



Recent Advances in Video Transmission: Challenges and Opportunities

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Plethora of exciting applications!

- Freedom of Multimedia Access (FOMA)
- Exciting Applications
- TiVo2Go
- Sales of phone digital cameras surpassed sales of digital cameras in 2001
- Seamless mobility, Ubiquitous access, UMA/UME



Video is the protagonist!

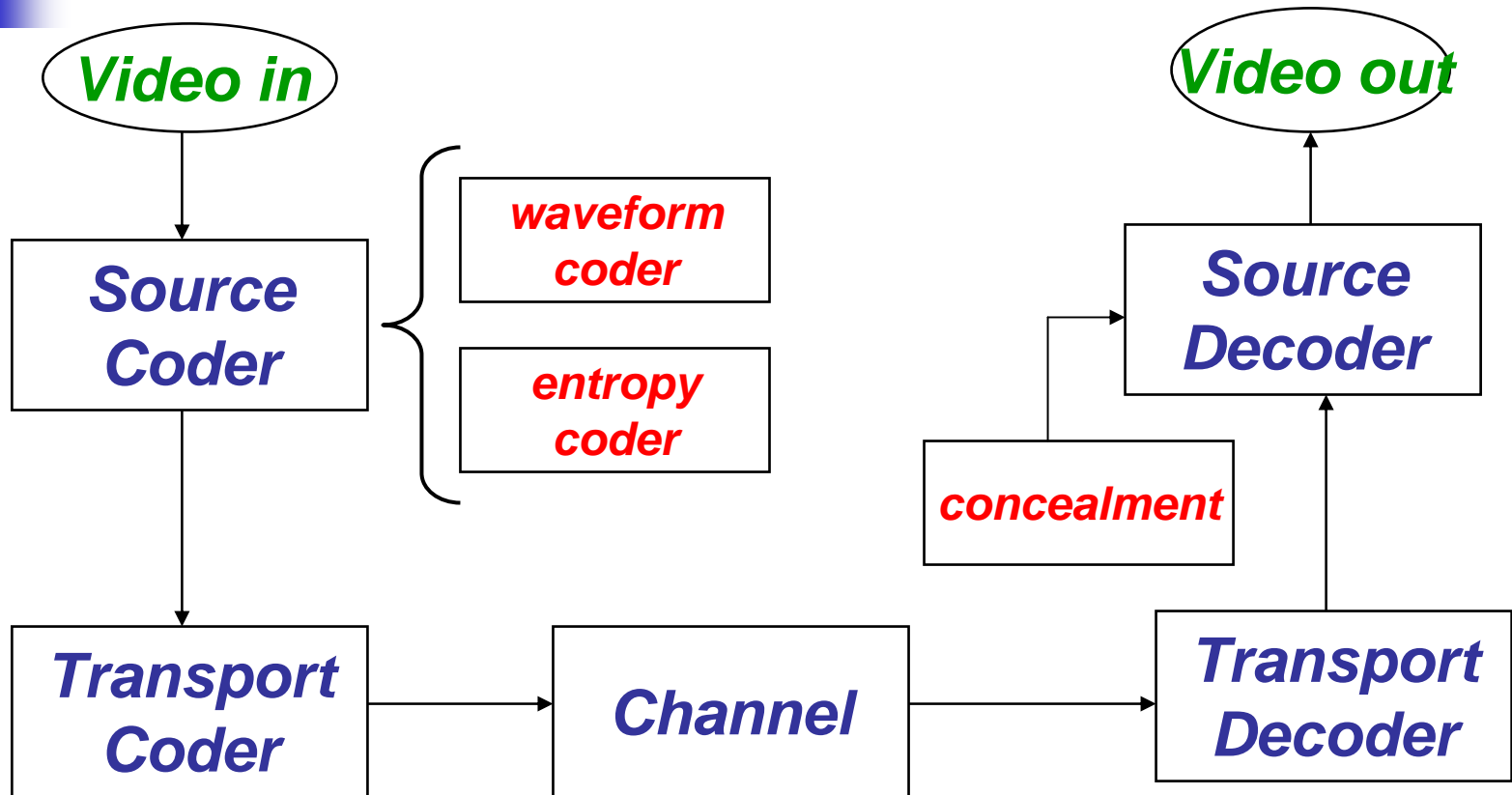
The Five+ Sense Communication Interface	Contribution to Human Perception
Visual	60%
Auditory	20%
Tactile	15%
Taste	3%
Smell	2%
Existence/Emotion	?%



Talk Objectives

- Describe recent advances in video transmission
- Towards this goal, describe in some detail the building blocks of a video transmission system, and provide specific examples
- Describe “where are things going”, challenges and opportunities

Video Communications System

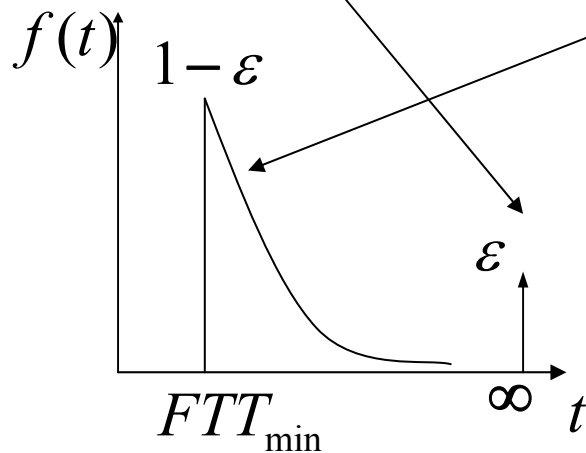


*Channel coding
packetization
modulation*

Channel Models

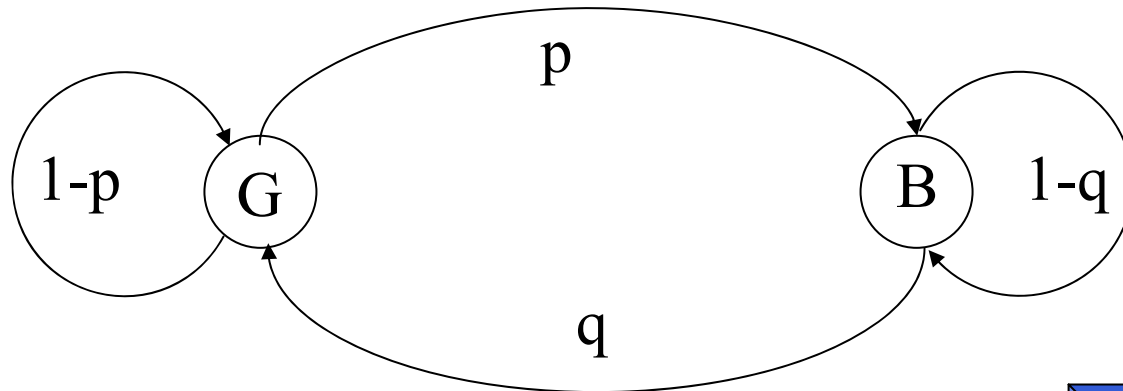
- Network model :
an independent time-invariant packet erasure channel +
random delays

$$\rho^k = \underbrace{\varepsilon}_{\text{erasure}} + (1 - \varepsilon) \underbrace{P\{\Delta T_n(k) > \tau\}}_{\text{delay}}$$



- 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

Gilbert Model

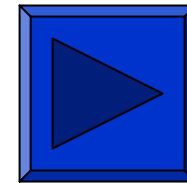


G: Good State, B: Bad State

While in B, error prob= $1-h$

burst length= $1/q * 1000/\text{bitrate}$ msec

Overall BER= $(1-h)*p/(p+q)$



Gilbert Model

BER = 5×10^{-3}

Burst Length = 2.5ms

Bad BER $h = 0.5$



Error Resiliency Challenges

- Non-Robust Nature of (VLBR) Video Coding
 - Highly Predictive
 - Variable Length Codes (requires resynchronization)
- Broad Error Conditions
 - Random Bit Errors
 - Burst Errors
 - Packet Loss Errors
- Low Delay (and additional constraints on resources)
 - Interleaving can be a problem

