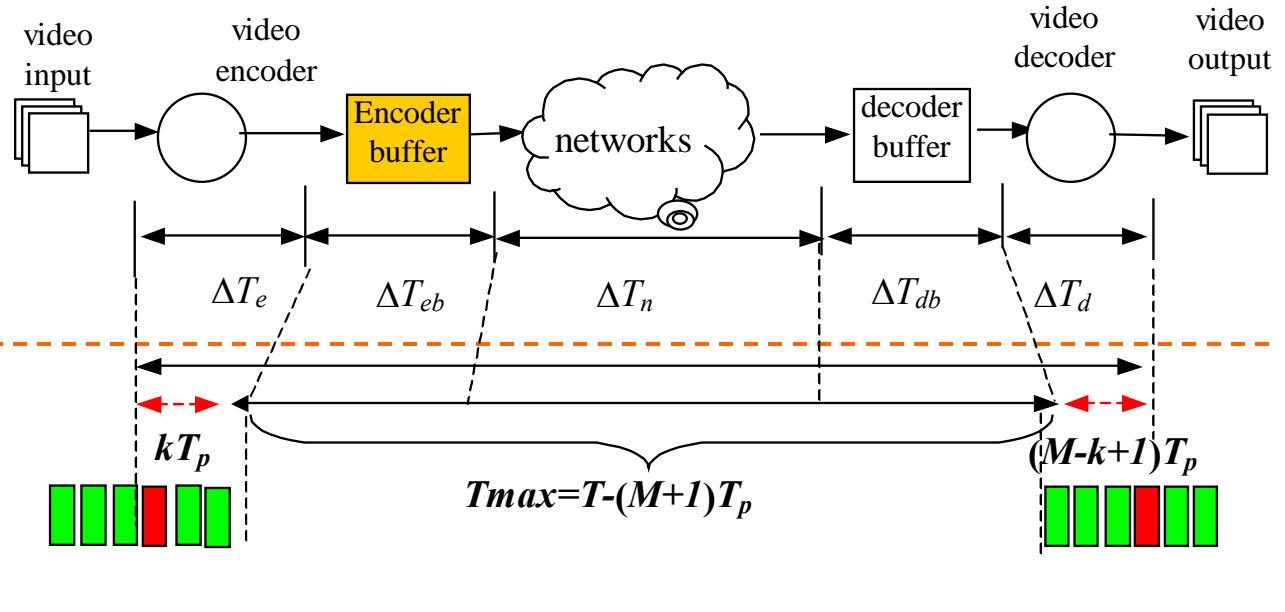
- Allocating resources discriminatorily to aggregated traffic flows
- Multiple service classes: different class has different end-to-end statistical behavior
- Sender is charged for each transmitted bit based on its service class

# Delay Components

- Constant end-to-end delay:

$$T = \Delta T_e + \Delta T_{eb} + \Delta T_n + \Delta T_{db} + \Delta T_d$$

- Frame level



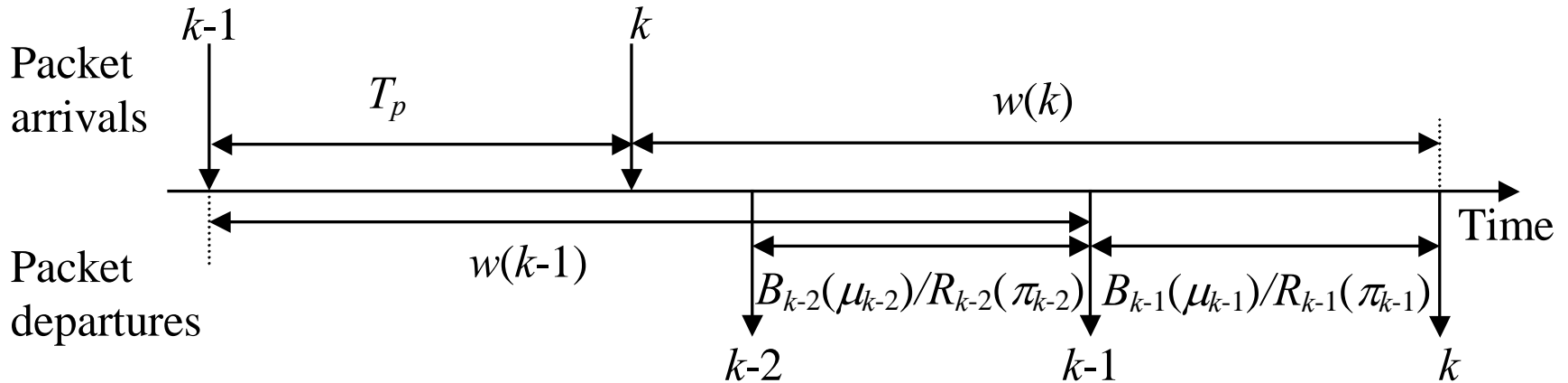
- Packet level  
An example  
( $M=6, k=3$ )

