

Analysis and Modeling of MPEG-4 and H.264 Multi-Layer Video Traffic

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Outline

- 1 Background
 - Motivation
 - Preliminaries
 - Challenges & Current Status
- 2 Our Work
 - Modeling single-layer video traffic
 - Modeling multi-layer video traffic
 - Model accuracy study

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Importance of Traffic Modeling

Importance

- Properly allocate network resources
- Evaluate protocols and effectively design networks
- Use as traffic descriptor to achieve certain Quality of Service (QoS) requirements
- Analyze and characterize a queue or a network

Goals of Traffic Modeling

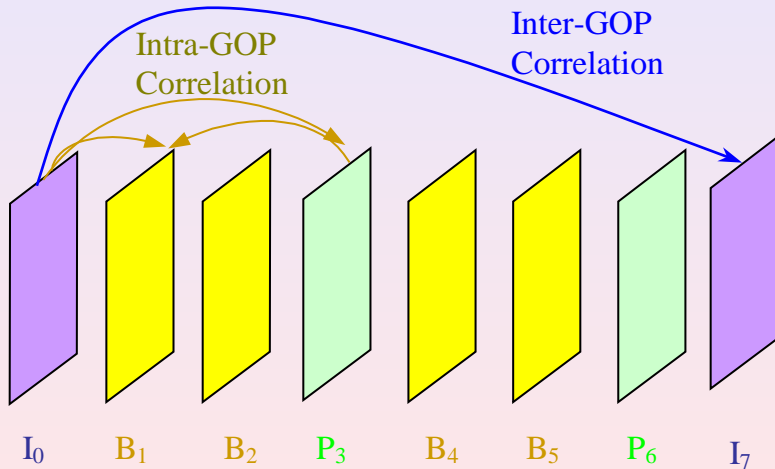
Goals

- Capture the characteristics of video frame size sequences
 - The marginal probability density function (PDF) of frame sizes
 - The autocorrelation function (ACF) of video traffic
- Accurately predict network performance
 - Buffer overflow probabilities
 - Temporal burstiness

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Group of Pictures



Inter- and Intra-GOP Correlation

Definition

- 1 Inter-GOP correlation is the correlation among various GOPs, which is well characterized by the ACF of the I-frames.
- 2 Intra-GOP correlation is the correlation between P/B-frames and the I-frame in the same GOP.

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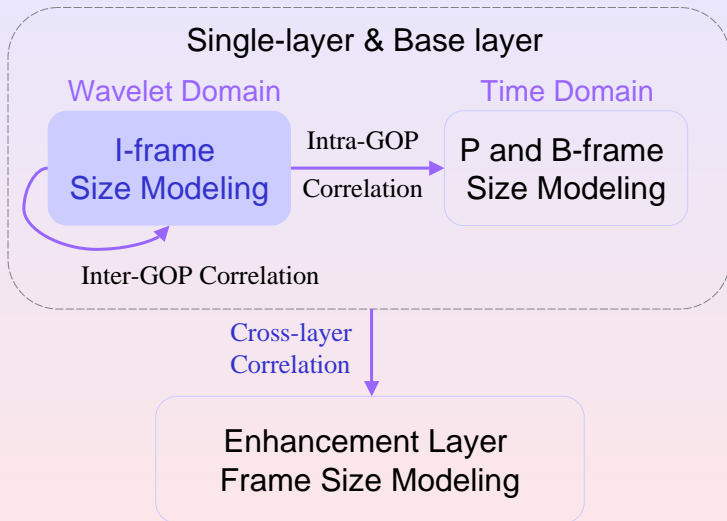
Challenges & Current Status

Challenges

- Coexistence of long range dependency (LRD) and short range dependency (SRD)
- Coexistence of inter- and intra-GOP correlation
- Strong cross-layer correlation in multi-layer video traffic
- Various PDF among I, P, and B-frame sizes distributions

Current Status

- Difficult to capture both LRD and SRD properties
- Little work has considered both inter- and intra-GOP correlation
- Most existing models only apply to single-layer video traffic
- Current multi-layer traffic models do not capture the cross-layer correlation