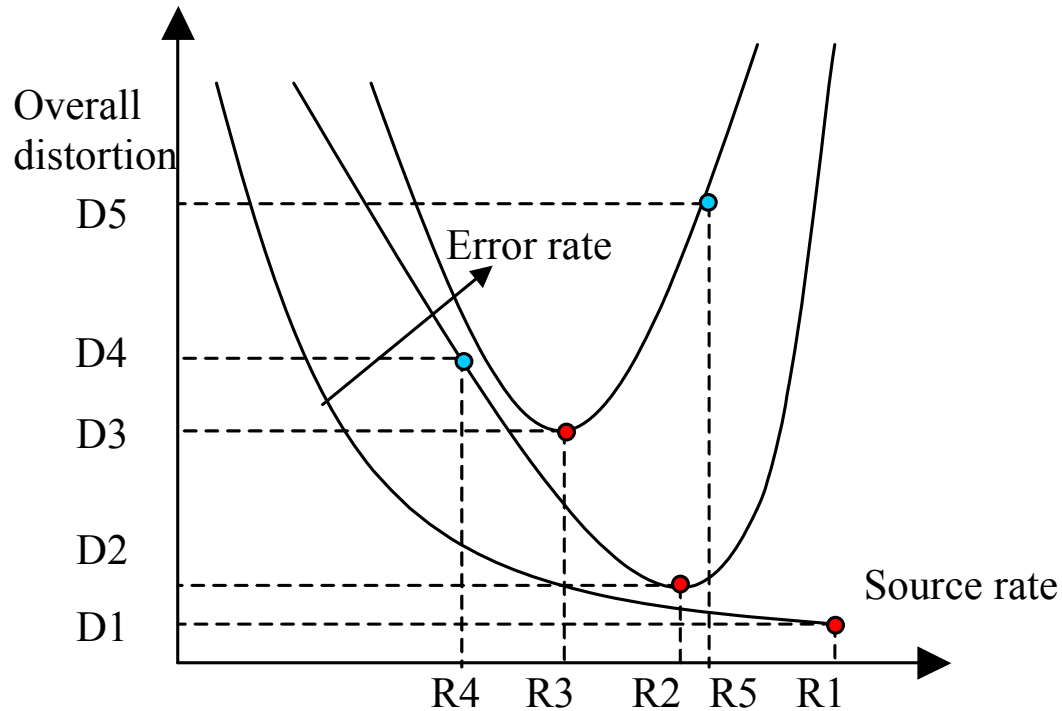


Considerations

Shannon's Separation Theorem

- Joint-Cross Channel Coding
- Network Adaptive Video Transmission
- Dynamic Resource Allocation
- Cross Layer Optimization
- Resource Allocation Based on "Importance of Content"
 - Protection and Energy are given to regions that are difficult to conceal
 - Regions of Interest / Object oriented coding

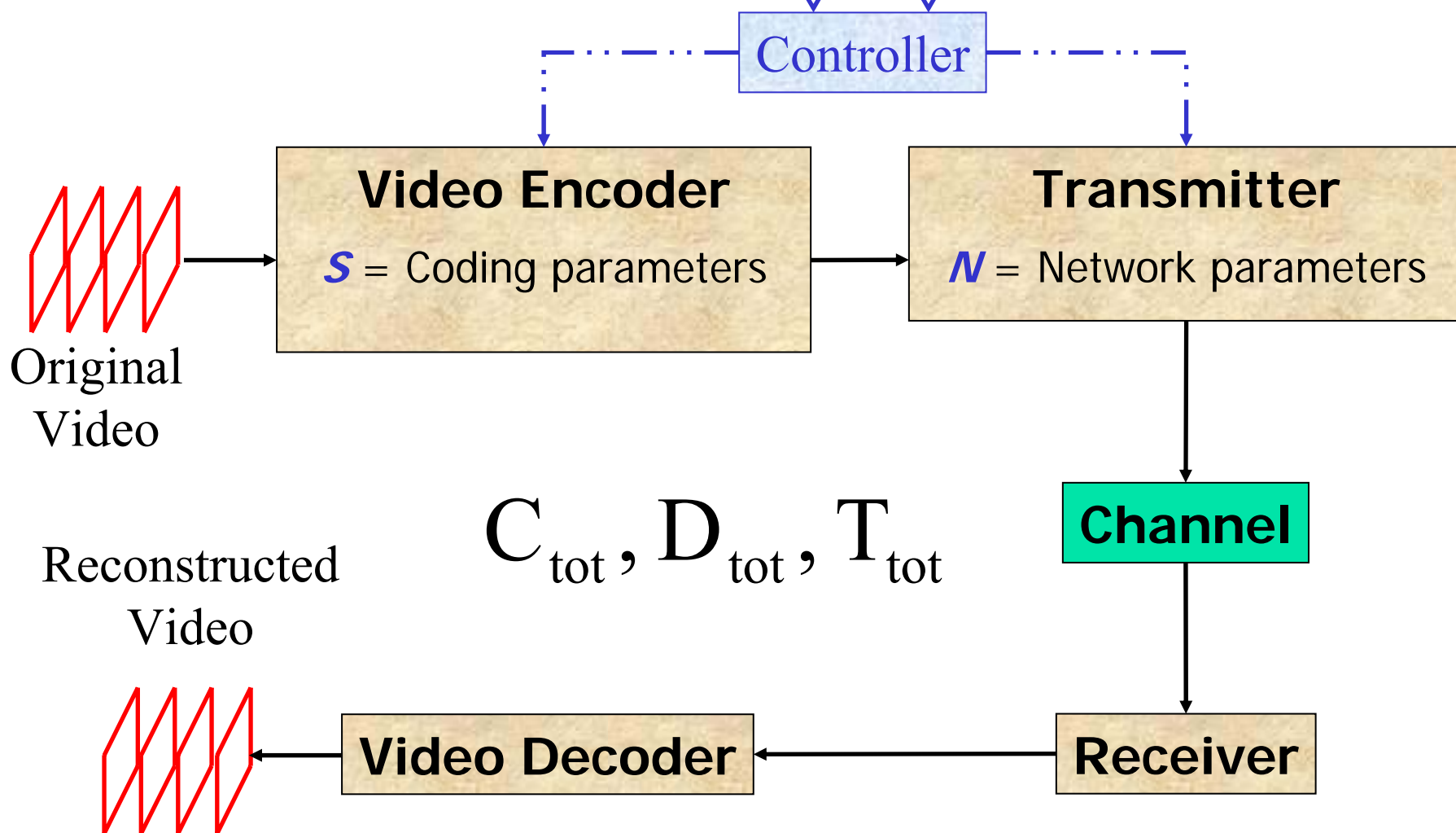
Joint Source-Channel Coding



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

System Model

Concealment Strategy — Channel State Info.



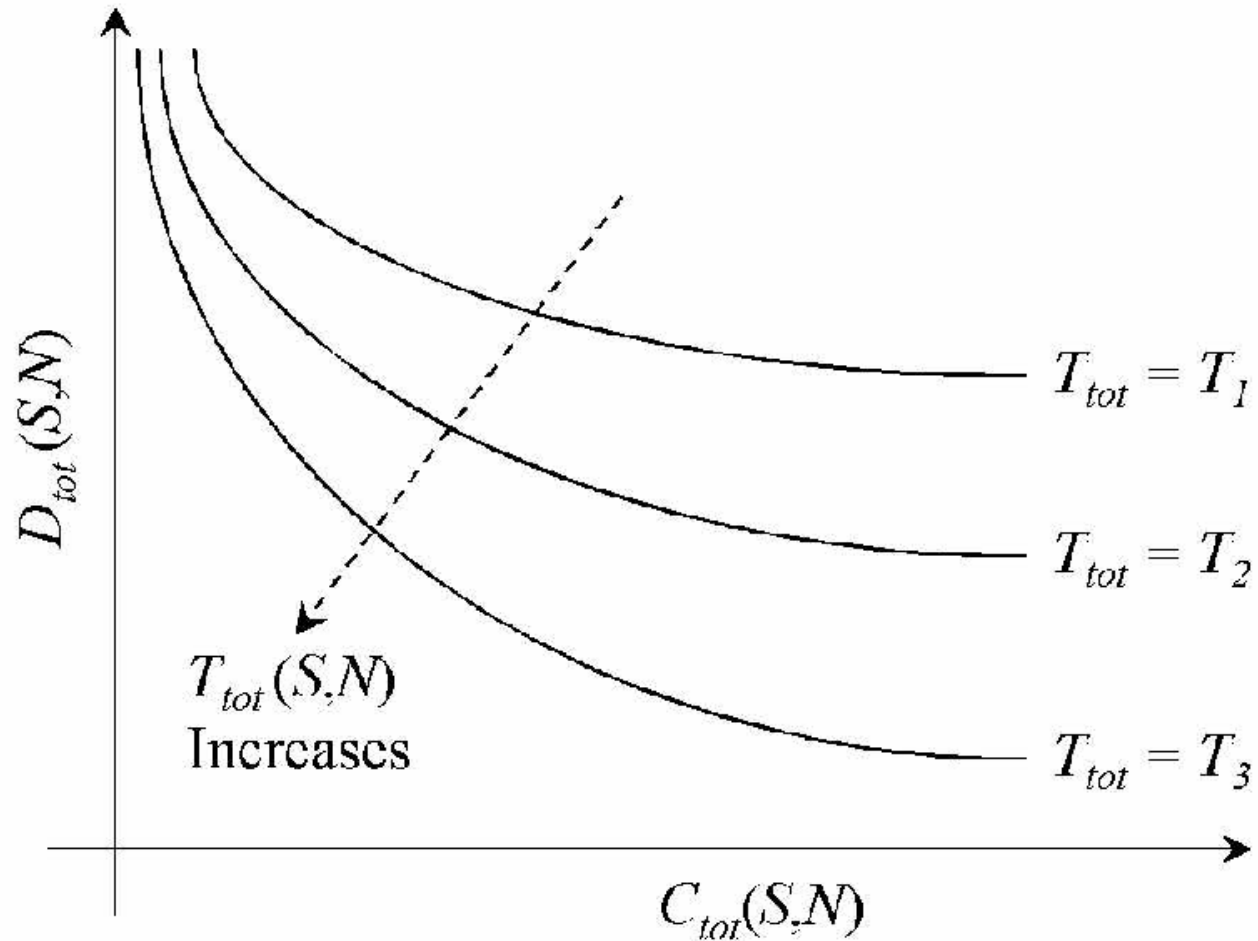


Examples of cost/constraints

- Cost
 - Transmission (computation) power
 - \$\$ for DiffServ
- Network Parameters
 - Scheduling
 - Transmission rate
 - Probability of packet loss
 - FEC
 - ARQ (fast, hybrid)

A. K. Katsaggelos et al, "Advances in Efficient Resource Allocation for Packet-Based Real-Time Video Transmission", *IEEE Proc.*, Jan. 2005.