$$\Delta T_{eb}(k) + \Delta T_n(k) + \Delta T_{db}(k) = T - (M+1)T_p = T_{max}$$

to avoid excessive delay:  $\Delta T_{db}(k) \geq 0 \Rightarrow$

$$\Delta T(k) = \Delta T_{eb}(k) + \Delta T_n(k) \leq T_{max}$$

# Delay Components—cont.



encoder buffer delay: 
$$\Delta T_{eb}(k) = w(k) + \frac{B_k(\mu_k)}{R_k}$$

waiting time: 
$$w(k) = w(k-1) + \frac{B_{k-1}(\mu_{k-1})}{R_{k-1}} - T_p$$

max allowable network delay: 
$$\tau(k) = T_{\max} - \Delta T_{eb}(k) = T_{\max} - w(k) - \frac{B_k(\mu_k)}{R_k}$$

# JSCPC– Solution Algorithm

➤ Lagrangian relaxation 1

$$\min_{\{\boldsymbol{\pi}, \boldsymbol{\mu}\}} L(\boldsymbol{\pi}, \boldsymbol{\mu}, \lambda_1, \lambda_2) = \sum_{k=1}^M \{D_k + \lambda_1 C_k + \lambda_2 T_k\}$$

➤ Lagrangian relaxation 2

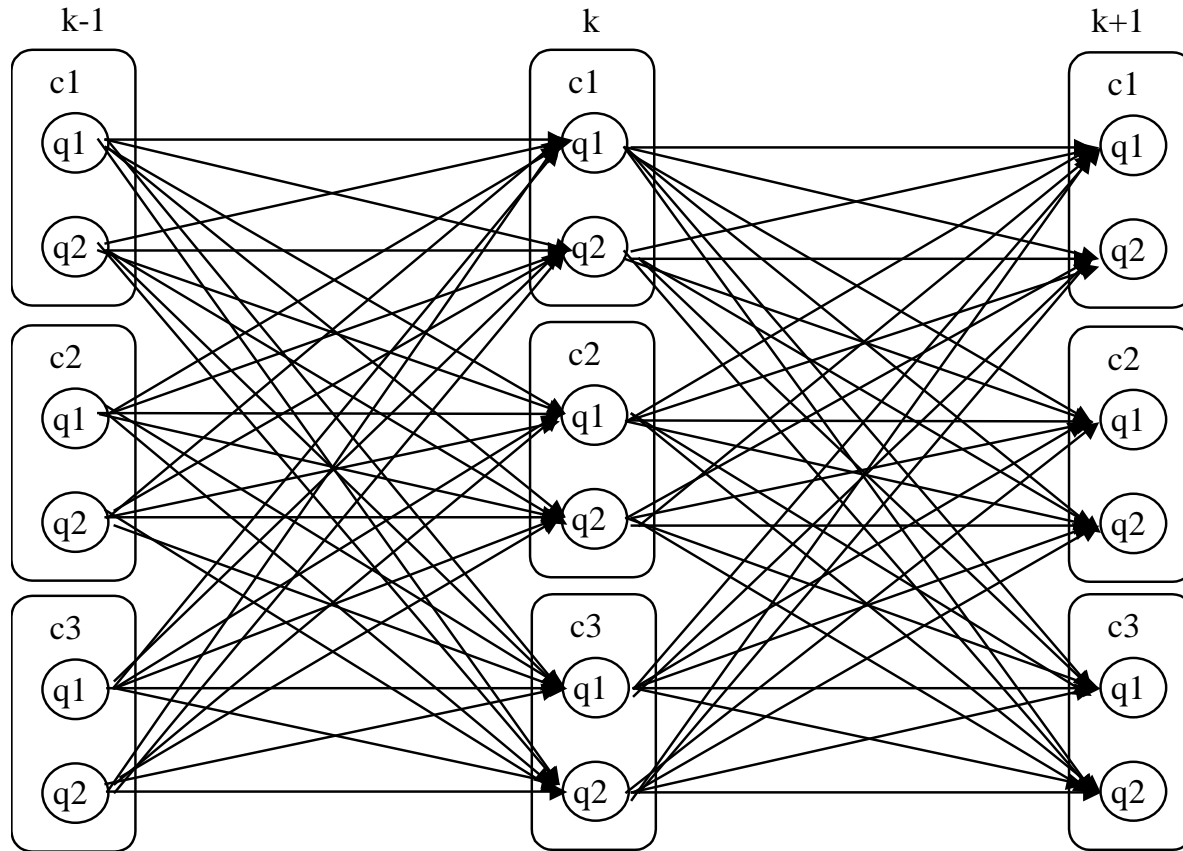
$$\begin{aligned} \min_{\{\boldsymbol{\mu}, \boldsymbol{\pi}\}} L(\boldsymbol{\mu}, \boldsymbol{\pi}, \lambda) &= \min_{\{\boldsymbol{\mu}, \boldsymbol{\pi}\}} \sum_{k=1}^M D_k + \lambda C_k \\ \text{s.t.} \quad \sum_{k=1}^M T_k &\leq T_0 \end{aligned}$$