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Wavelet Decomposition

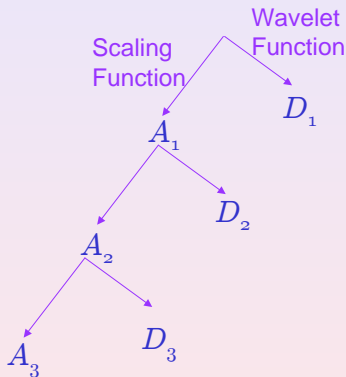


Figure: A typical wavelet decomposition

Wavelet Decomposition

- Wavelet decomposition can be simply considered as passing the original signal to a high-pass filter (wavelet function) and a low-pass filter (scaling function).
- Wavelet function generates the detailed coefficients $\{D_j\}$ and scaling function generates the approximation coefficients $\{A_j\}$, where j is the decomposition level.

Definition of Frame Sizes

Definition

Assuming that n is the GOP number,

- 1 $\phi^I(n)$ is the I-frame size of the n -th GOP.
- 2 $\phi_i^P(n)$ is the size of the i -th P-frame in the n -th GOP.
- 3 $\phi_i^B(n)$ is the size of the i -th B-frame in the n -th GOP.

For example, $\phi_3^P(10)$ represents the size of the third P-frame in the 10-th GOP.

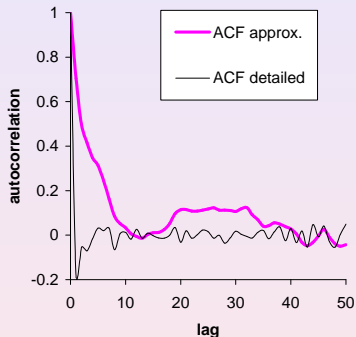
Modeling I-Frame Sizes

Algorithm

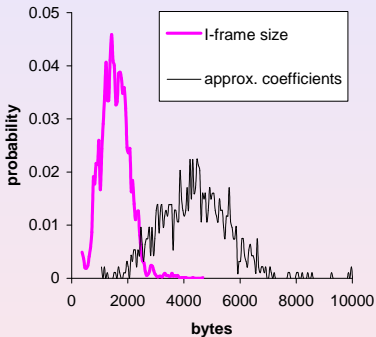
- 1 Perform wavelet decomposition to I-frame sizes $\phi^I(n)$ till decomposition level J
- 2 Estimate the coarsest approximation coefficients $\{A_J\}$
- 3 Estimate the detailed coefficients at each level
- 4 Perform inverse wavelet transform to obtain the synthetic I-frame sizes

Modeling I-Frame Sizes

Analysis of Wavelet Coefficients



(a)



(b)

Figure: (a) The ACF structure of coefficients $\{A_3\}$ and $\{D_3\}$ in single-layer Star Wars IV. (b) The histogram of I-frame sizes and that of approximation coefficients $\{A_3\}$ in the same sequence.

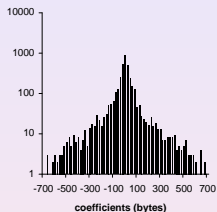
Modeling I-Frame Sizes

Estimation of Wavelet Coefficients

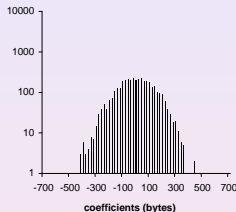
- Estimate the coarsest approximation coefficients $\{A_J\}$:
 - Prior work uses independent random Gaussian or Beta variables
 - Our model uses **dependent** random variables with marginal Gamma distribution
- Estimate detailed coefficients $\{D_j\}$ at each level:
 - Prior work uses i.i.d. Gaussian random variables
 - Our model uses i.i.d. **mixture Laplacian** random variables