TRADE-OFFS





Resource-Distortion Optimization Framework

- Goal: Minimize transmission **cost** while limiting the end-to-end **distortion** and **delay**.

$$\min \quad C_{\text{tot}}(S, N) \quad \text{Transmission Cost}$$

s.t.:

$$D_{\text{tot}}(S, N) \leq D_0 \quad \text{End-to-End Distortion Constraint}$$

$$T_{\text{tot}}(S, N) \leq T_0 \quad \text{Transmission Delay Constraint}$$

Expected Distortion

Depends on coding parameters
for the current packet

Depends on concealment scheme
(μ and ρ for other packets)

$$E[D^k] = (1 - \rho^k) E[D_R^k(\mu^k)] + (\rho^k) E[D_L^k]$$

- What Affects the Expected Distortion?
 1. Source coding
 2. Probability of loss in the channel
 3. Error concealment

End-to-End Frame Distortion

Average Expected Distortion $\rightarrow D_{\text{tot}} = \frac{1}{K} \sum_{k=1}^K E[D^k]$

Maximum Expected Distortion $\rightarrow D_{\text{tot}} = \max_{k=[1, \dots, K]} \{ E[D^k] \}$

Vector of Expected Distortions $\rightarrow D_{\text{tot}} = [E[D^1], E[D^2], \dots, E[D^K]]^T$

Expected Distortion for the kth packet

Transmission Cost

- Wireless = Transmission energy

Total Energy $\rightarrow E_{\text{tot}} = \sum_{k=1}^K E^k = \sum_{k=1}^K B^k (\mu^k) \frac{P^k}{R^k}$

Packet index \rightarrow (points to k)
 Number of Bits \rightarrow (points to B^k)
 Source Coding Parameters \rightarrow (points to μ^k)
 Transmission Power \rightarrow (points to P^k)
 Transmission Rate \rightarrow (points to R^k)

The fraction $\frac{P^k}{R^k}$ is circled in green and labeled "Cost per bit".

- DiffServ = Pricing

Total Cost $\rightarrow C_{\text{tot}} = \sum_{k=1}^K C^k = \sum_{k=1}^K B^k (\mu^k) c^k$

The term c^k is circled in green.

Power vs. Probability of Loss

Transmission power
for the kth packet

Probability of loss
for the kth packet

$$P^k = g(\rho^k)$$

- Empirical measurements or analytical models can be used by the transmitter to obtain the function $g(\bullet)$
- Example: (outage probability)
 - Narrowband slowly fading channel with AWGN and i.i.d channel fading per packet (L. Ozarow et al, *VT* '94)

$$P^k = g(\rho^k) = \frac{-C}{\ln(1 - \rho^k)} \quad \text{where} \quad C = \frac{N_0 W}{E[H]} (2^{R/W} - 1)$$

R = channel rate; $E[H]$ = expected channel fade; $N_0 W$ = noise power; W = bandwidth

Minimum Energy Approach

- Goal: Use the minimum energy to achieve an acceptable level of quality and delay.

$$\text{minimize}_{\{\mu^k, P^k\}} E_{\text{tot}} = \sum_{k=1}^K \frac{B^k(\mu^k)}{R} P^k$$

Total Transmission
Energy for frame

s.t.:

$$E[D^k] = \begin{cases} D_0 & \forall k: E[D_R^k(\mu^k)] \leq D_0 \leq E[D_L^k] \\ E[D_L^k] & \forall k: D_0 > E[D_L^k] \end{cases}$$

Maximum Expected
Distortion Constraint
per packet

and

$$T_{\text{tot}} = \sum_{k=1}^K \frac{B^k(\mu^k)}{R} \leq T_0$$

Delay Constraint
for frame

Min Energy Solution

- Coupling Power to Source Coding Parameters
 - For the distortion constraint
 - Assumption: spatially causal concealment strategy

$$\underset{\{\mu^k\}}{\text{minimize}} \quad J_{tot} = \sum_{k=1}^K J^k(\mu^k)$$

$$J^k(\mu^k) = \begin{cases} \frac{B^k(\mu^k)}{R} g \left(\frac{E[D_L^k] - D_0^k}{E[D_L^k] - E[D_R^k(\mu^k)]} \right) + \lambda \frac{B^k(\mu^k)}{R} & \text{if } E[D_R^k] \leq D_0^k \leq E[D_L^k] \\ 0 & \text{if } D_0^k > E[D_L^k] \end{cases}$$

Energy Delay

Generalized Skip

Experimental Results

Compare:

■ Variable Power Approach

- Joint Source Coding and Power Allocation
- $\min E[D(S,P)] ; \text{s.t.}: B_{tot}(S) \leq B_{max} , E_{tot}(S,P) \leq E_{max}$

■ Fixed Power Approach

- Independent Source Coding and Power Allocation
- $\min E[D(S)] ; \text{s.t.}: B_{tot}(S) \leq B_{max}$

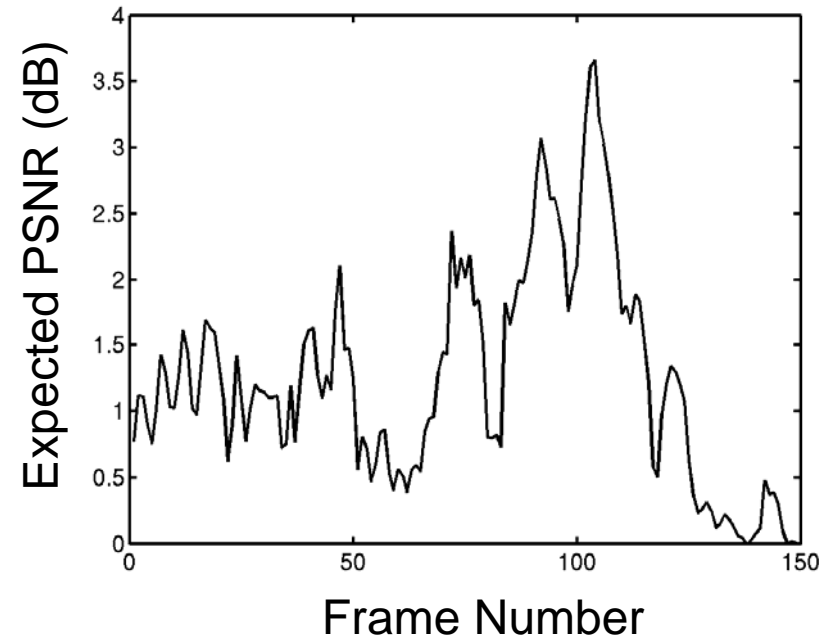
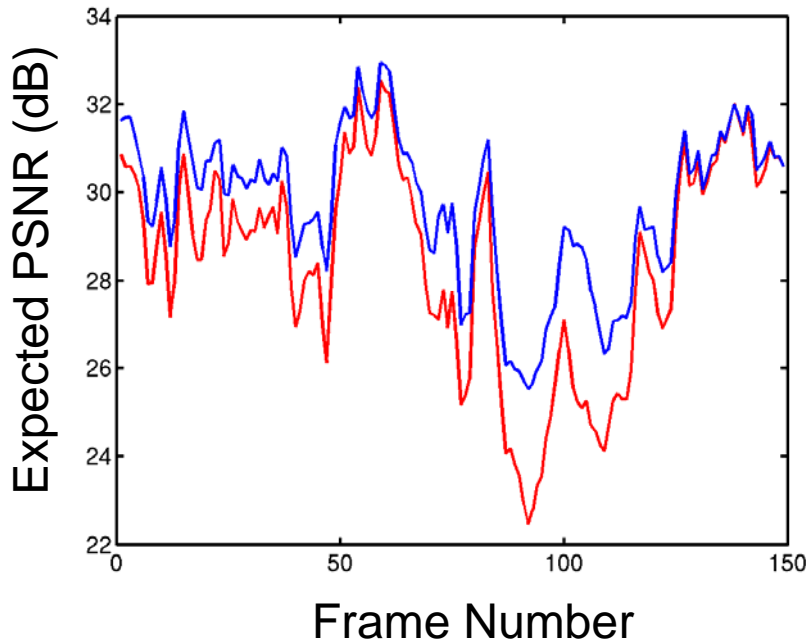
Setup:

- Fixed Delay Constraint ($T_{max} = 33 \text{ ms}$)
- Fixed Transmission Rate ($R = 300 \text{ kbps}$)
- Packetization: One MB per packet
- Concealment: Based on Neighboring MB to the left

Expected Distortion Per Frame

— Variable Power
— Fixed Power

— PSNR Improvement

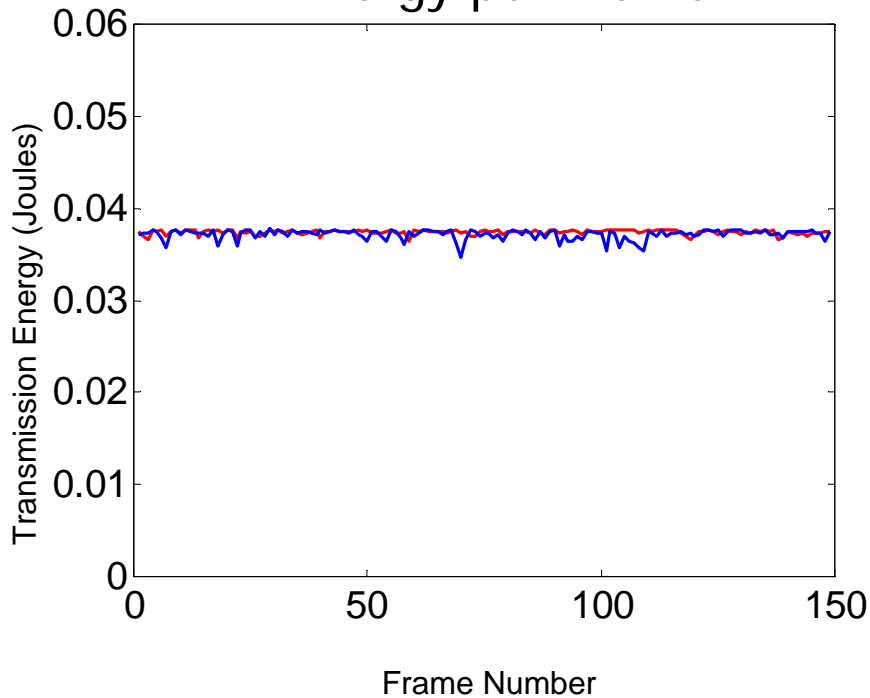


“Foreman” Sequence: transmission rate = 300 kbps ; avg. prob. of error = 0.20

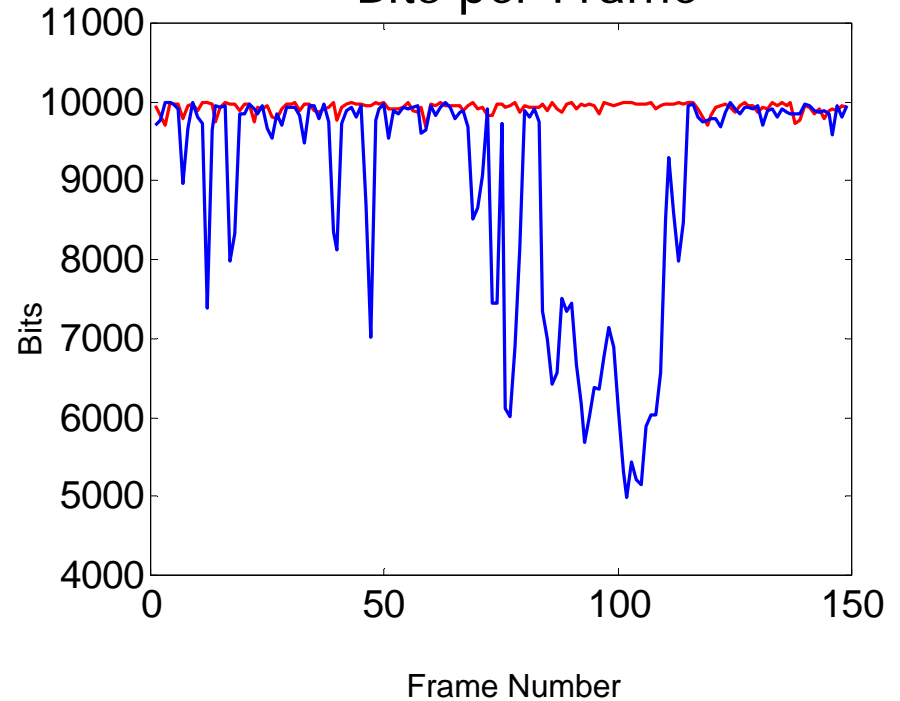
Energy and Bits per Frame

— Variable Power Approach
— Fixed Power Approach

Energy per Frame



Bits per Frame



Visual Comparison: Slow Motion

Fixed Power Approach

Variable Power Approach



- Same energy and delay constraints per frame
- Approaches differ in source coding and power allocation

Visual Comparison: Real-Time

Fixed Power Approach

Variable Power Approach



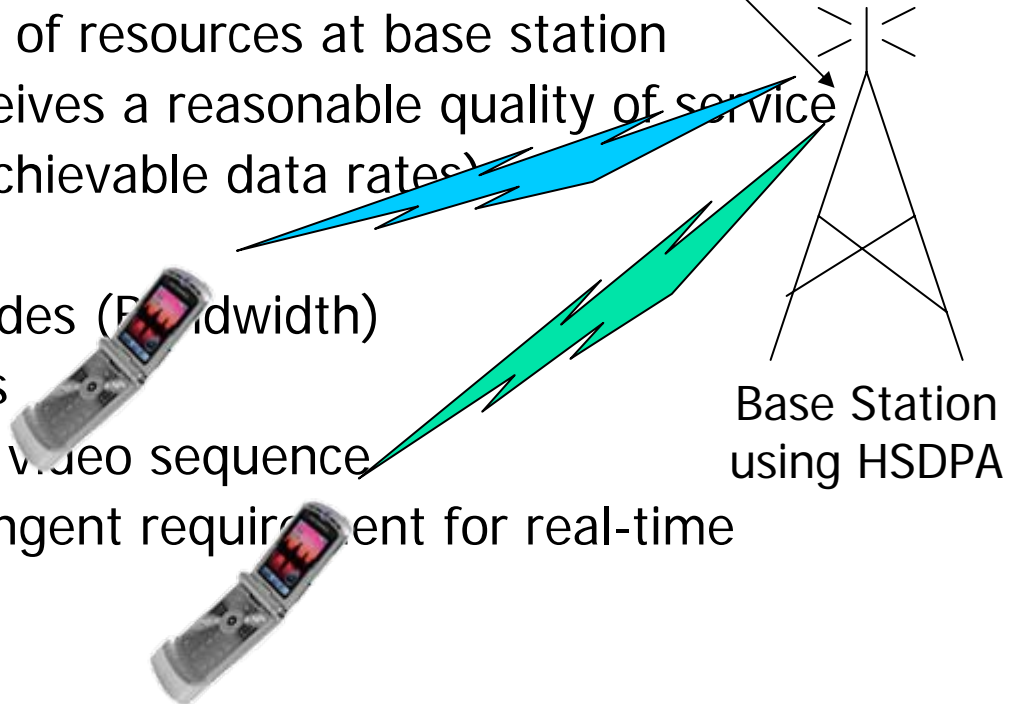
- How do spatio-temporal artifacts affect the *perceived* video quality?