➤ Dependency of the Lagrangian

$$\min_{\{\boldsymbol{\pi}, \boldsymbol{\mu}\}} L(\boldsymbol{\pi}, \boldsymbol{\mu}, \lambda_1, \lambda_2) = \sum_{k=1}^M J_k$$

$$J_k = J_k(\pi^1, \mu^1, \dots, \pi^{k-1}, \mu^{k-1}, \pi^k, \mu^k)$$

# JSCPC— Solution Algorithm contd.





# JSCPC— Solution Algorithm contd.

$$J_k = J_k(\pi_1, \mu_1, \dots, \pi_{k-1}, \mu_{k-1}, \pi_k, \mu_k)$$

$$E[D_k] = (1 - \rho_k) E[D_{R,k}] + \rho_k (1 - \rho_{k-1}) E[D_{C,k}] + \rho_k \rho_{k-1} E[D_{Z,k}]$$

$$J_k = J_k(\mu^{k-1}, \pi^{k-1}, \mu^k, \pi^k, w(k))$$

- Do DP with respect to “**system state**”,  $w(k)$ , instead of to choice of source coding parameters and service classes

$$\rho_k = \varepsilon + (1 - \varepsilon) P\{\Delta T_n(k) > \tau(k)\}$$

$$\tau(k) = T_{\max} - \Delta T_{eb}(k) = T_{\max} - w(k) - \frac{B_k(\mu_k)}{R_k} = T_{\max} - w(k+1) + T_p$$

$$w(k) = w(k-1) + \frac{B_{k-1}(\mu_{k-1})}{R_{k-1}} - T_p$$

# JSCPC— Solution Algorithm contd.

$$s(k) = \sum_{j=1}^k T(j) : \quad \sum_{k=1}^M T(k) \leq T_0 \Rightarrow \quad s(M) \leq T_0$$

$$\begin{aligned} w(k) &= w(k-1) + T(k-1) - T_p \\ &= w(1) + \sum_{j=1}^{k-1} T(j) - (k-1)T_p = w(1) + s(k-1) - (k-1)T_p \end{aligned}$$

