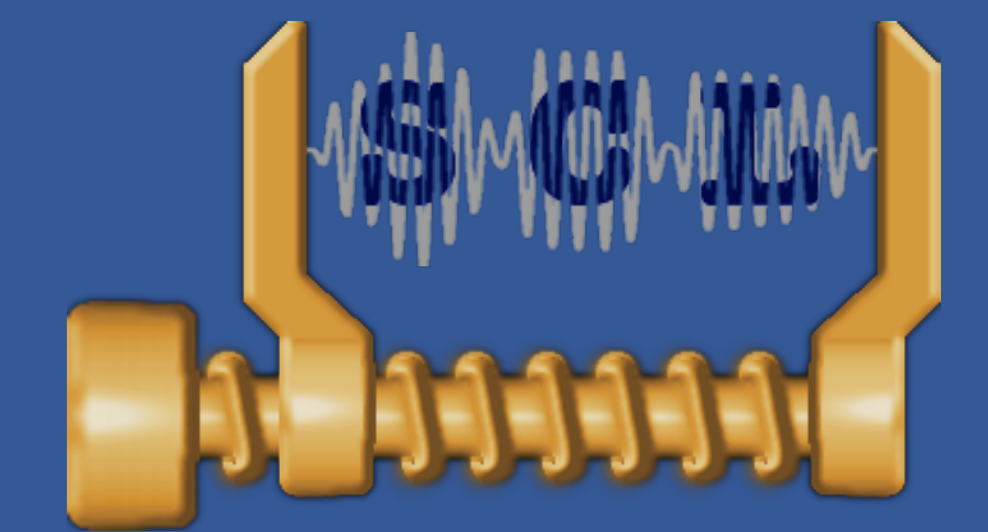




ASYMPTOTIC CLOSED-LOOP DESIGN FOR TRANSFORM DOMAIN TEMPORAL PREDICTION

SHUNYAO LI, TEJASWI NANJUNDASWAMY, YUE CHEN and KENNETH ROSE
SIGNAL COMPRESSION LAB, ECE DEPARTMENT, UNIVERSITY OF CALIFORNIA, SANTA BARBARA

