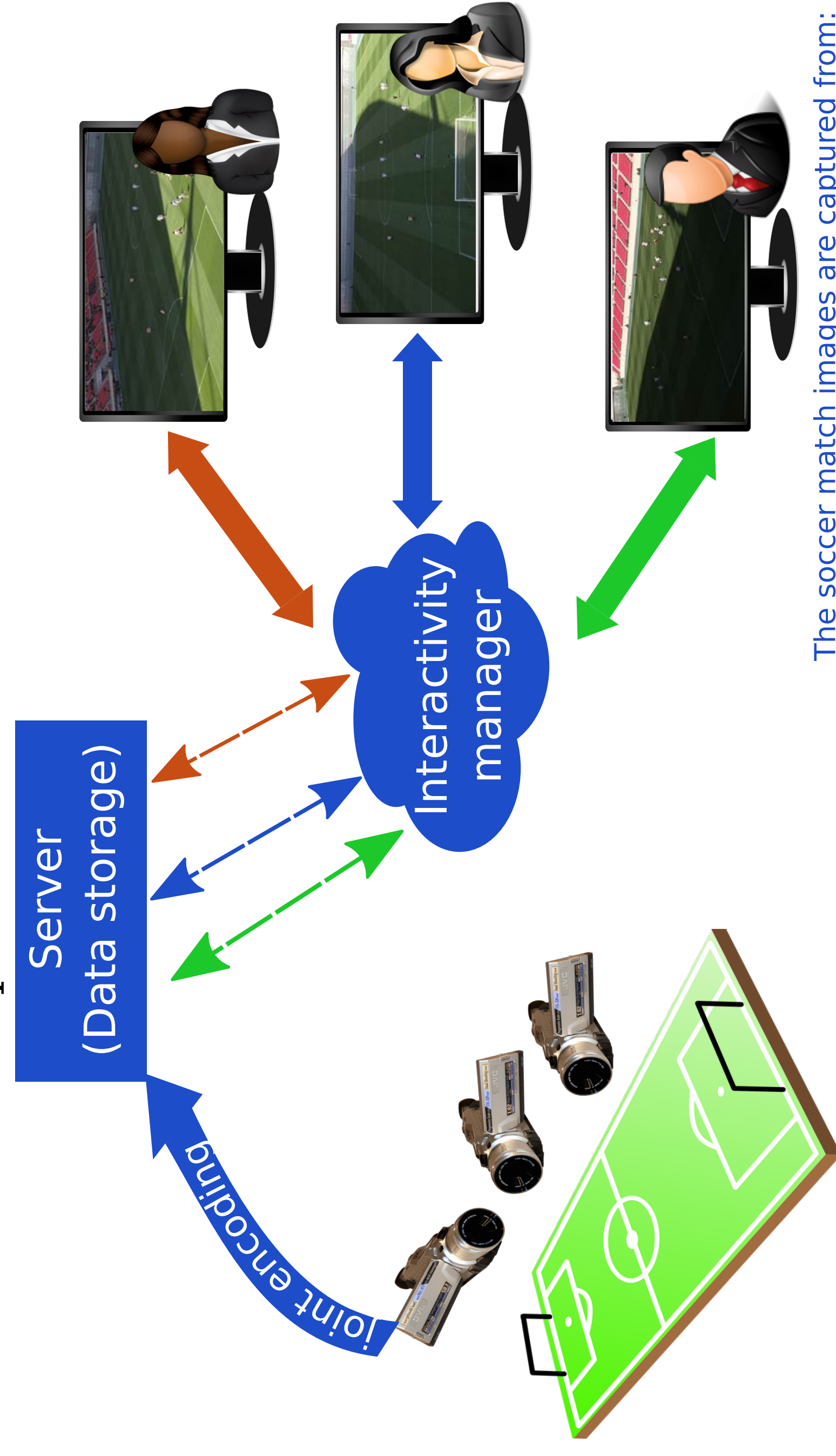


Correlative Model Selection for Interactive video Communication

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Introduction

Interactivity: user can switch within views to see the scene from different viewpoints.