Another Application

- General Problem Definition
 - Transmit multiple pre-encoded video sequences
 - To multiple users
 - Over HSDPA
 - Find optimal distribution of resources at base station
 - Such that each user receives a reasonable quality of service
- Limited Resources (limits achievable data rates)
 - Transmission power
 - Number of spreading codes (Bandwidth)
- Quality of Service Measures
 - End-to-end distortion of video sequence
 - Transmission delay (stringent requirement for real-time applications)



Video Sequences



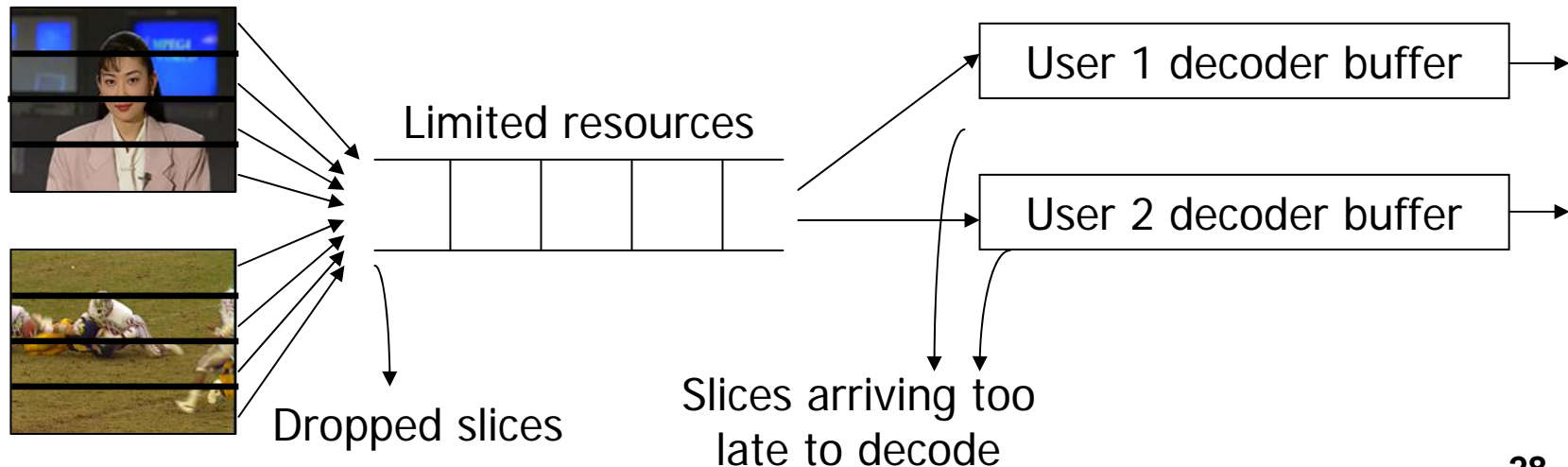


Downlink Packet Scheduling

- “Dumb” Method
 - Round Robin
- State of the Art
 - **Basic Idea:** Allocate resources to users with better channel quality
 - Maximum Throughput Methods
 - Proportionally Fair Methods
 - Fairness criterion based on current average throughput
 - Gradient Based Scheduling
 - Maximize rate to users that will gain the most, subject to channel conditions
- Scheduling for Streaming Video
 - Minimize queue length (delay of head-of-line packet)
 - **Current work does not consider rate-distortion trade-offs for individual video packets**

Video Transmission Assumptions

- Frames are split into **independently decodable slices**
- Video will be viewed in **real-time** (slices from the current frame must be received by the decoder before it finishes decoding the previous frame)
- **Achievable data rates** may not be sufficient to transmit every slice of every frame to all the users within the real-time constraints





Formulation

- Key Idea

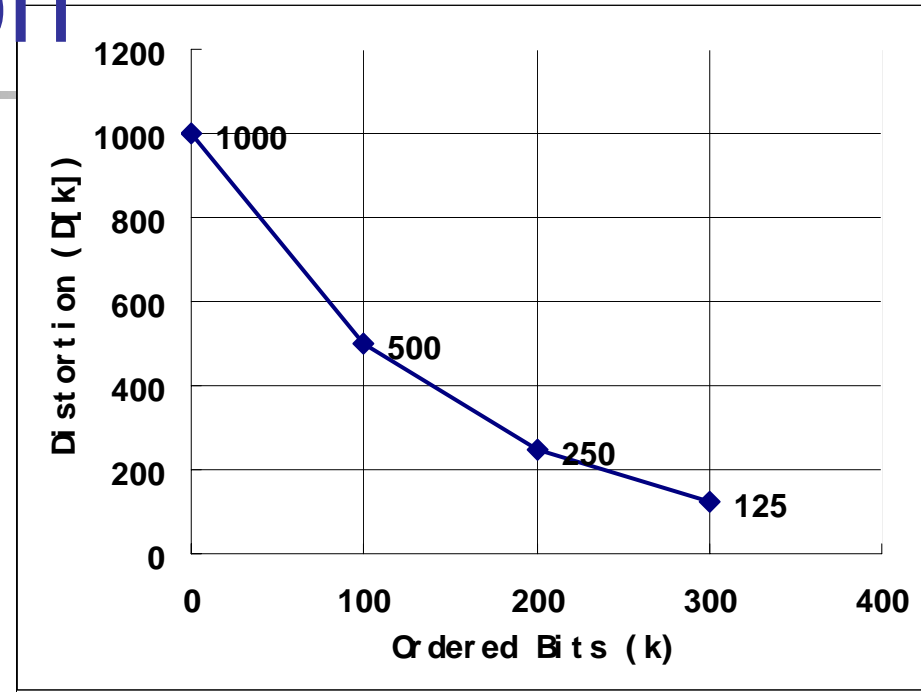
- Order slices by amount of distortion reduction
- Define Slice Utility in terms of distortion reduction

- Assumptions

- Channel states may change by an order of magnitude over one frame's duration
 - Feedback is available every 2ms in HSDPA
 - Optimization performed every time slot (2ms)
- Simple error concealment
 - Copying MB from the same position in previous frame
 - Complex error concealment adds dependencies between slices
 - complicates slice ordering
 - Will eventually include complex error concealment

Utility Function

- Utility Per User Based on Slice Ordering
 - $D[k] :=$ Distortion given k slices are received
 - $D[M] =$ Minimum distortion where, $M =$ total # of slices
- $$U[k] = (D[M] - D[k])$$

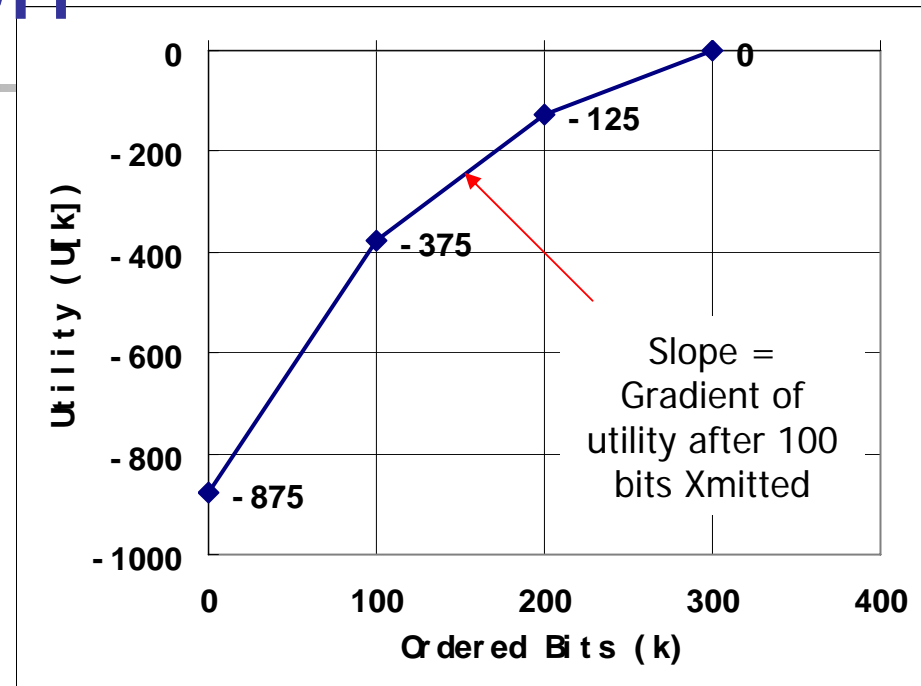


Distortion vs bits after slice ordering

Utility Function

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- $$U[k] = (D[M] - D[k])$$

$$u[k] = U'[k] = \frac{D[k] - D[k+1]}{B_{k+1}}$$



Utility vs bits after slice ordering

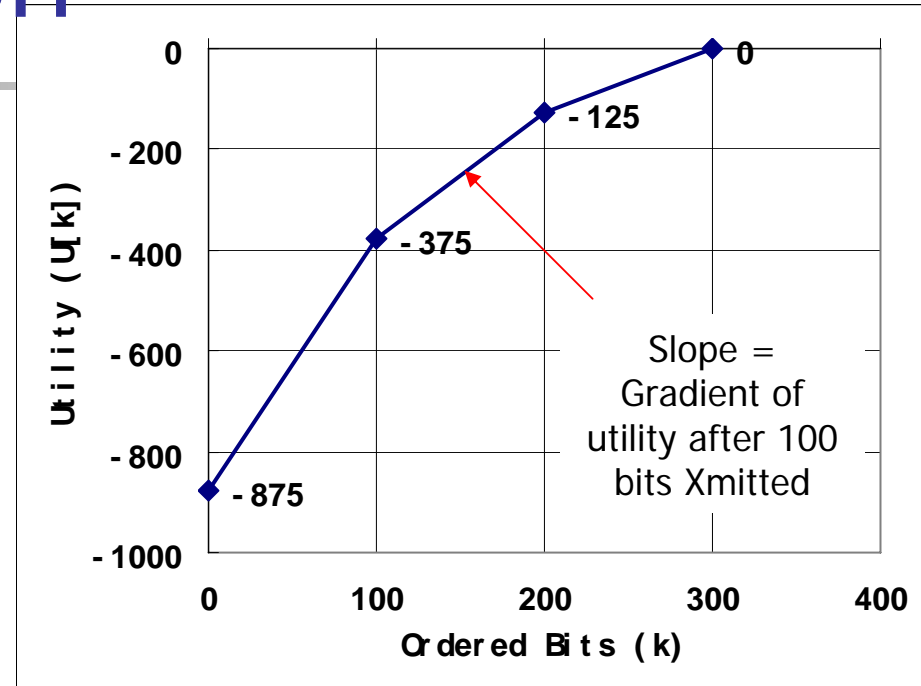
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$$U[k] = (D[M] - D[k])$$