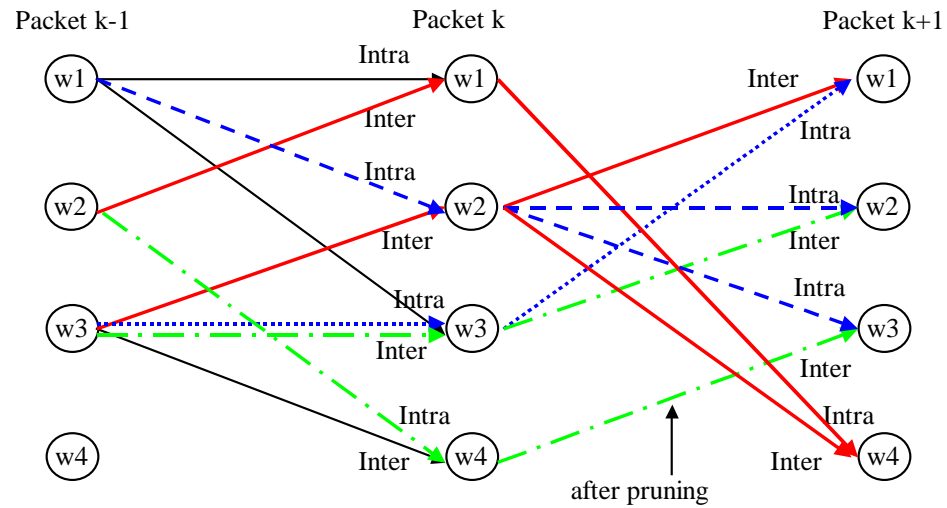
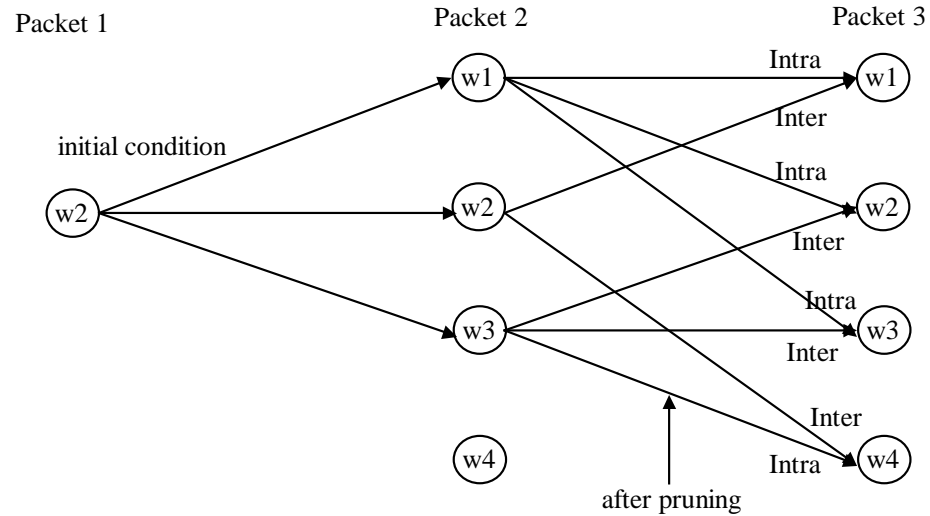
$$w(k) - w(1) + (k-1)T_p \leq T_0 \quad \text{for } k = 1, \dots, M$$