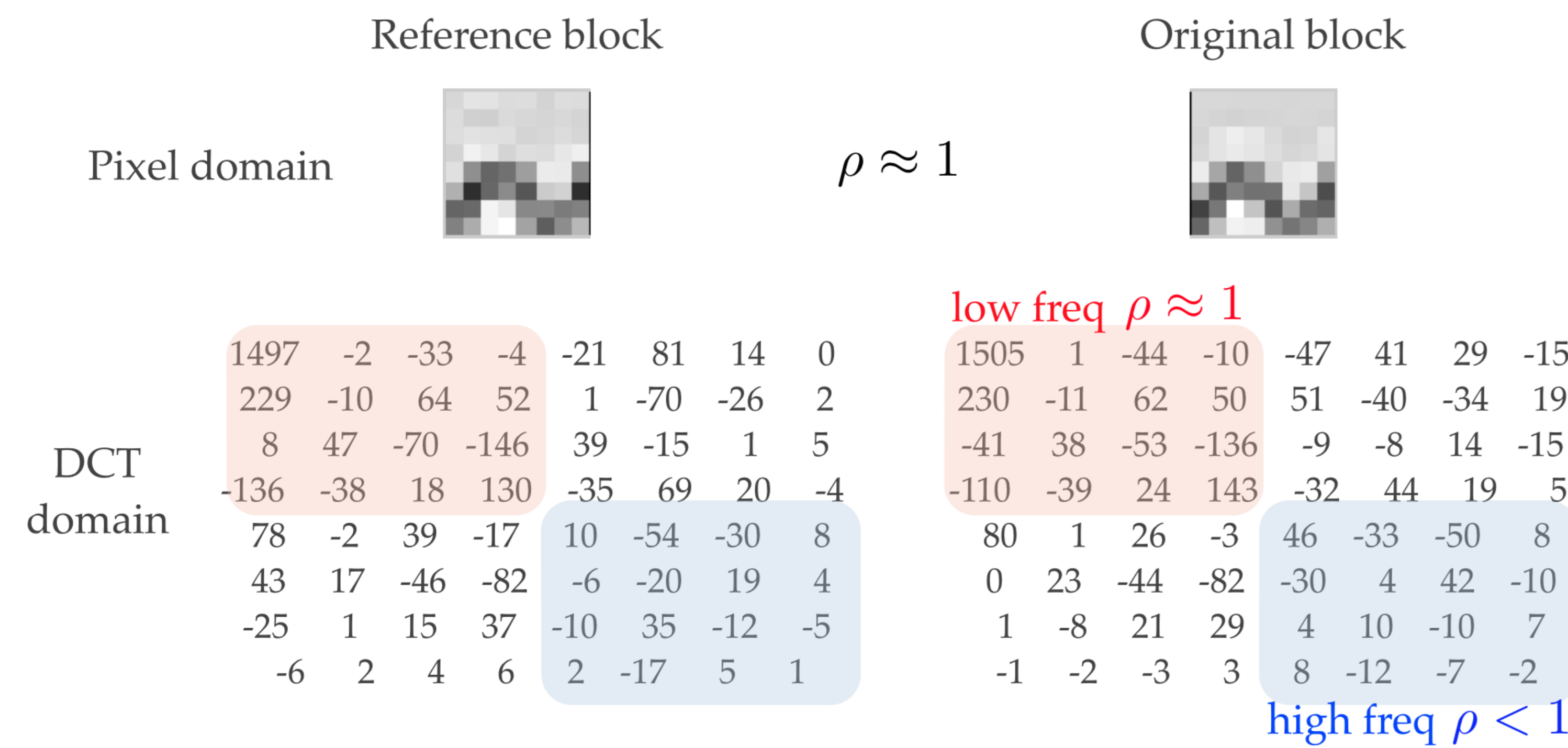


TRANSFORM DOMAIN TEMPORAL PREDICTION (TDTP)

Pixel-Domain Inter Prediction is Suboptimal

- Traditional inter prediction copies pixels one-by-one.
- Suboptimal because it ignores spatial correlation.
- Transform Domain Temporal Prediction (TDTP): DCT (largely) achieves spatial decorrelation, enabling optimal one-to-one prediction.

Hidden Temporal Correlation at High Frequency



- Correlation in pixel domain is dominated by the low frequencies ($\rho \approx 1$), inspiring the traditional pixel copying prediction.
- TDTP: Accounts for variation in temporal correlation across frequency, which is hidden in pixel domain.

Sub-Pixel Motion Compensation Interferes with TDTP

- The interpolation low-pass filters scale down high-frequencies as per its magnitude response.
- Thus we apply TDTP conditioned on the sub-pixel location.