$$u[k] = U'[k] = \frac{D[k] - D[k+1]}{B_{k+1}}$$



Utility vs bits after slice ordering

- Gradient Based Scheduling – Maximizes rates per user weighted by gradient of utility function

$$\max \sum_{i=1}^K w_i u_i \cdot r_i$$

K : # of users

r_i : rate per user

w_i : additional weighting

HSDPA System Constraints

■ K users

■ Maximum transmission power:

$$\sum_{i=1}^K p_i \leq P$$

■ $p_i :=$ power per user

■ Number n_i of codes per user:

$$n_i \leq N_i$$

■ Total number of spreading codes:

$$\sum_{i=1}^K n_i \leq N$$

■ Achievable rates: $r_i = n_i \Gamma(\xi_i \cdot SINR_i)$

■ $SINR_i := \frac{p_i}{n_i} e_i$

■ Where $e_i :=$ channel state (SINR per unit power)

■ $\Gamma :=$ Shannon capacity assuming Gaussian noise channel

■ $\xi_i \in (0, 1] :=$ gap from capacity



Problem Definition

- Maximize sum of user rates weighted by utility gradients:

$$V^* = \max_{\mathbf{n}, \mathbf{p}} \sum_{i=1}^K w_i u_i \cdot n_i B \log \left(1 + \frac{p_i e_i}{n_i} \right)$$

subject to : $\sum_{i=1}^K n_i \leq N$

additional constraint :

$$n_i \leq N_i$$

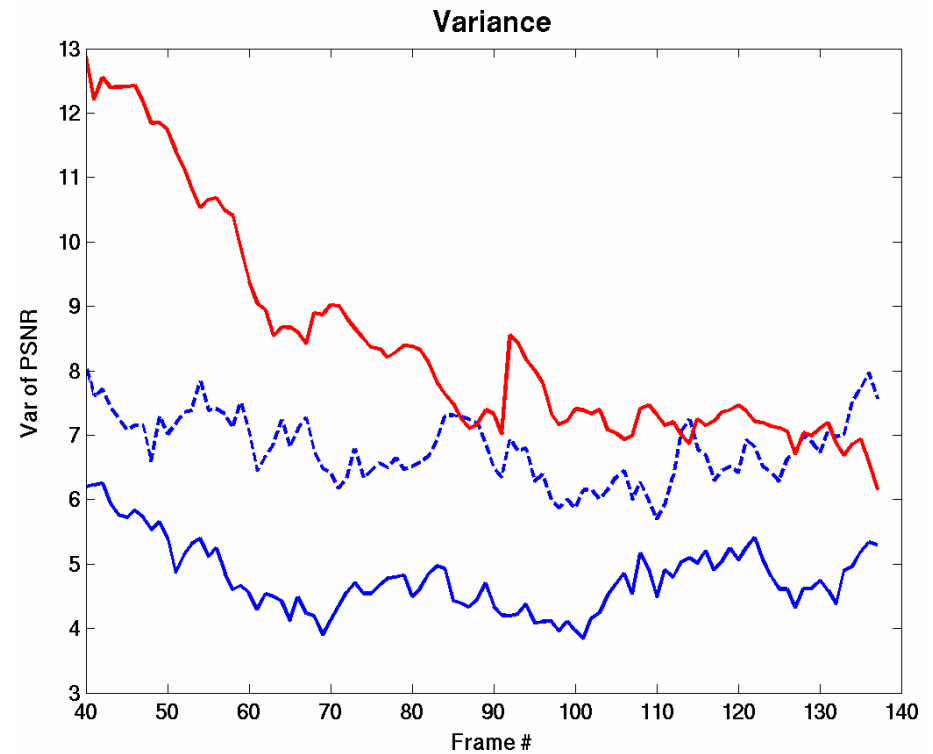
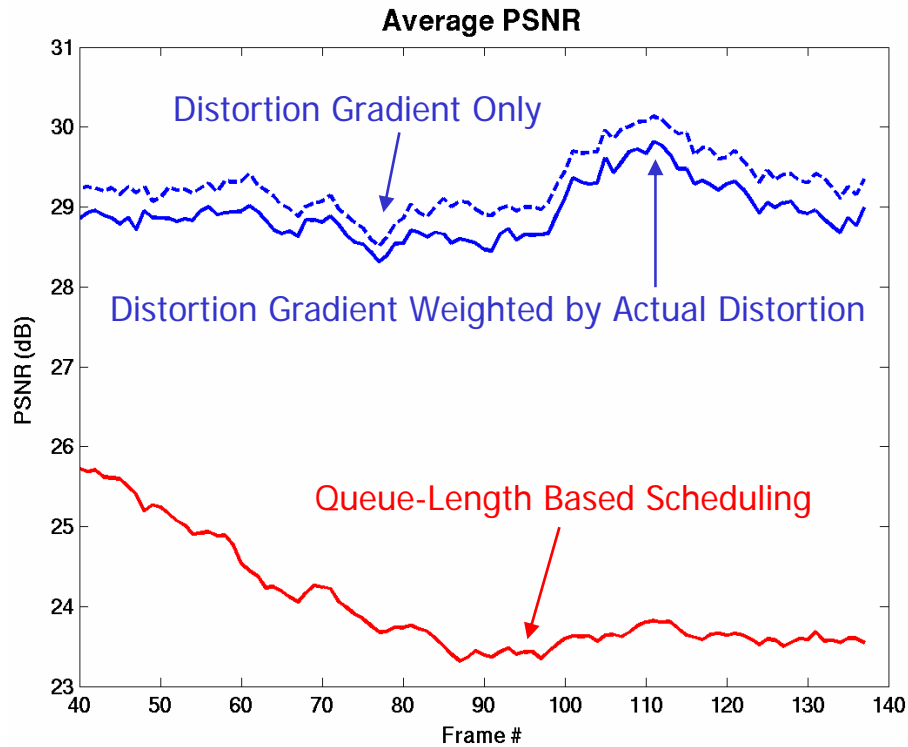
and $\sum_{i=1}^K p_i \leq P$



Simulation I : Similar Fading Stats

- HSDPA System
 - # of Users (K) = 12
 - Total # of spreading codes (N) = 15
 - Max codes per user (N_i) = 5
 - Total Power (P) = 9.9W
 - Max SINR = 1.59
 - Simulated uncorrelated Rayleigh fading channels, with same average channel conditions, for each user
- Video Source
 - H.264 encoded at 256kbps
 - QCIF (176x144) {foreman, mother and daughter, carphone, news, silent, 0...139, 150...289}
 - 11 MBs per slice