## ➤ DP solution

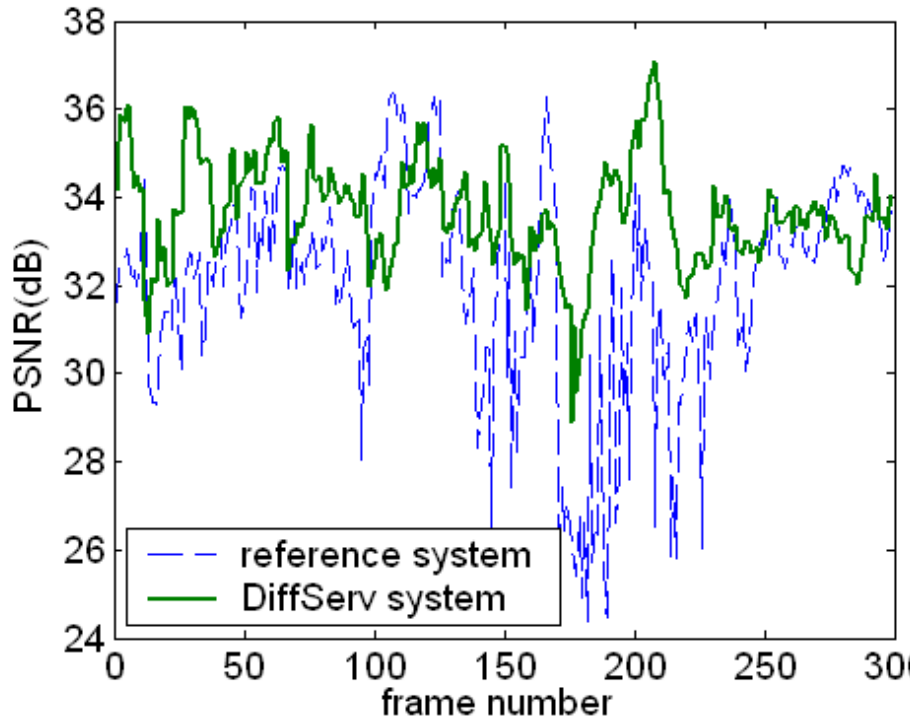
$$\min_{\{u(k) \in \mathbf{U}(w(k))\}} \sum_{k=1}^M J(k) = \sum_{k=1}^M J(\mu_{k-1}, \pi_{k-1}, \mu_k, \pi_k, w(k))$$

$$\mathbf{U}(w(k)) = \{u(k) \in \mathbf{\Pi} \times \mathbf{Q} : 0 \leq \frac{B_k(\mu_k)}{R_k(\pi_k)} + w(k) - T_p \leq \min(T_{\max}, T_0 + w(1) - kT_p)\}$$

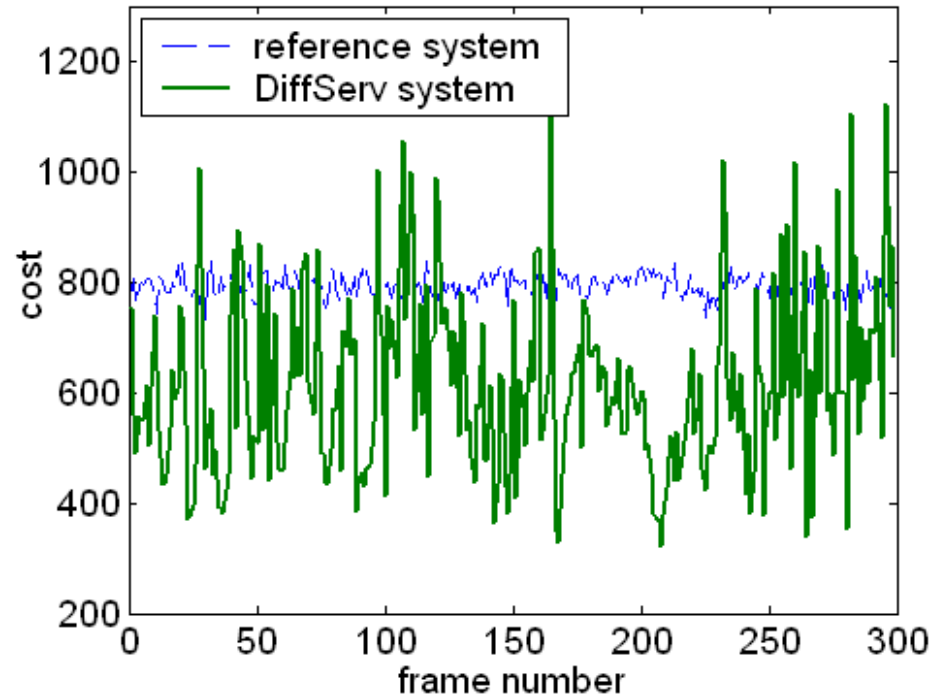
# JSCPC— Solution Algorithm contd.