CLOSED-LOOP PREDICTOR DESIGN

- The optimal predictor for each transform domain coefficient is given by,

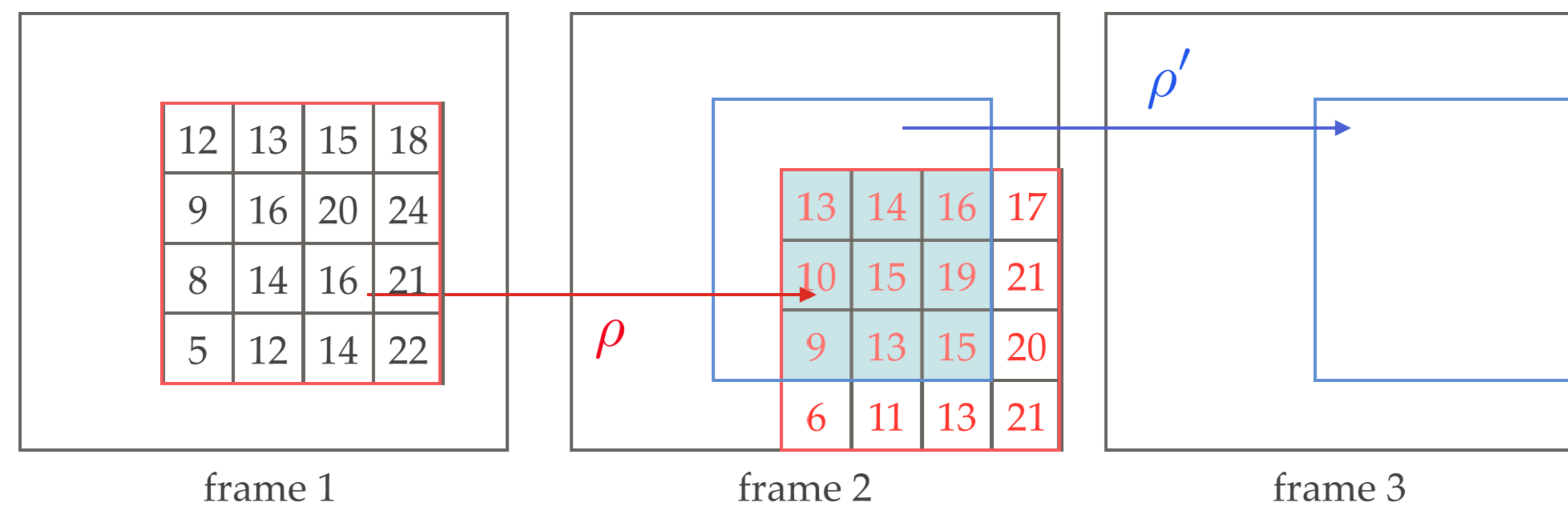
$$\tilde{x}_n = \rho \hat{x}_{n-1}$$

- Given the motion compensated reference blocks, the optimal prediction coefficient is the correlation coefficient,

$$\rho = \frac{E(x_n \hat{x}_{n-1})}{E(\hat{x}_{n-1}^2)}$$