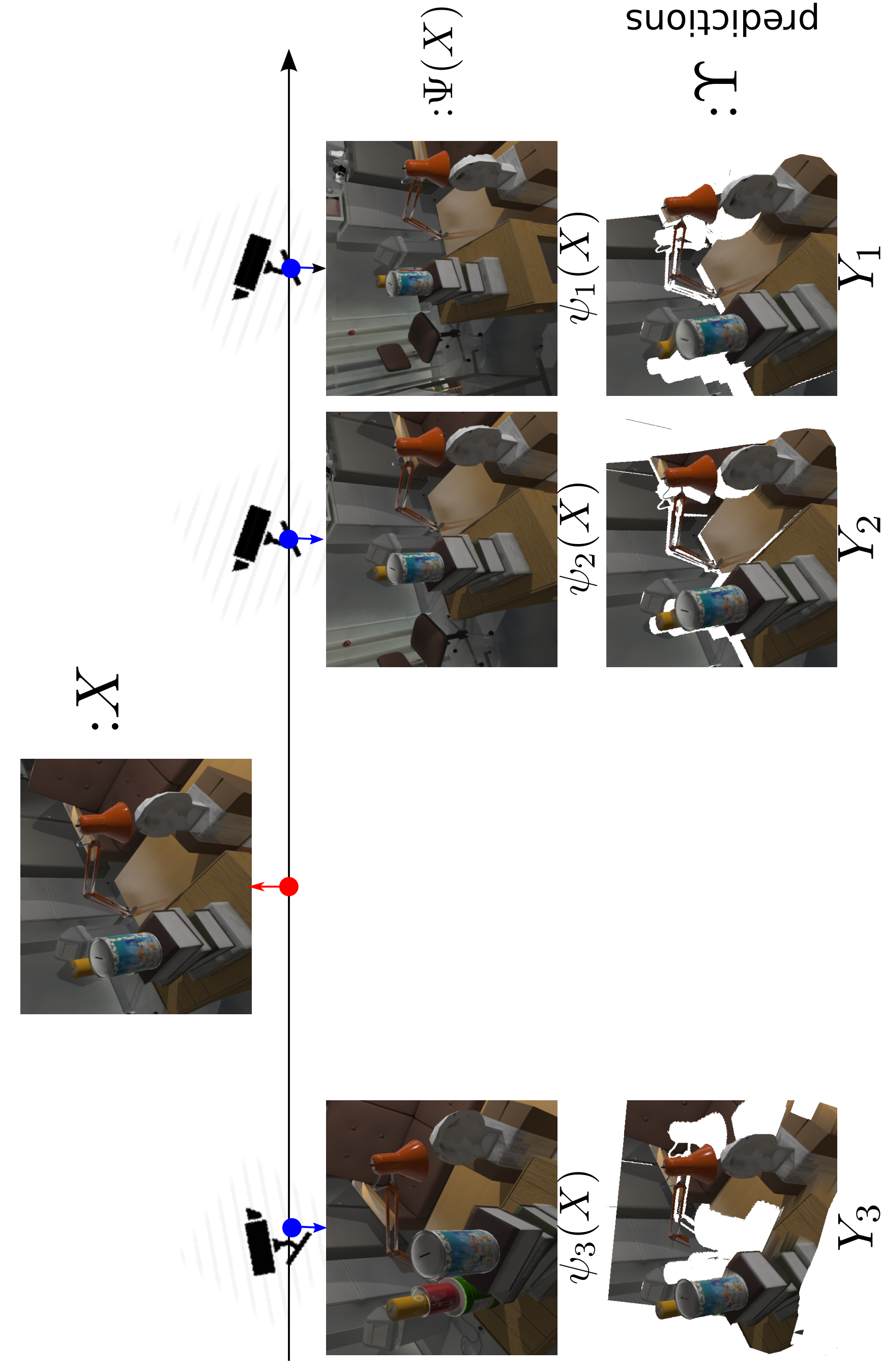


compress All videos

- Storing the videos on the server as **compact** as possible.
- Transmitting at **minimum rate** between server & the user.
- **Low complexity** at server (Massive Random Access)

Interactive Video Coding (IVC)

The views are correlated and one should exploit this correlation through prediction:



Model based coding

Classic (predictive) coding:

- Encode residual between prediction and frame, i.e. $X-Y$

In IVC reference is different for each user.

