

Temporal Resolution Enhancement in Compressed Video

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Motivation

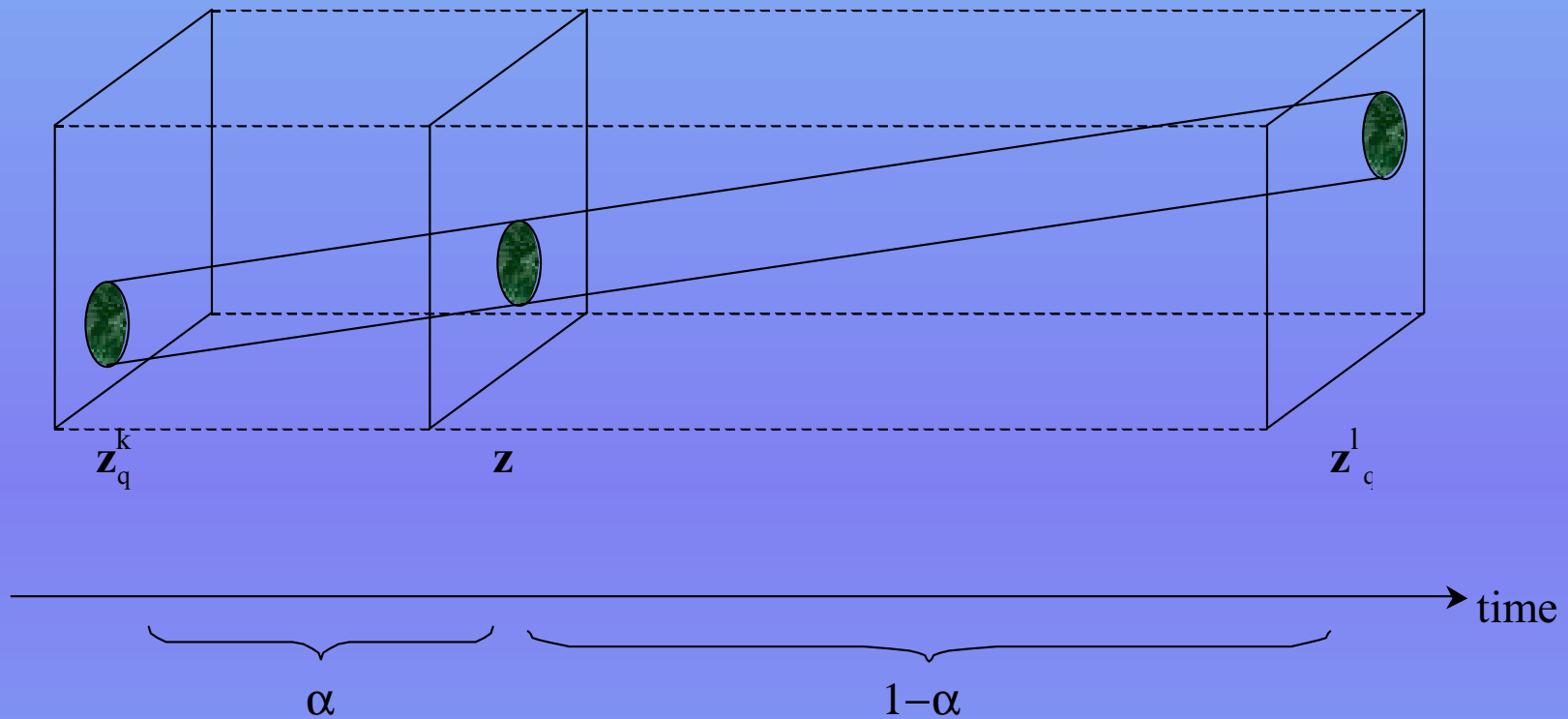
1. Sometimes, we may want to increase video frame rate to avoid jerky, uneven motion:
 - (a) Frames can be “dropped” during encoding
 - (b) Might want a “slow-motion” effect
2. Frame rate is increased by inserting frames between received frames.
3. Want to prevent propagation of compression artifacts from received frames to inserted frames.



Notation:

- \mathbf{z}_q^k Decoded frame at time k (with quantization noise)
- \mathbf{z}_q^l Decoded frame at time l (with quantization noise)
- \mathbf{z} Frame to insert between \mathbf{z}_q^k and \mathbf{z}_q^l





Distance between z_q^k and z_q^l is normalized to one

Assume a MAP criterion for estimation:

$$\begin{aligned}\hat{\mathbf{z}} &= \arg \max_{\mathbf{z}} p(\mathbf{z} | \mathbf{z}_q^k, \mathbf{z}_q^l) \\ &= \arg \max_{\mathbf{z}} p(\mathbf{z}) p(\mathbf{z}_q^k, \mathbf{z}_q^l | \mathbf{z})\end{aligned}$$

$p(\mathbf{z}_q^k, \mathbf{z}_q^l | \mathbf{z})$ relates observations to image we are estimating

$p(\mathbf{z})$ is a prior image model that encourages smoothness



Need to relate observations \mathbf{z}_q^k and \mathbf{z}_q^l with frame to insert \mathbf{z} .