INSTABILITY DUE TO QUANTIZATION ERROR PROPAGATION

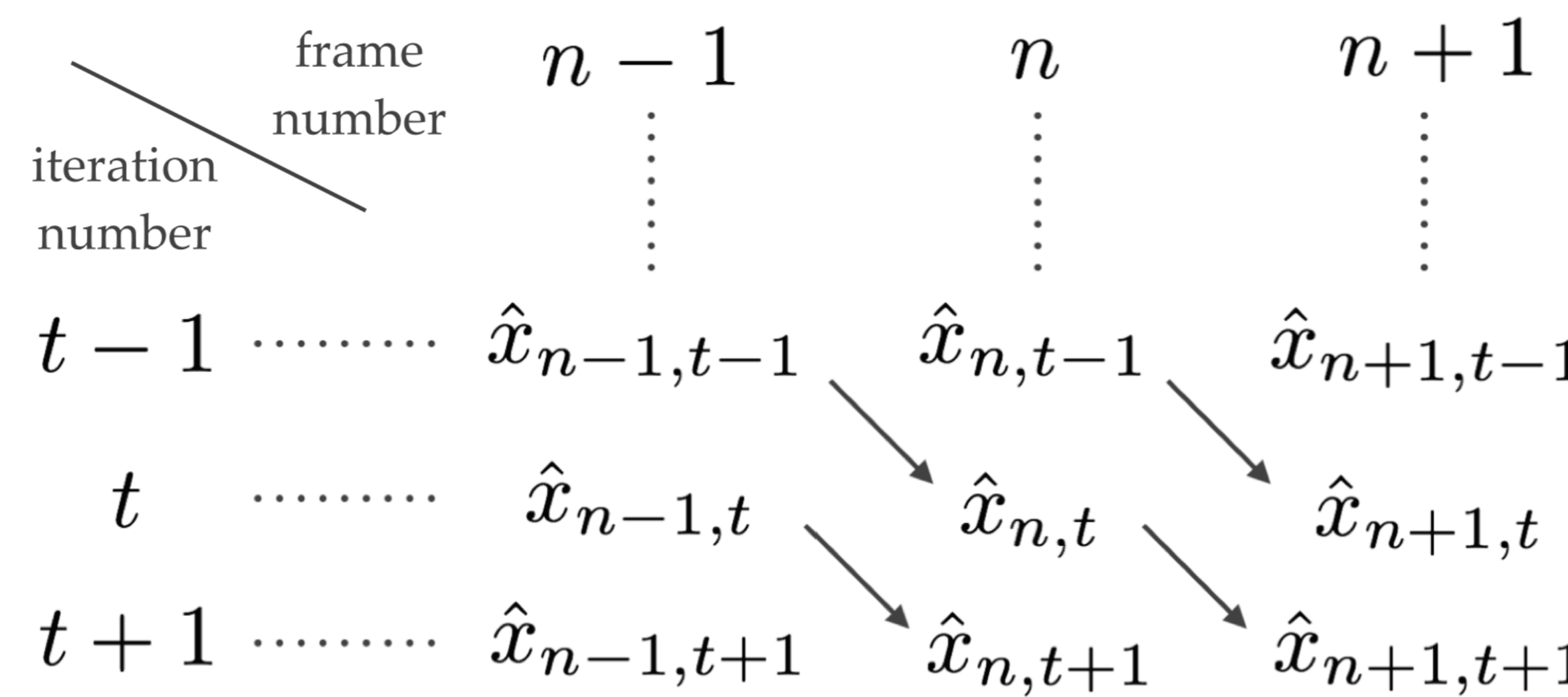
- As we operate this predictor in closed loop, the new reconstructed frames (which are prediction reference for future frames) have different statistics, for which the correlation coefficient $\rho' \neq \rho$.