# JSCPC — Experimental Results



Minimum distortion approach

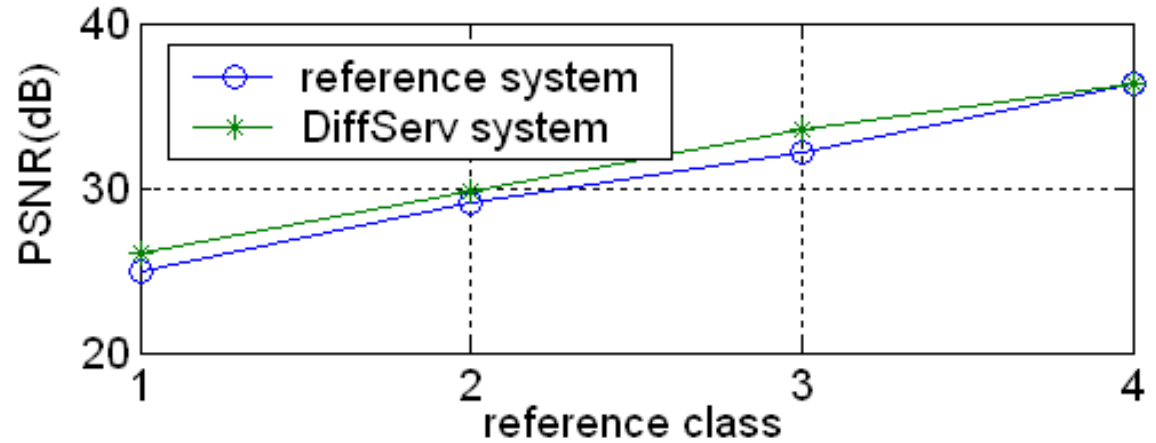


Minimum cost approach

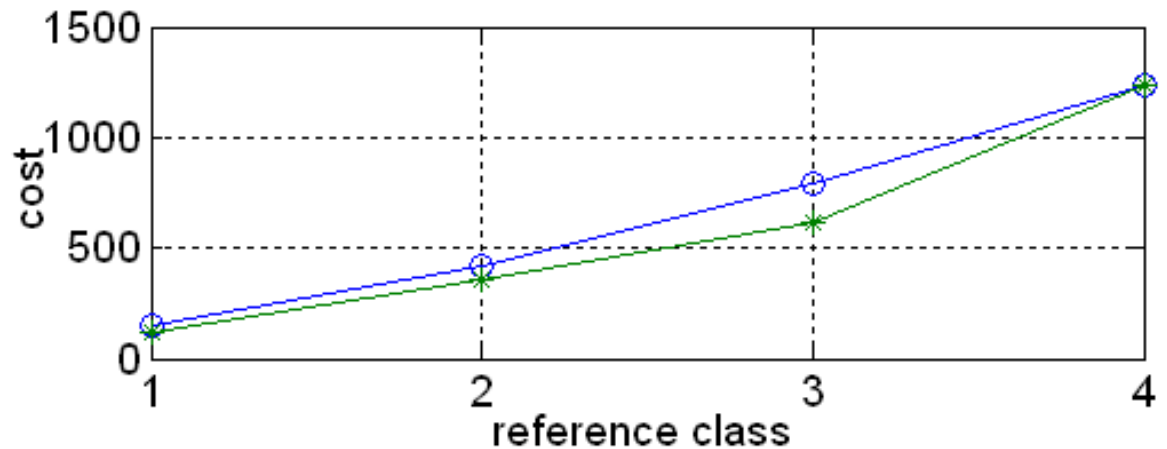
QCIF Foreman,  $F=30$  fps, reference class = 3