\implies start with original images (include compression noise later):

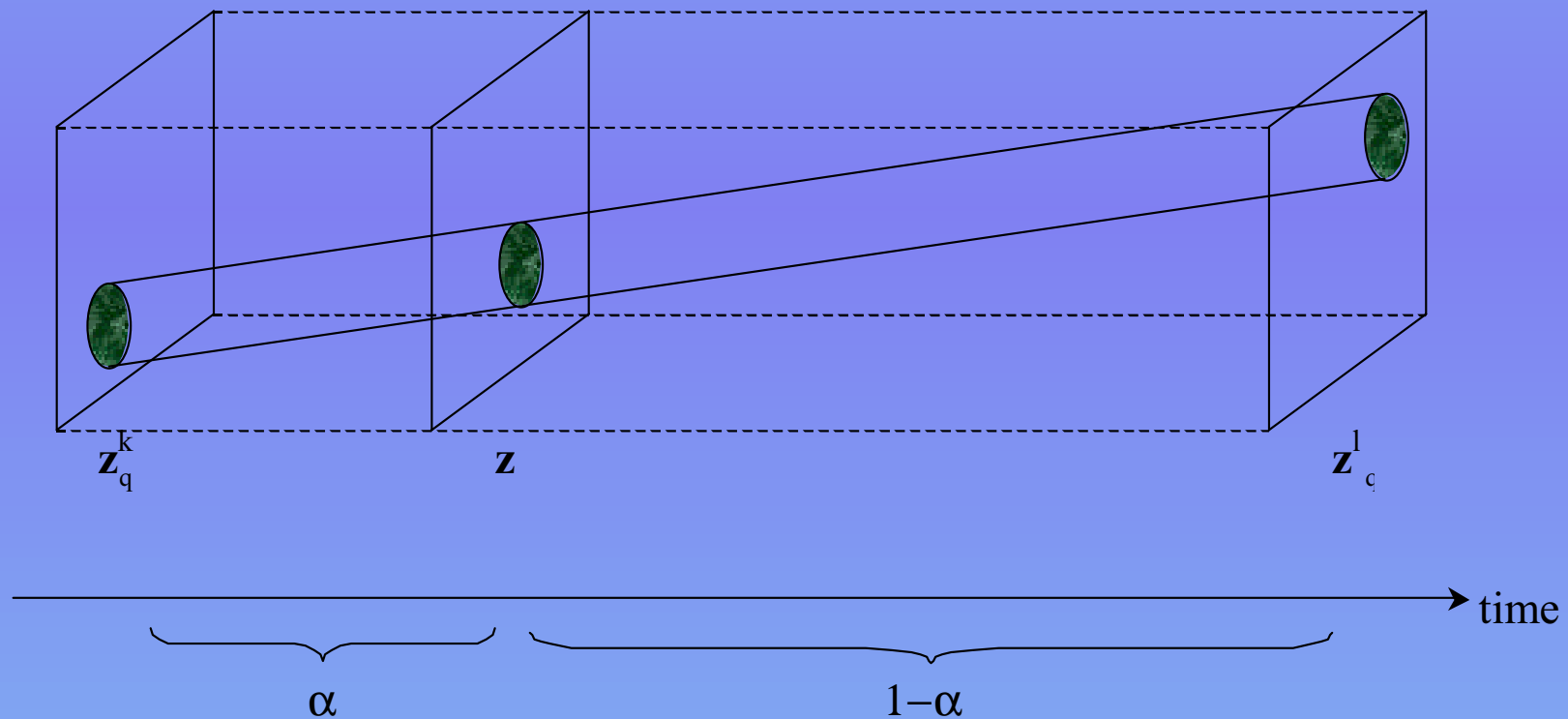
$$\mathbf{z}^k = \mathbf{A}^k \mathbf{z} + \mathbf{n}_m^k$$

\mathbf{A}^k is the *motion-compensation* matrix.

\mathbf{n}_m^k is the error in predicting \mathbf{z}^k from \mathbf{z}



Motion compensation matrix from \mathbf{z}^k to \mathbf{z} is determined based on motion from \mathbf{z}_q^k to \mathbf{z}_q^l assuming constant velocity:



Now, include quantization noise in the model:

$$\begin{aligned}\mathbf{z}_q^k &= \mathbf{z}^k + \mathbf{e}_z^k \\ &= \mathbf{A}^k \mathbf{z} + \mathbf{n}_m^k + \mathbf{e}_z^k \\ &= \mathbf{A}^k \mathbf{z} + \mathbf{n}^k\end{aligned}$$

\mathbf{n}^k is the noise (error) between \mathbf{z}_q^k and its motion compensation from \mathbf{z} .

$$\begin{aligned}\mathbf{e}_z^k &\implies \text{assume Gaussian with } \mathbf{K}_{\mathbf{e}_z}^k \\ \mathbf{n}_m^k &\implies \text{assume Gaussian with } \mathbf{K}_m^k\end{aligned}$$

\mathbf{e}_z^k Quantization noise is characterized in the spatial domain by explicitly considering the quantization intervals used on DCT-domain signal. In general, *not* IID.