Model based coding:

- Use Slepian-Wolf (SW) coder to perform compression.
- No need for the exact residual. Only its distribution is needed
- Model the correlation between current frame and set of predictions coming from possible previous frames.

Objective:

- What is the **impact** if we use an **approximate model**?
- What is a **good model** to **select**?

Optimal storage and transmission rates

Let (X, Y) be a static source determined by $\{P(X, Y_i), p_i\}_i$. The achievable rate-storage pair in lossless coding is:

$$R \geq H(X|Y_{i^*})$$

no loss
(better than expected)

$$S \geq \max_{j \in \{1, \dots, K\}} H(X|Y_j)$$

loss in storage only
(as expected)

A. Roumy and T. Maugey,
"Universal lossless coding with random user access: The cost of interactivity," ICP 2015

where Y_{i^*} is the prediction available at the decoder.

Fixed Length source coding in IVC

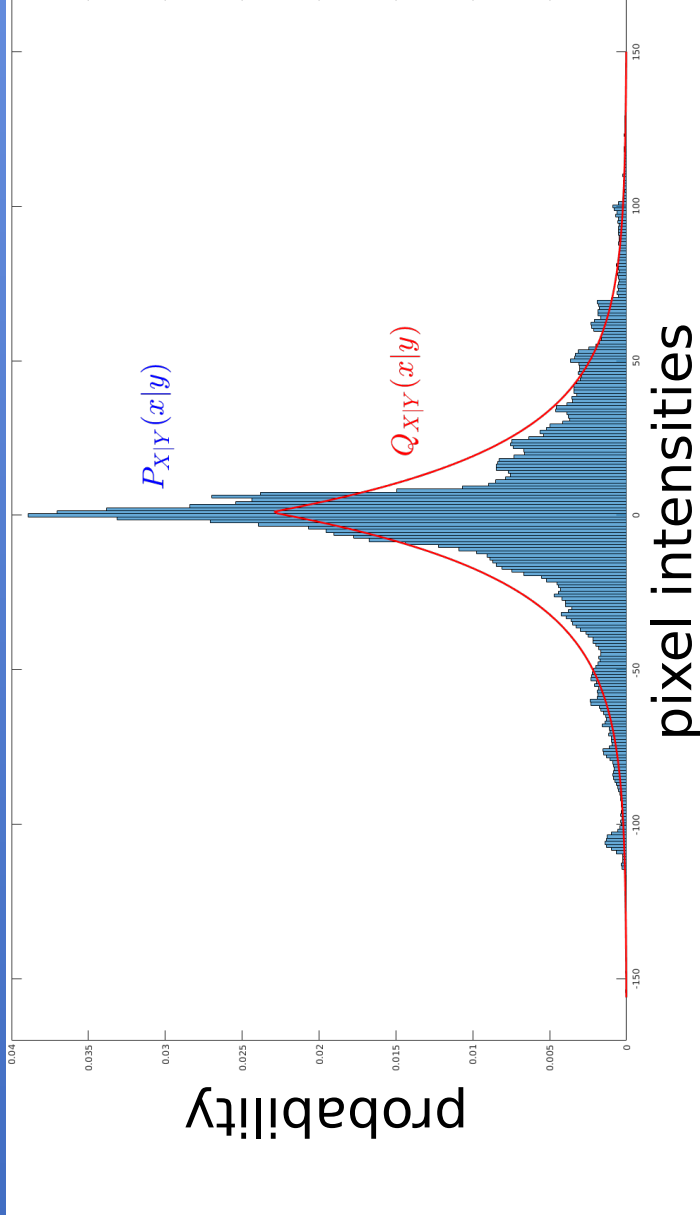
$A(X|Y) :=$ set of typical sequences of $X^n = (X_1, X_2, \dots, X_n)$ given Y^n . We need $nH(X|Y)$ bits on average to encode X^n given Y^n

