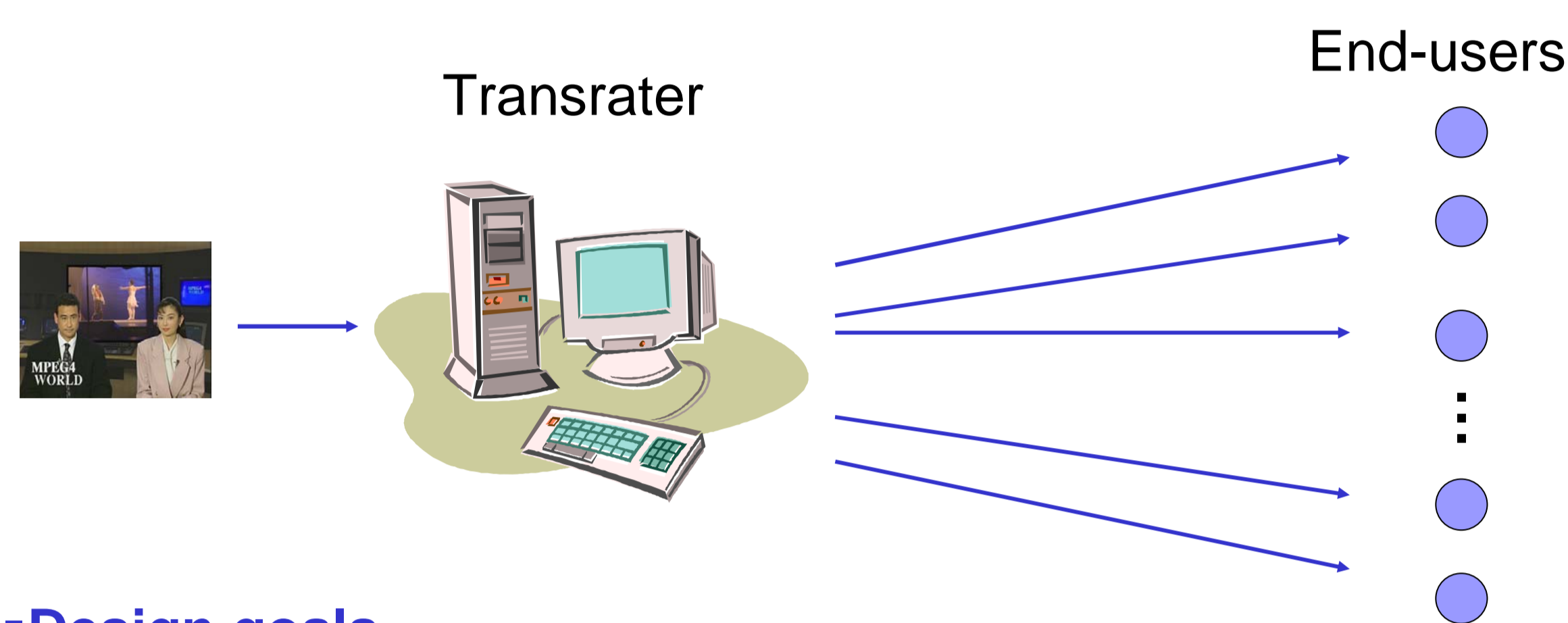


Transrating of coded video

- **Transrating:** bit rate reduction of pre-encoded video.
- **Method:** transform coefficients requantization.
- **Applications:** television broadcast, internet streaming



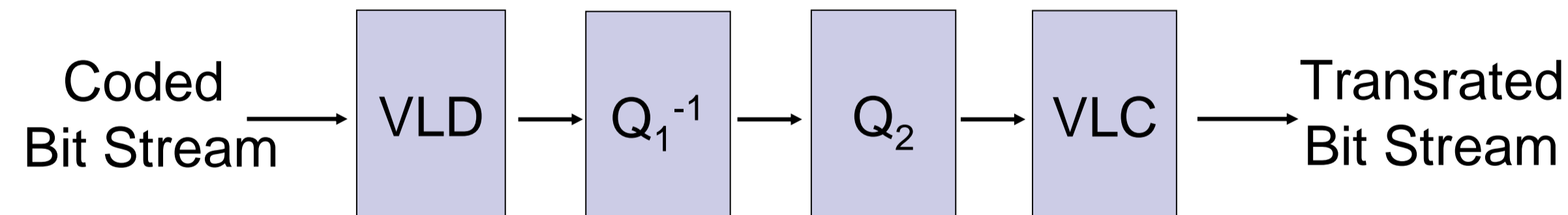
- **Design goals**
- Achieve the highest quality at the target bit rate.
- Minimal complexity – use as much information from the coded bit stream.