Modeling I-Frame Sizes

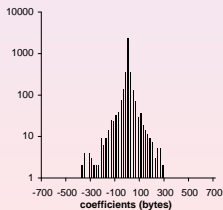
Detailed Coefficients Estimates



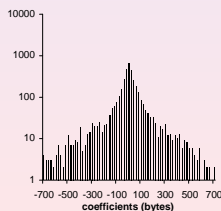
(a) Actual



(b) Gaussian



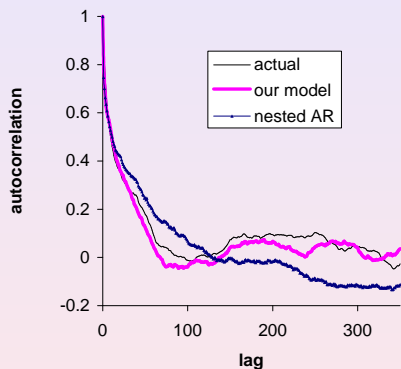
(c) GGD



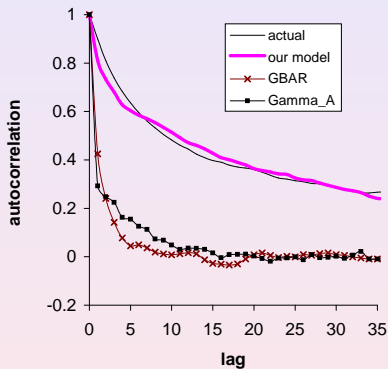
(d) Mix-Laplacian

Modeling I-Frame Sizes

Performance Comparison

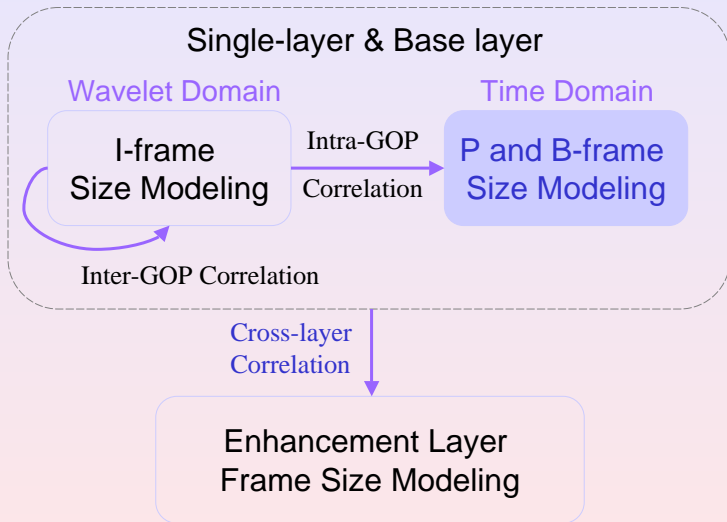


(a) LRD



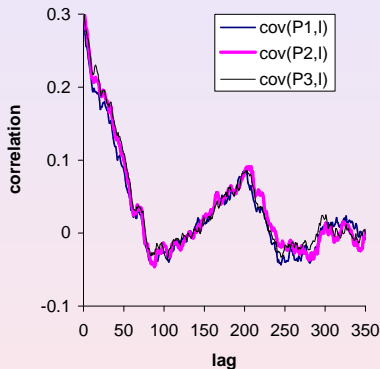
(b) SRD

Figure: The ACF of the actual I-frame sizes and that of the synthetic traffic in (a) long range and (b) short range.

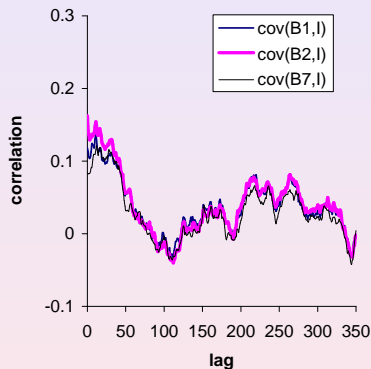


Modeling P/B-Frame Sizes

Intra-GOP Correlation



(a)



(b)

Figure: (a) The correlation between $\{\phi^I(n)\}$ and $\{\phi_i^P(n)\}$, $i = 1, 2, 3$, and (b) that between $\{\phi^I(n)\}$ and $\{\phi_i^B(n)\}$, $i = 1, 2, 7$ in Star Wars.

Modeling P/B-Frame Sizes

Model Comparison

Previous Work

Previous work does not consider the intra-GOP correlation and estimates P/B-frame sizes as **i.i.d. random variables**.