Example with 2 predictions Y_1, Y_2 :

- Assign codeword of length $nH(X|Y)$ to each $x^n \in A(X|Y_1)$
- upon encoding $x^n \in A(X|Y_2)$:
 - $x^n \in A(X|Y_1) \cap A(X|Y_2)$: same length- n_1 as used in $A(X|Y_1)$ padded with $n_2 - n_1$ zeros
 - use the remaining length- n_2 codewords for the rest

Cost of using approximate distribution

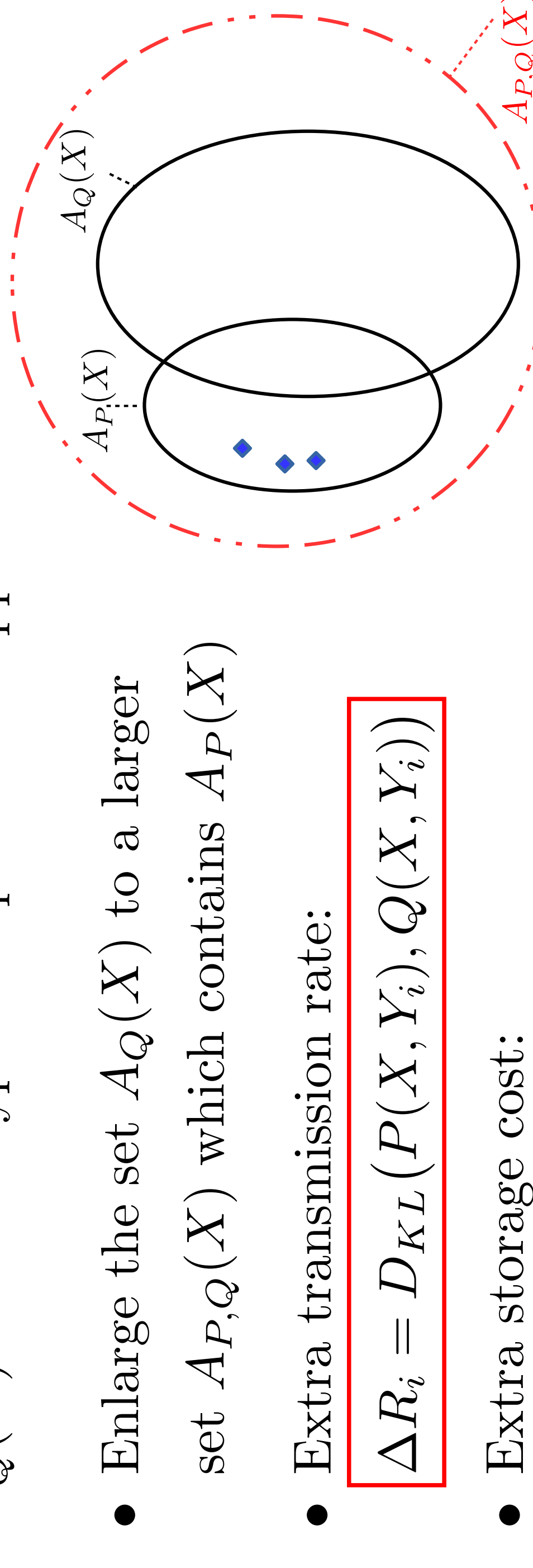
Idea: send θ_i parameters of approximate Distribution $Q_{X|Y}$ instead of whole $P_{X|Y}$.