- This deviation in statistics between design and operation grows over frames.
- Thus we propose the asymptotic closed-loop (ACL) design approach for TDTP.

ASYMPTOTIC CLOSED-LOOP (ACL) DESIGN

- ACL is an iterative open-loop design technique that asymptotically optimizes the system for closed-loop operation.



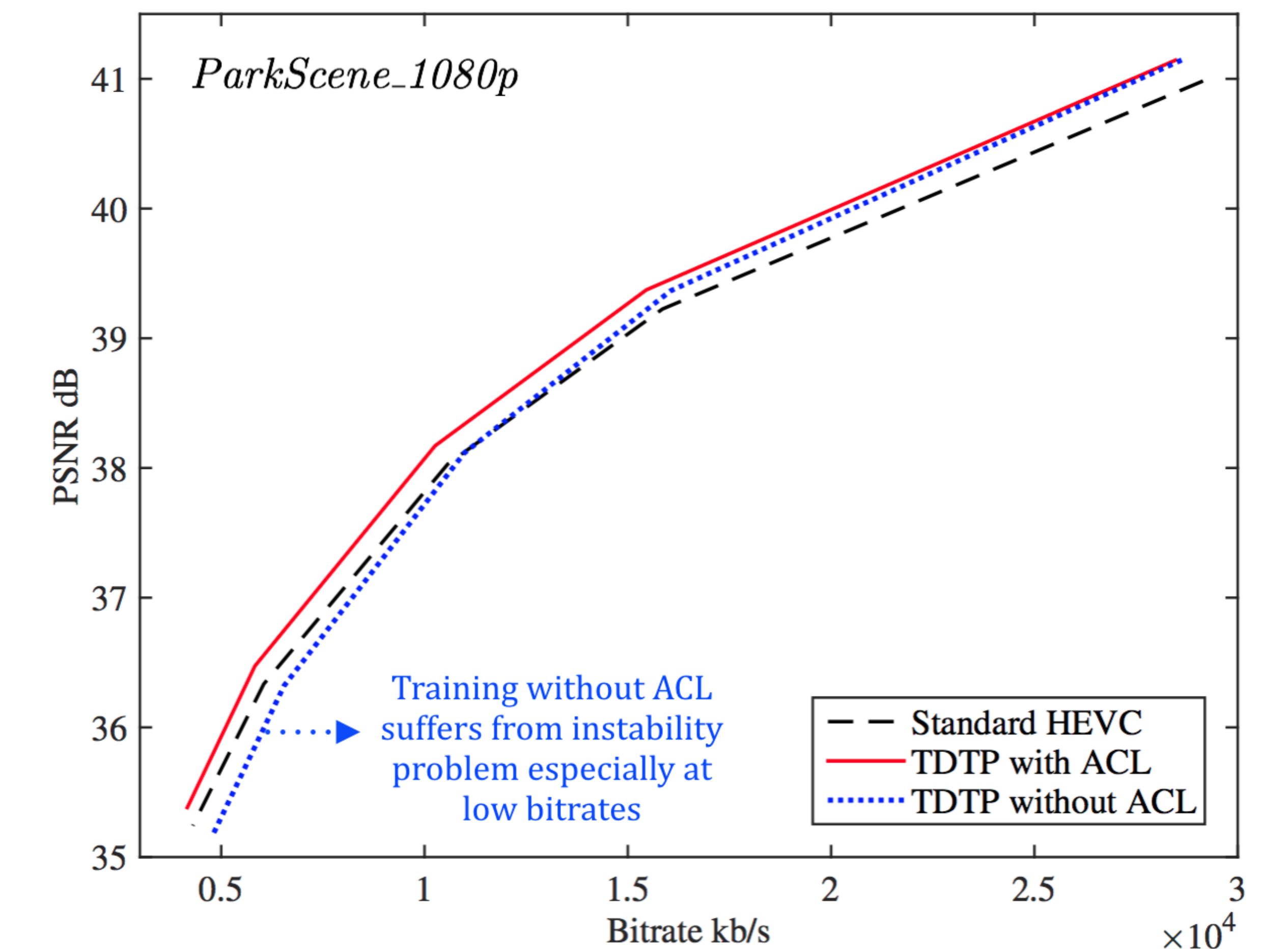
- Given the reconstructed data at iteration $t - 1$, $\hat{x}_{n-1,t-1}$ ($n = 2 \dots N$), estimate the prediction coefficient for iteration t as, $\rho_t = E(x_n \hat{x}_{n-1,t-1}) / E(\hat{x}_{n-1,t-1}^2)$.
- Then employ open-loop prediction to generate, $\tilde{x}_{n,t} = \rho_t \hat{x}_{n-1,t-1}$.
- Since ρ_t is directly optimized for the statistics of $\hat{x}_{n-1,t-1}$ ($n = 2 \dots N$), the prediction $\tilde{x}_{n,t}$ is guaranteed to improve.
- Better prediction usually leads to better reconstruction, $\hat{x}_{n,t}$, and vice versa.
- The reconstruction error decreases over iterations and on convergence, $\hat{x}_{n,t-1} = \hat{x}_{n,t}$, which is equivalent to the closed-loop system.

TWO LOOP ASYMPTOTIC CLOSED-LOOP (ACL) DESIGN FOR TDTP

- In video coding, encoder decisions (e.g. mode decisions, motion vectors, quantization, etc.) are dependent on the prediction.
- Thus we proposed a two-loop design scheme:
Inner loop: Estimate prediction coefficient ρ via ACL with encoder decisions fixed.
Outer loop: Update encoder decisions with ρ fixed.
- We design different prediction coefficients conditioned on: sub-pixel location, quantization parameter (QP), skip/non-skip mode

EXPERIMENTAL RESULTS

- The proposed approach was implemented in HM 14.0. Both prediction size and transform size are restricted to 8x8, and the motion search is at half-pixel precision.
- The average bitrate reduction over standard HEVC is **6.53%** for training set (Exp1) and **4.96%** outside training set (Exp2).
- In Exp2 we provide a choice of 8 sets of prediction coefficients at the encoder with an overhead of 3 bits per sequence. The difference between the two experiments suggests further scope for adaptivity.



Sequence	Exp1	Exp2	Sequence	Exp1	Exp2
BQTerrace	12.88%	10.23%	Waterfall	9.29%	4.38%
BasketballDrive	6.21%	5.36%	Vidvo1	5.75%	4.56%
Kimono	7.88%	4.53%	Bus	5.59%	5.37%
ParkScene	7.41%	7.00%	Tennis	2.23%	1.50%
Keiba	5.38%	5.17%	Tempete	5.56%	5.30%
RaceHorse	3.44%	3.03%	FourPeople	6.81%	3.04%