Our Model

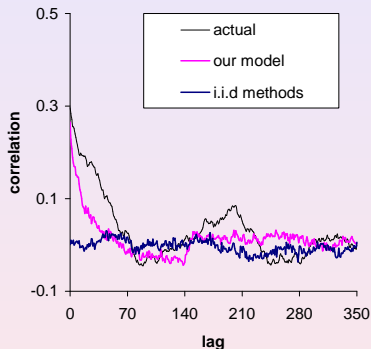
The size of the i -th P-frame in the n -th GOP is:

$$\phi_i^P(n) = a\tilde{\phi}^I(n) + \tilde{v}(n), \quad \text{where} \quad a = \frac{r(0)\sigma_P}{\sigma_I}.$$

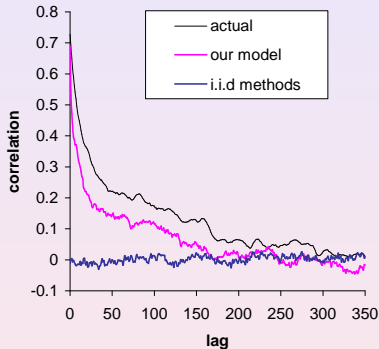
- Process $\tilde{\phi}^I(n) = \phi^I(n) - E[\phi^I(n)]$ and process $\tilde{v}(n)$ is independent of $\tilde{\phi}^I(n)$.
- Parameter σ_P is the standard deviation of $\{\phi_i^P(n)\}$, σ_I is the standard deviation of $\{\phi^I(n)\}$.

Modeling P/B-Frame Sizes

Performance Comparison



(a)



(b)

Figure: (a) The correlation between $\{\phi_1^P(n)\}$ and $\{\phi^I(n)\}$ in Star Wars and (b) that between $\{\phi_1^B(n)\}$ and $\{\phi^I(n)\}$ in Jurassic Park.

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Brief Description

Layered Video Coding

- Generates one base layer (BL) and one or more enhancement layers(EL)
- BL provides a low but guaranteed level of quality and EL provides quality improvement
- The input to an EL is the residual between the original image and the reconstructed image from the BL.

Cross-Layer Correlation

The enhancement layer has a strong dependency on the base layer, which is referred to as **cross-layer correlation**.

Brief Description (cont.)

Definition

Assuming that $n \geq 1$ represents the GOP number in an enhancement layer,

- 1 $\varepsilon^I(n)$ is the I-frame size in this GOP.
- 2 $\varepsilon_i^P(n)$ is the size of the i -th P-frame in this GOP.
- 3 $\varepsilon_i^B(n)$ is the size of the i -th B-frame in this GOP.

Modeling Enhancement Layer

Algorithm

- 1 We estimate I-frame sizes in wavelet domain
- 2 Estimate P and B-frame sizes using the cross-layer correlation:

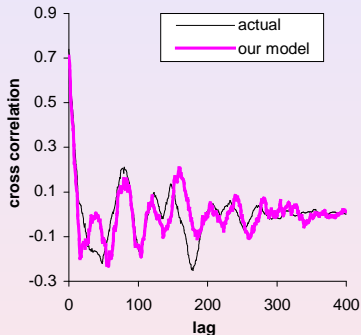
$$\begin{aligned}\varepsilon_i^P(n) &= a\phi_i^P(n) + \tilde{w}_1(n), \\ \varepsilon_i^B(n) &= a\phi_i^B(n) + \tilde{w}_2(n),\end{aligned}$$

where $a = r(0)\sigma_\varepsilon/\sigma_\phi$.

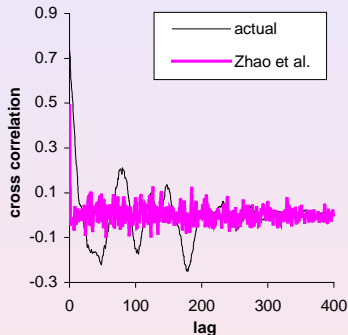
- Parameter $r(0)$ is the lag-0 cross-correlation coefficient, σ_ε and σ_ϕ are the standard deviation of the EL and the corresponding BL.
- Processes $\{\tilde{w}_1(n)\}, \{\tilde{w}_2(n)\}$ are independent of $\{\phi_i^P(n)\}$ and $\{\phi_i^B(n)\}$.