\mathbf{n}_m^k Motion compensation noise allows high variance to be assigned to areas of low confidence—e.g., appearing or disappearing objects, lighting changes, motion estimation errors.

(more details in paper)



Assuming conditional independence

$$p(\mathbf{z}_q^k, \mathbf{z}_q^l | \mathbf{z}) = p(\mathbf{z}_q^k | \mathbf{z}) p(\mathbf{z}_q^l | \mathbf{z})$$

where

$$p(\mathbf{z}_q^k | \mathbf{z}) = \frac{(2\pi)^{-\frac{N}{2}}}{|\mathbf{K}^k|^{\frac{1}{2}}} \exp \left\{ -\frac{1}{2} (\mathbf{A}^k \mathbf{z} - \mathbf{z}_q^k)^t \mathbf{K}^k^{-1} (\mathbf{A}^k \mathbf{z} - \mathbf{z}_q^k) \right\}$$

and

$$\mathbf{K}^k = \mathbf{K}_{\mathbf{e}_z}^k + \mathbf{K}_m^k$$

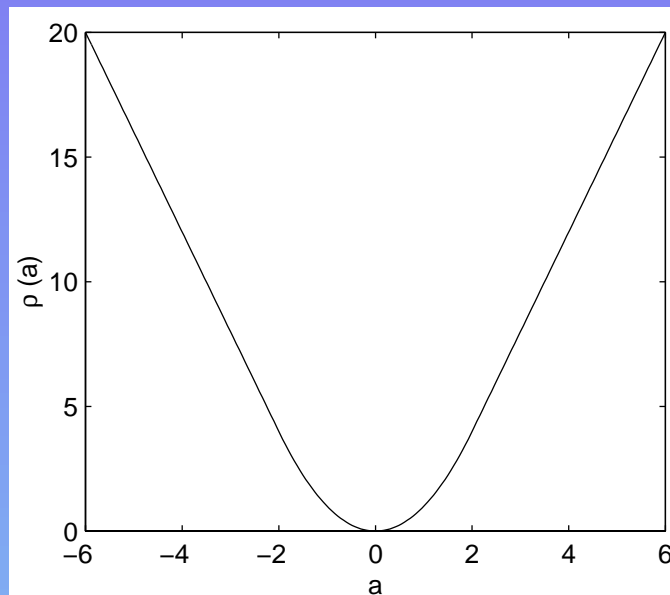


Prior Term

Huber Markov Random Field (HMRF):

$$p(\mathbf{z}) = \frac{1}{G} \exp \left\{ -\lambda_I \sum_{c \in \mathcal{C}} \rho_T(\mathbf{d}_c^t \mathbf{z}) \right\}$$

where Huber function $\rho_T(\cdot)$ looks like



Prior term has two main effects:

- Smooths artifacts such as blocking and ringing
- Fills in “gaps” in \mathbf{z} , to which no motion vectors from \mathbf{z}_q^k or \mathbf{z}_q^l point



Resulting optimization:

$$\hat{\mathbf{z}} = \arg \min_{\mathbf{z}} [\lambda_F r(\mathbf{z}) + s(\mathbf{z})]$$

where $r(\mathbf{z})$ is from the prior smoothness term, and $s(\mathbf{z})$ is from the likelihood term.



Solve using gradient descent:

Gradient is $\mathbf{g}(\mathbf{z}) = \lambda_F \nabla r(\mathbf{z}) + \nabla s(\mathbf{z})$

Estimate at k^{th} iteration is updated as

$$\mathbf{z}^{(k+1)} = \mathbf{z}^{(k)} - \tau^{(k)} \mathbf{g} \left(\mathbf{z}^{(k)} \right)$$



received frame 1: ■ □ □ □



received frame 2: □□□■



received frame 1: ■ □ □ □



inserted frame 1:



inserted frame 2: □ □ ■ □



received frame 2: □□□■



received frame 1: ■ □ □ □ □



received frame 2: □□□□■



received frame 1: ■ □ □ □ □



inserted frame 1:



inserted frame 2: □ □ ■ □ □



inserted frame 3: □ □ □ ■ □



received frame 2: □□□□■



Conclusion

Frame rate adjustment in compressed video that:

- Uses Bayesian estimation
- Explicitly considers the effect of compression noise in received video frames
- Uses prior image model to smooth compression artifacts, as well as “fill in” motion gaps in inserted frames