Transrating architectures

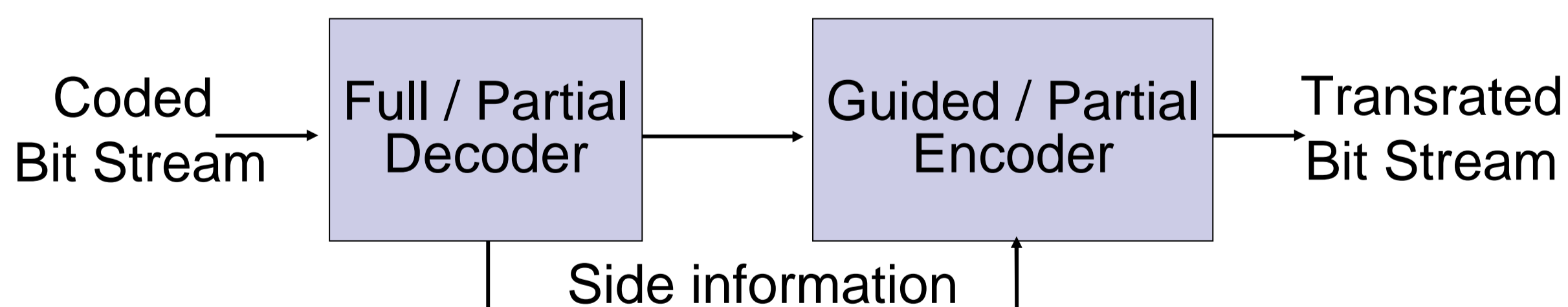
- **Trivial solution – re-encoding**
- ✗ high computational complexity due to full encoding



- **Lowest complexity – open-loop**
- ✗ drift error



- **Reuse of input coding modes**
- ✓ compromise of quality vs. computational complexity



Intra-frame transrating architectures

- **Partial Decoder – Partial Encoder**
- ✗ Decode up to the residual transform coefficients

Non-linear operations cause a drift error that cannot be fully compensated

Full Decoder – Guided Encoder

- ✓ No drift error
- Guided encoding, based on the input intra prediction modes with optional modes modification.

Model-based uniform requantization

Uniform requantization

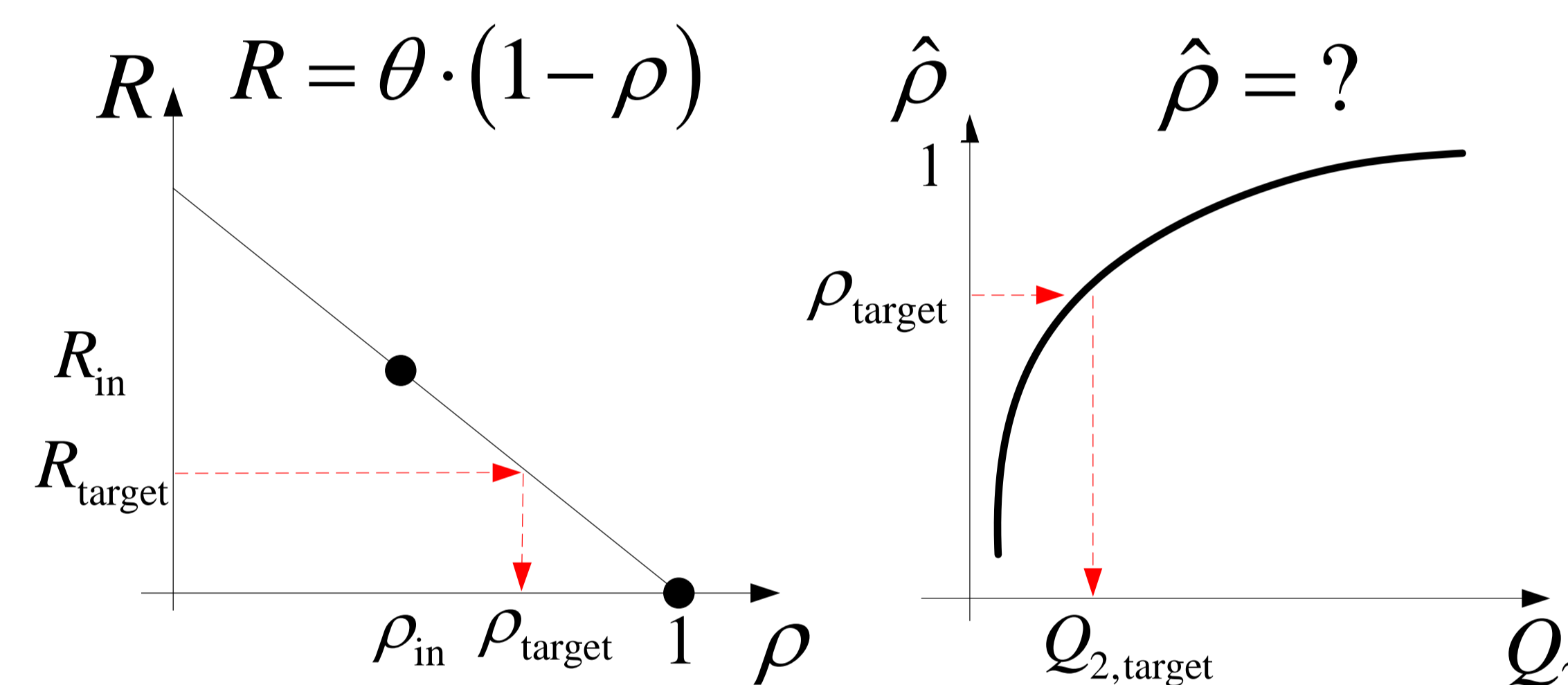
- A sufficiently high bit budget is typical (intra-frames).
- Spatial prediction → block dependencies