Simulation I Results



Demo 1

- 6 users
- $P = 7W$



Weighted Distortion Gradient Metric



Queue-based Utility Metric



Simulation II : One Degraded User

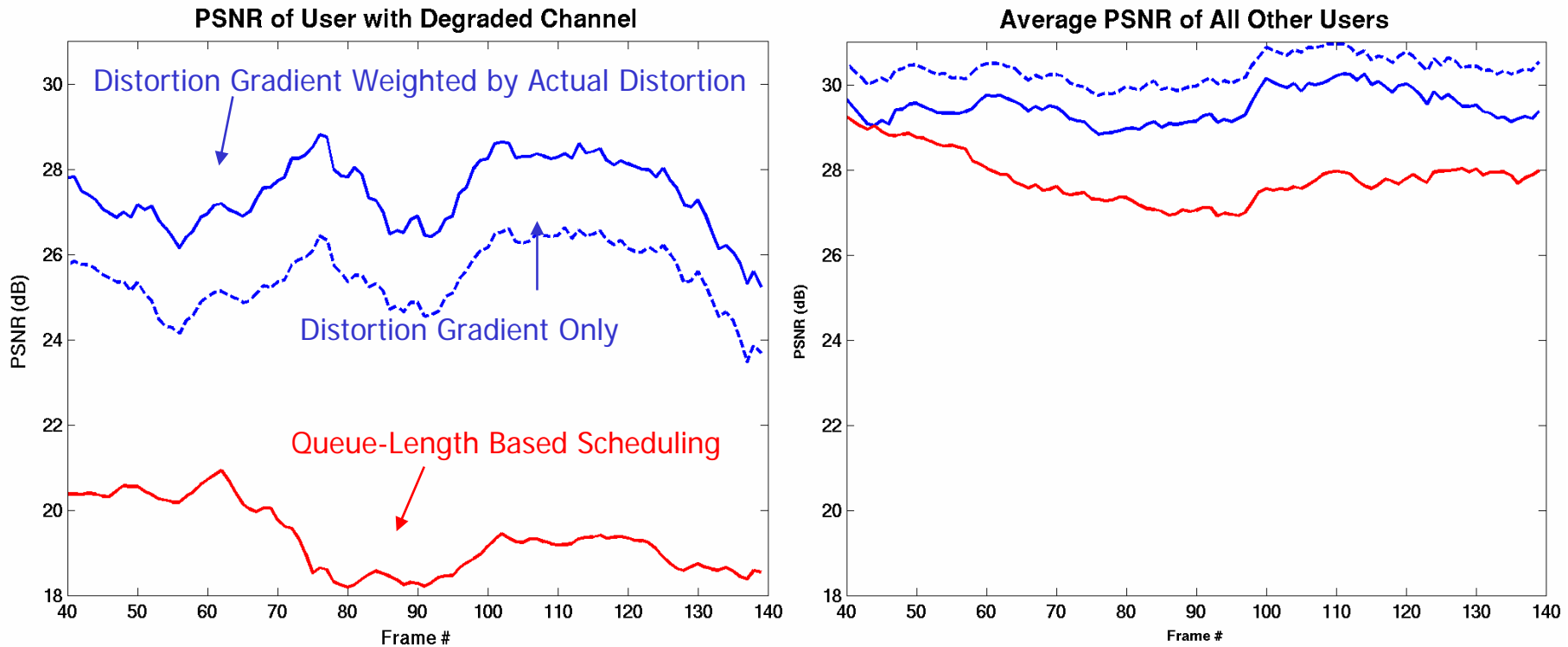
- HSDPA System

- # of Users (K) = 10
- Total # of spreading codes (N) = 15
- Max codes per user (N_i) = 5
- Total Power (P) = 9.9W
- Max SINR = 1.59
- Simulated uncorrelated Rayleigh fading channels, with one user having lower average channel quality

- Video Source

- H.264 encoded at 256kbps
- QCIF (176x144) {foreman, mother and daughter, carphone, news, silent, 0...139, 150...289}
- 11 MBs per slice

Simulation II Results



- The weighted distortion gradient tends to be fair to all users while not significantly sacrificing overall quality

Demo 2

Reception Quality for User with Degraded Channel

Distortion-based Utility



Queue length-based Utility





Closing Remarks

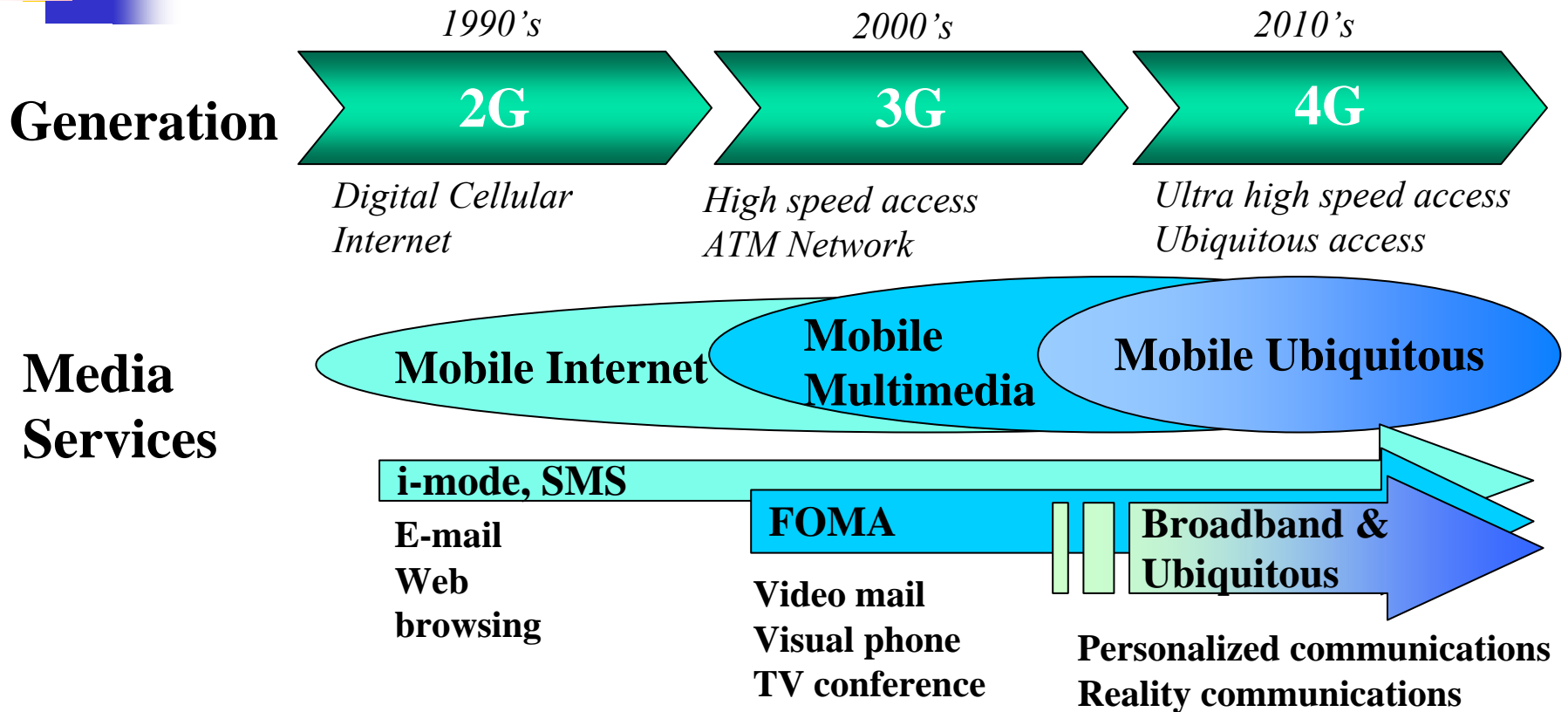
- Rich topic at the intersection of multimedia signal processing, communications, and networking
- Increasing number of applications
- Touched the tip of the iceberg