Modeling Enhancement Layer

Cross-correlation Comparison



(a)



(b)

Figure: The cross-correlation between $\{\varepsilon^I(n)\}$ and $\{\phi^I(n)\}$ in The Silence of the Lambs and that in the synthetic traffic generated from (a) our model and (b) a popular model in related work.

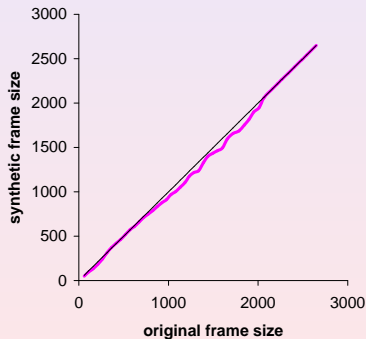
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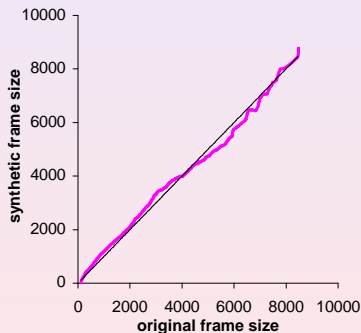
Performance Evaluation Methods

QQ Plots

- To verify the distribution similarity between the original traffic and the synthetic one.



(a) Base-layer

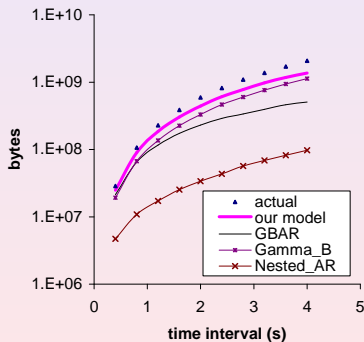


(b) Enhancement layer

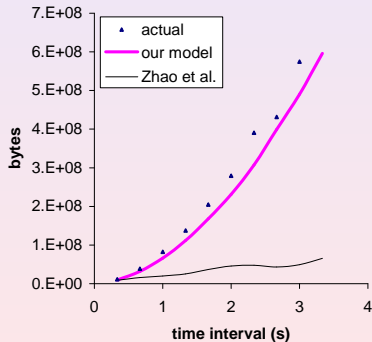
Performance Evaluation Methods (cont.)

Variance of traffic during various time intervals

- To check whether the second-order moment of the synthetic traffic fits that of the original one.



(a) Star Wars BL



(b) Star Wars EL

Performance Evaluation Methods (cont.)

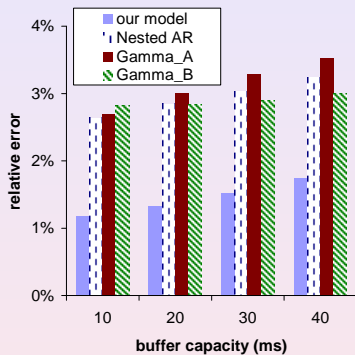
Leaky-bucket simulation

- To examine how well the traffic model preserves the temporal information of the original traffic
- Implementation: Pass the original and synthetic traffic through a generic buffer with capacity c and drain rate d .
- Evaluation metric:

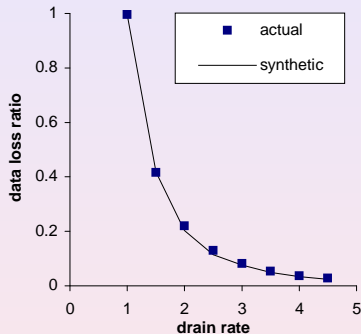
$$e = \frac{|p - p_{model}|}{p},$$

where p is the actual data loss ratio and p_{model} is the synthetic one.

Performance Evaluation Methods (cont.)



(a)



(b)

Figure: (a) Comparison of several models in H.264 coded Starship Troopers. (b) The loss ratio p of the original and synthetic enhancement layer traffic for The Silence of the Lambs.

Conclusion

- This paper proposed a traffic model applicable to **both single-layer** and **multi-layer** VBR video traffic.
- The presented traffic modeling framework captures **both LRD and SRD** properties of video traffic.
- This framework accurately describes both **inter-/intra-GOP** correlation and the **cross-layer** correlation.

- Future Work
 - Develop a unified model for multi-layer video traffic
 - More applications in overlay networks

Thank you!

Any questions?