Vanishing error probability:

$A_P(X) :=$ set of typical sequences of true distribution

$A_Q(X) :=$ set of typical sequences of approximate distribution



- Enlarge the set $A_Q(X)$ to a larger set $A_{P,Q}(X)$ which contains $A_P(X)$

- Extra transmission rate:

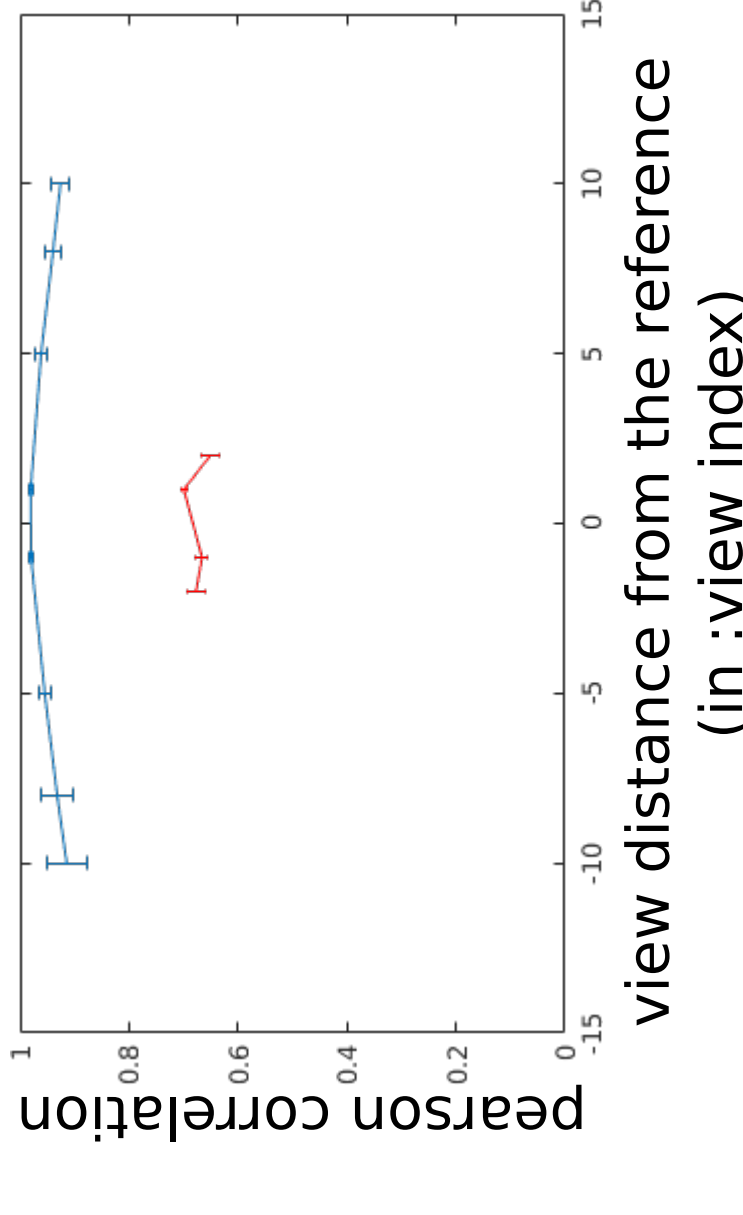
$$\Delta R_i = D_{KL}(P(X, Y_i), Q(X, Y_i))$$

- Extra storage cost:

$$\Delta S_X = \max_j [H(X|Y_j) + \Delta R_j] - \max_j H(X|Y_j)$$

Experimental results

Predictions are provided by Depth-Image Based Rendering



Validity of additive model:

- Linear dependency evaluated
- $X = Y+Z$

Model selection:

	Tsukuba		Ballet	
	ΔS	ΔR	ΔS	ΔR
Laplace	0.634	0.610	0.552	0.3
Normal	1.43	1.401	0.996	0.555
Gaussian Mixture	0.211	0.2	0.364	0.19
Erasure channel	Inf	Inf	Inf	Inf

Extra rate and extra storage cost in bit per pixel. Inf means infinity

Conclusion

- A criterion is proposed for model selection in IVC.
- Several distribution evaluated. It is not possible to code a view losslessly by completely relying on DIBR prediction.