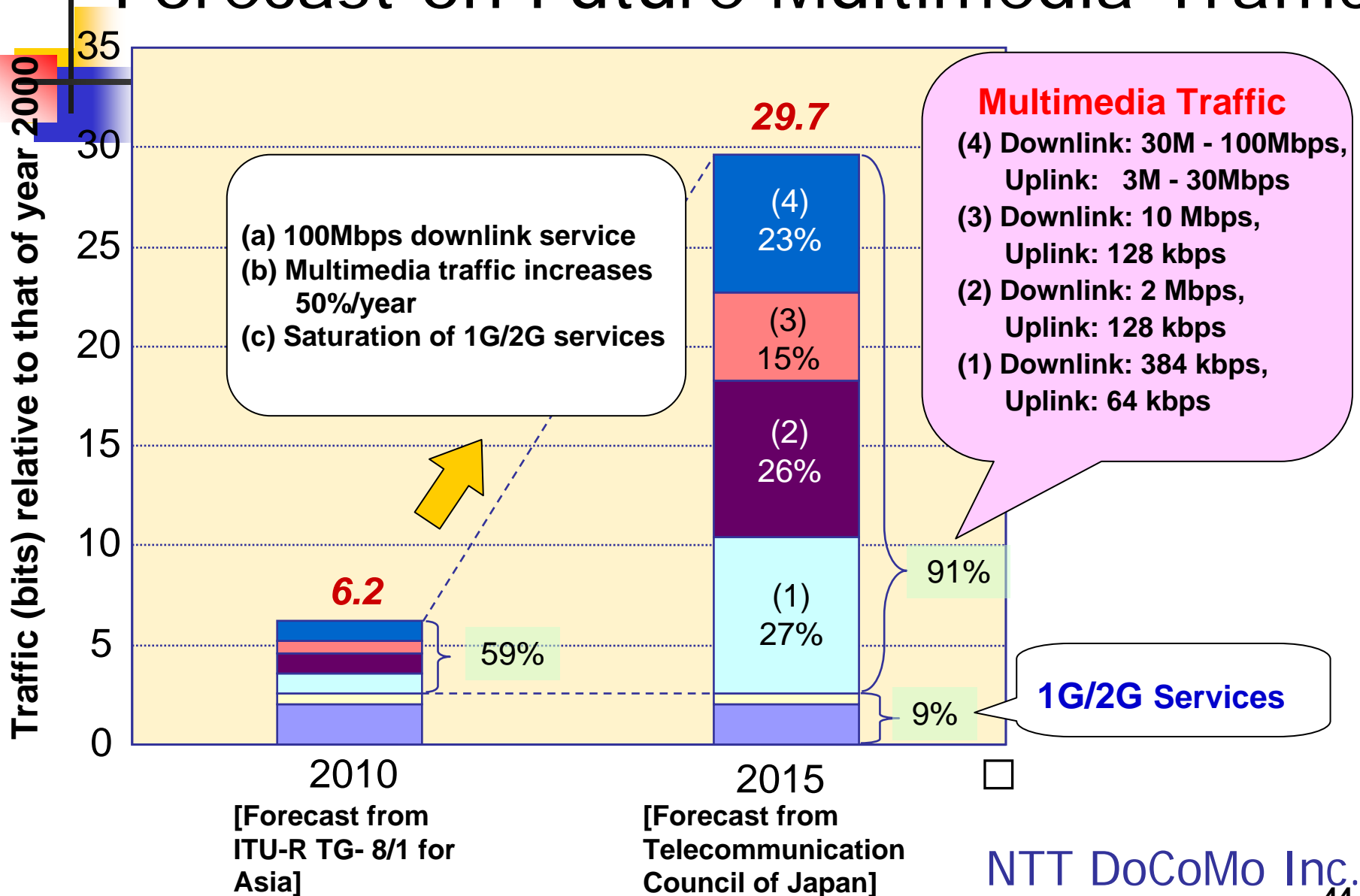
Future Research Directions

- Multi-hop, Image/Video Sensor Networks
- Multi-user, up- and down-link
- Hybrid networks
- More sophisticated concealment methods
- Distortion metrics, utility functions
- Scalable coding, MDC
- Data hiding approaches

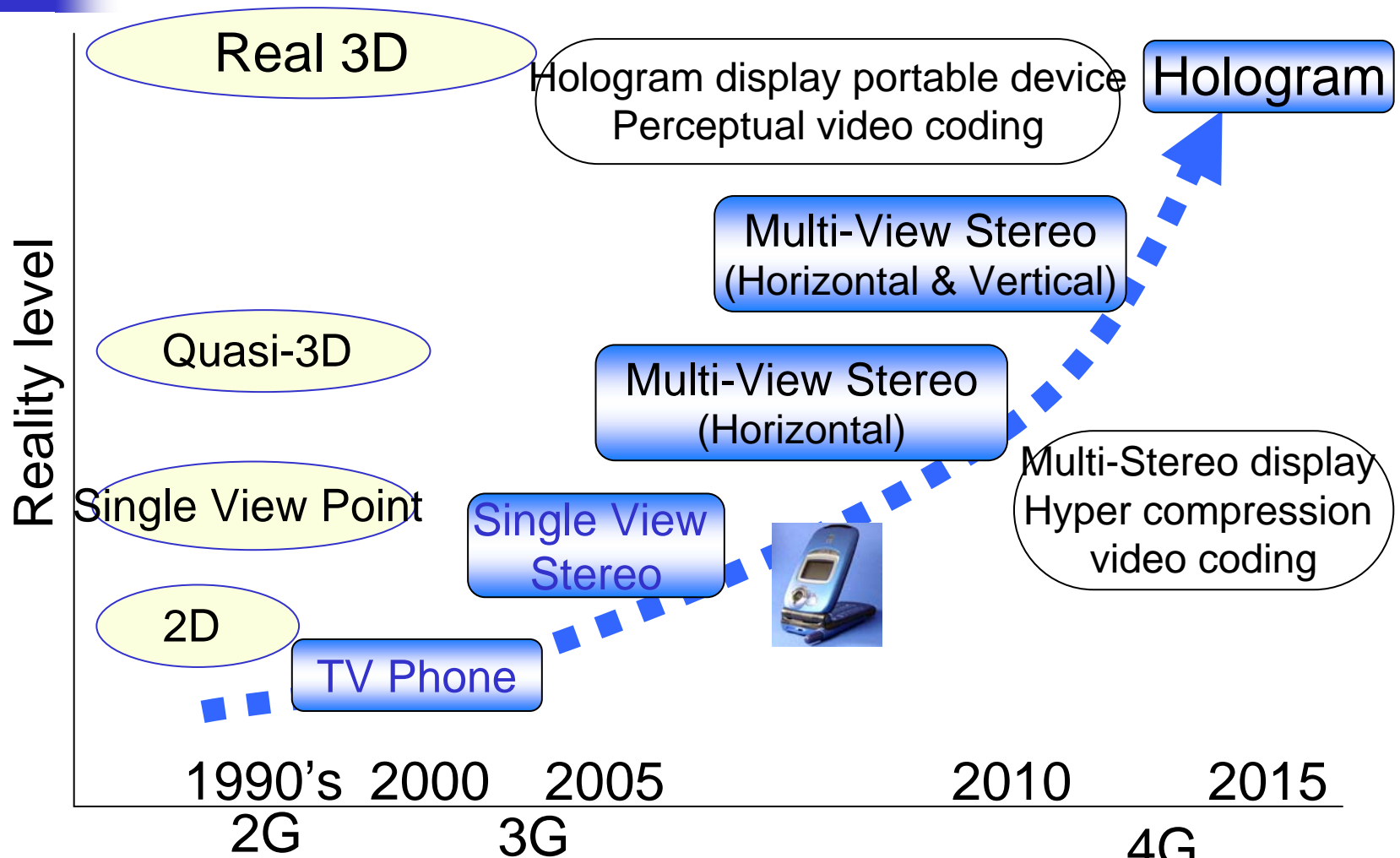
Network Evolution



Forecast on Future Multimedia Traffic



Reality Visual Communications



Requirements for Future Mobile Networks

Media	Transmission speed	Delay	Connection Latency	Terminal capabilities
Speech/ 3D Audio	< 1 Mbps	<50ms	<1sec	<ul style="list-style-type: none"> • 3D sound field control • High efficiency loud speakers
Video/ 3D video	10Mbps (2D video) ~ 30Gbps (3D video)	<50ms	<1sec	<ul style="list-style-type: none"> • Real time hologram
Enhanced Reality	< 1Mbps	<< 50ms Should be predictable	N/A	<ul style="list-style-type: none"> • Eyeglass display • 3D and multimodal UI
Five senses communications	< 1Mbps	<50ms	N/A	<ul style="list-style-type: none"> • Five sense sensors
Tele-existence	<10Mbps (Robotic I/F) < 1Gbps (Virtual avatar) < 100Mbps (Alter-ego existence)	< 10ms < 30ms < 5ms (Small and known jitter)	<1sec	<ul style="list-style-type: none"> • Alter-ego robot



Additional Work

- Cross Layer Design
- Resource-Distortion Optimization Framework (A. K. Katsaggelos et al, "Advances in Efficient Resource Allocation for Packet-Based Real-Time Video Transmission", *IEEE Proc.*, Jan. 2005)
- JSCC (L. Kondi, F. Ishtiaq, and A. K. Katsaggelos, "Joint Source-Channel Coding for Motion-Compensated DCT-Based Scalable Video," *IEEE Trans. IP*, vol. 11, pp. 1043-1052, Sept. 2002)
- Energy Efficient Wireless Video Streaming (Y. Eisenberg, C. Luna, T. Pappas, R. Berry, A. K. Katsaggelos, "Joint Source Coding and Transmission Power Management for Energy Efficient Wireless Video Communications", *IEEE CASVT*, pp. 411-424, June 2002)
- Data Rate Adaptation and Scheduling (C. Luna, Y. Eisenberg, R. Berry, T. Pappas, A. K. Katsaggelos, "Joint Source Coding and Data Rate Adaptation for Energy Efficient Wireless Video Streaming", *IEEE JSAC*, pp. 1710-1720, Dec. 2003)
- DiffServ (F. Zhai, C. Luna, Y. Eisenberg, R. Berry, T. Pappas, A. K. Katsaggelos, "Joint Source Coding and Packet Classification for Real Time Video Transmission over DiffServ Networks", *IEEE Trans. Multimedia*, to appear).
- Variance-Aware Distortion Estimation (Y. Eisenberg, F. Zhai, C. Luna, T. Pappas, R. Berry, A. K. Katsaggelos, "Variance-Aware Distortion Estimation for Wireless Video Communications", *IEEE Image Processing*, to appear)

Additional Work (cont'ed)

- Hybrid (FEC, retransmission) Rate Control (F. Zhai, Y. Eisenberg, T. Pappas, R. Berry, A. K. Katsaggelos, "Rate Distortion Optimized Hybrid Error Control for Real-Time Packetized Video Transmission", *IEEE Trans. Image Processing*, to appear).
- Hybrid Intra/Inter FEC
- Hybrid (wired/wireless) Networks
- Object-based Encoded Video Transmission (*CSVT 2006*)
- FGS Scalable Video Transmission (*ICC '04*)
- Correlated Fading, Channel Mismatch Sensitivity (*ICIP'04*)