# JSCPC — Experimental Results

minimum distortion  
approach

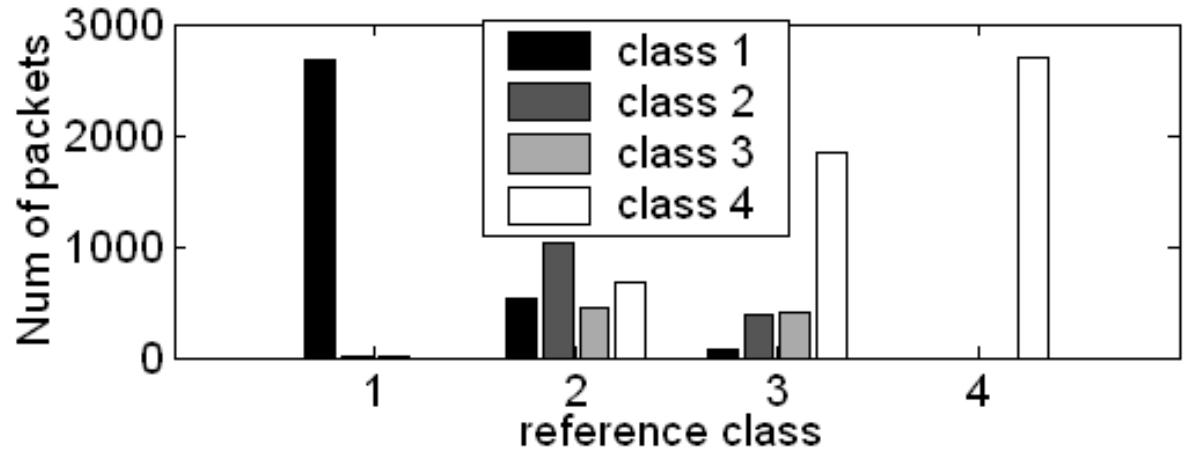


minimum cost  
approach

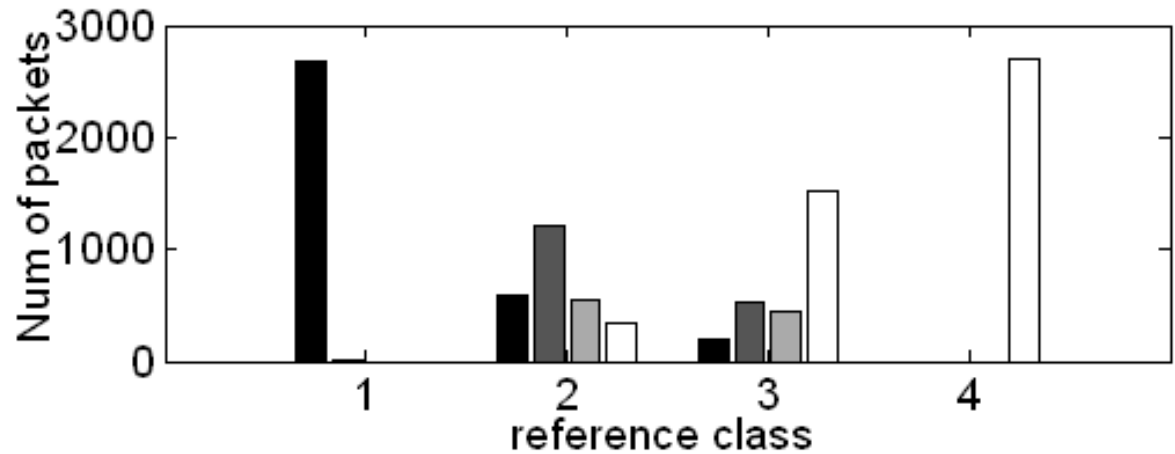


# Distribution of packet classification

minimum distortion  
approach

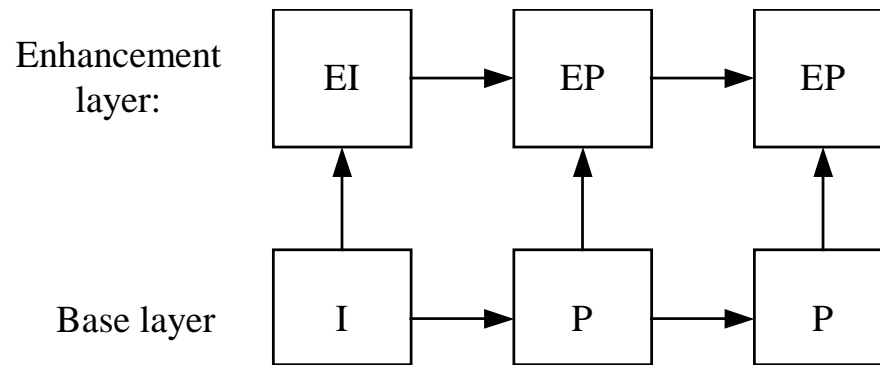


minimum cost  
approach



# Internet Scalable Video

- JSCC based on H.263+ SNR scalability codec

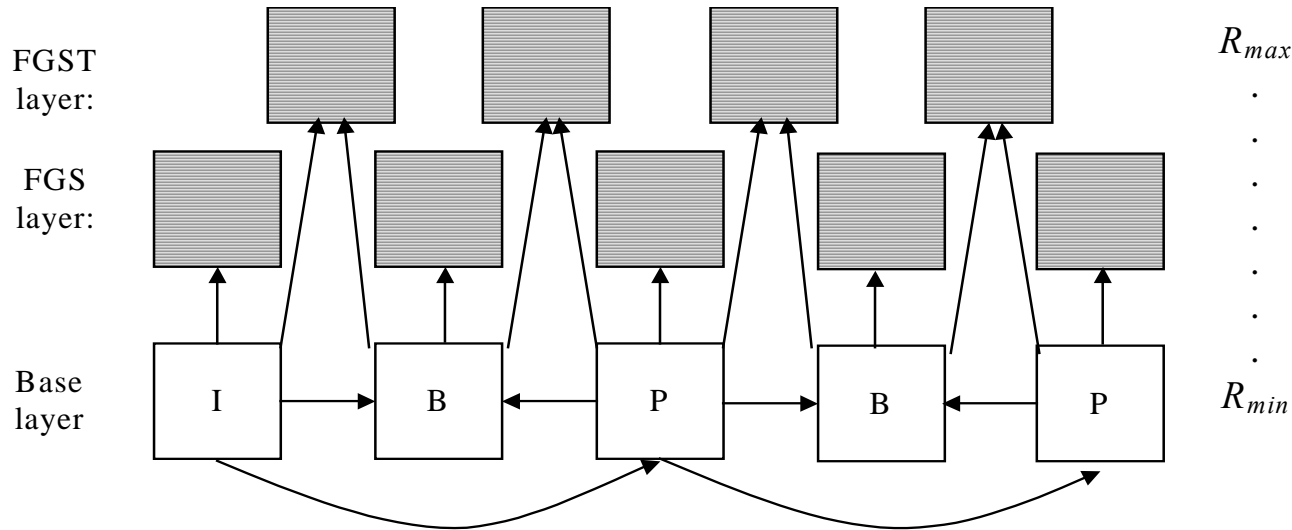


- Problem Formulation

$$\min_{\{\mu^{(b)}, \nu^{(b)}, \mu^{(e)}, \nu^{(e)}\}} E[D] = E[D^{(b)}(\mu^{(b)}, \nu^{(b)})] + E[D^{(e)}(\mu^{(e)}, \nu^{(e)})]$$
$$\text{s.t. } T^{(b)} \leq T_0^{(b)}$$
$$T^{(b)} + T^{(e)} \leq T_0$$

# Wireless Scalable Video

- Optimal power allocation based on MPEG-4 FGS codec



- Problem Formulation

$$\min_{\{P_i^{(e)}\}} \left\{ D^{(b)} - E[\Delta^{(e)}] \right\}$$

s.t.  $E_{tot} = E_0$

- Assume base layer is pre-encoded and always received correctly



# Conclusions

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- Objective:
  - Optimal error control to achieve the best video quality
- Method:
  - Optimal cross-layer resource allocation
- Base:
  - Resource-distortion optimization framework
- Scope:
  - End-system design for video communication
- Have studied following applications:
  - JSCC—Internet video (Allerton'03, ICC'04, ICIP'04, TIM'04)
  - JSCCPA—wireless video (Allerton'03, ICASSP'04)
  - JSCPC— DiffServ video (ICIP'03, TMM'04)
  - Scalable video (ICME'03, ICC'04)

# Future Work

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- Rate control
  - Incorporate rate control into the optimization framework
- Channel model
  - Take into account the correlation of packet loss, e.g., Gilbert model
- Cross-layer Design
  - Consider modifying the current network protocols:  
Link-layer ARQ, UDPLite,

# Q&A

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**Thank You!**