Model-based

- Transform coefficients requantization → evaluation of rates at many step sizes → high complexity.
- Reduce the complexity by model-based evaluation.
- Use a robust **rate model in the ρ domain** (fraction of zeroed coefficients), (He and Mitra, 2002)

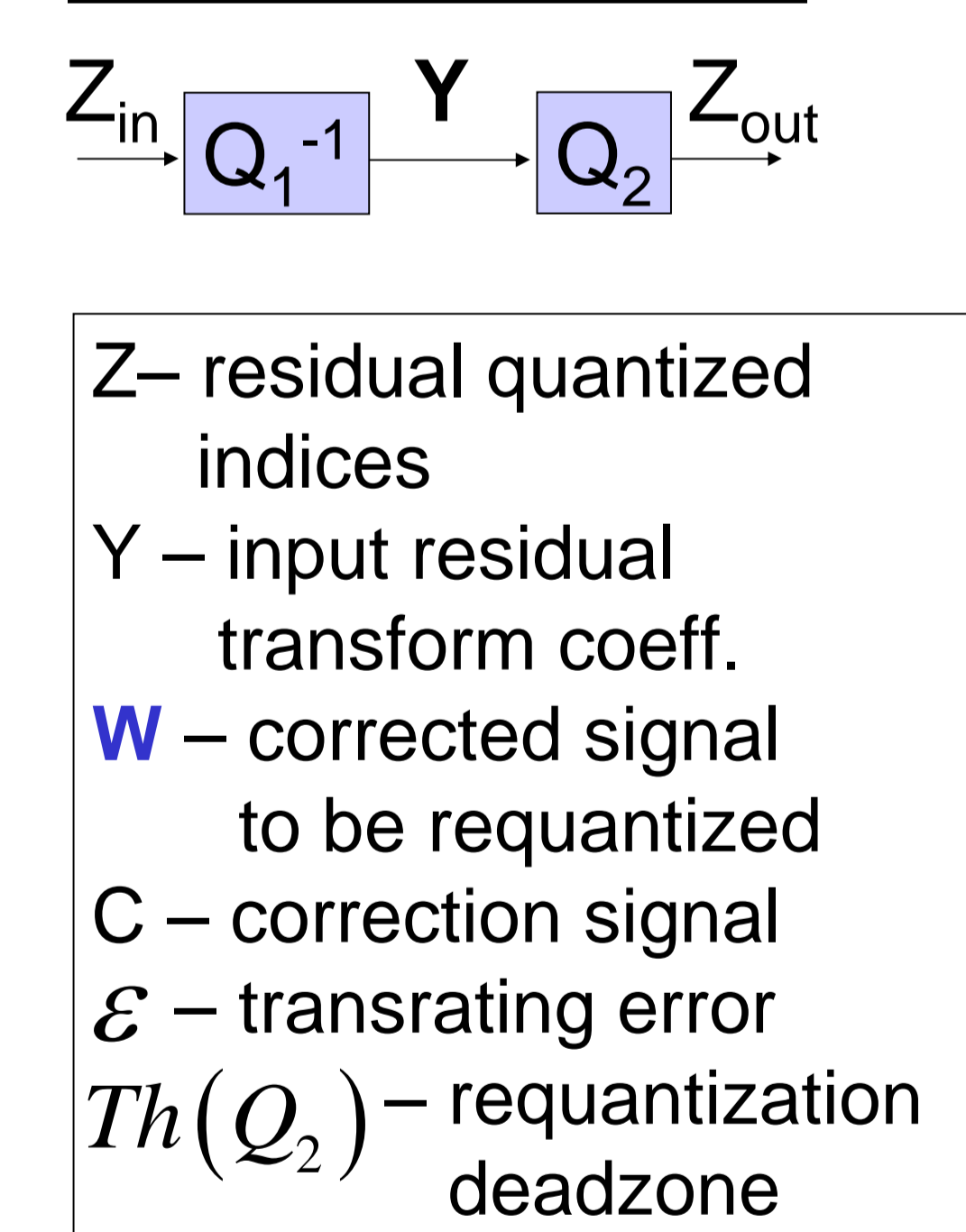
Procedure:

1. Estimate **rate-ρ** model
2. Find the target **ρ**
3. Estimate **ρ-Q₂** relation
4. Find the target **Q₂**

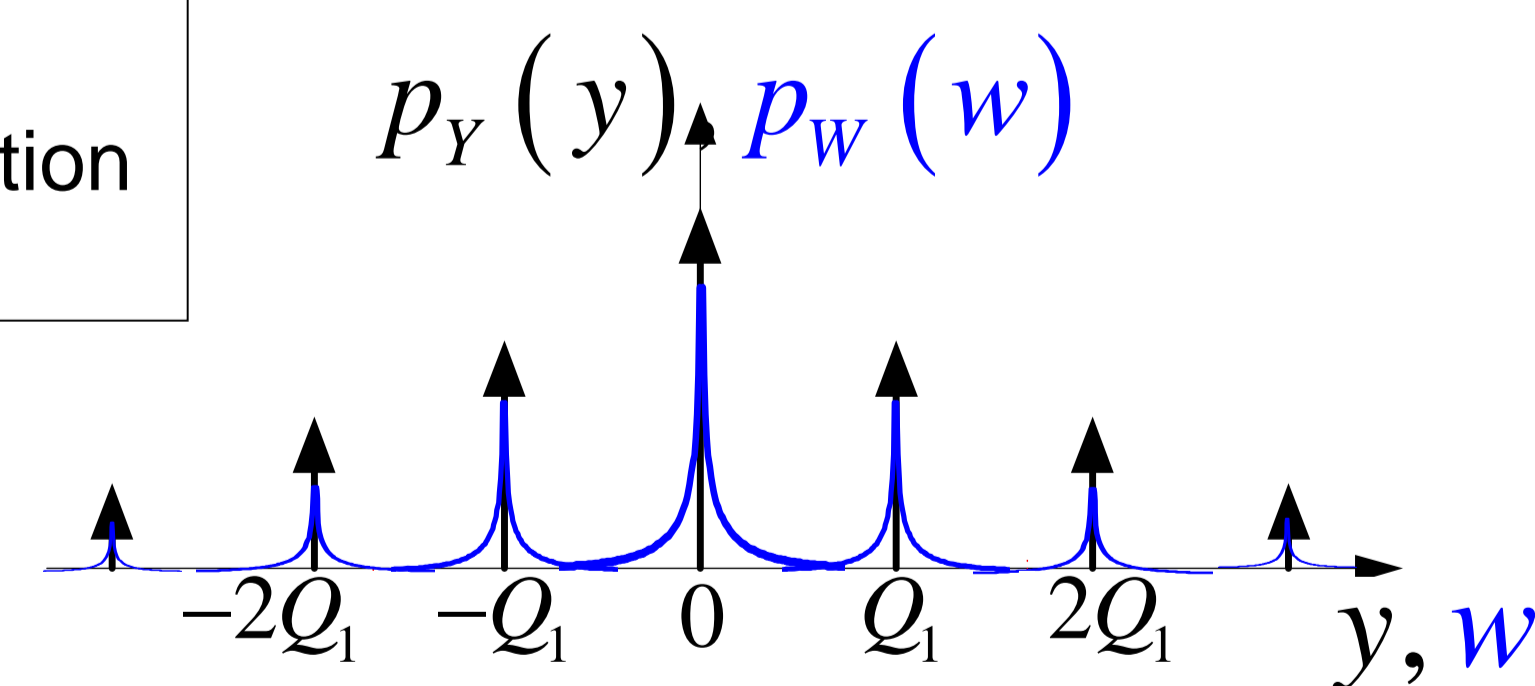
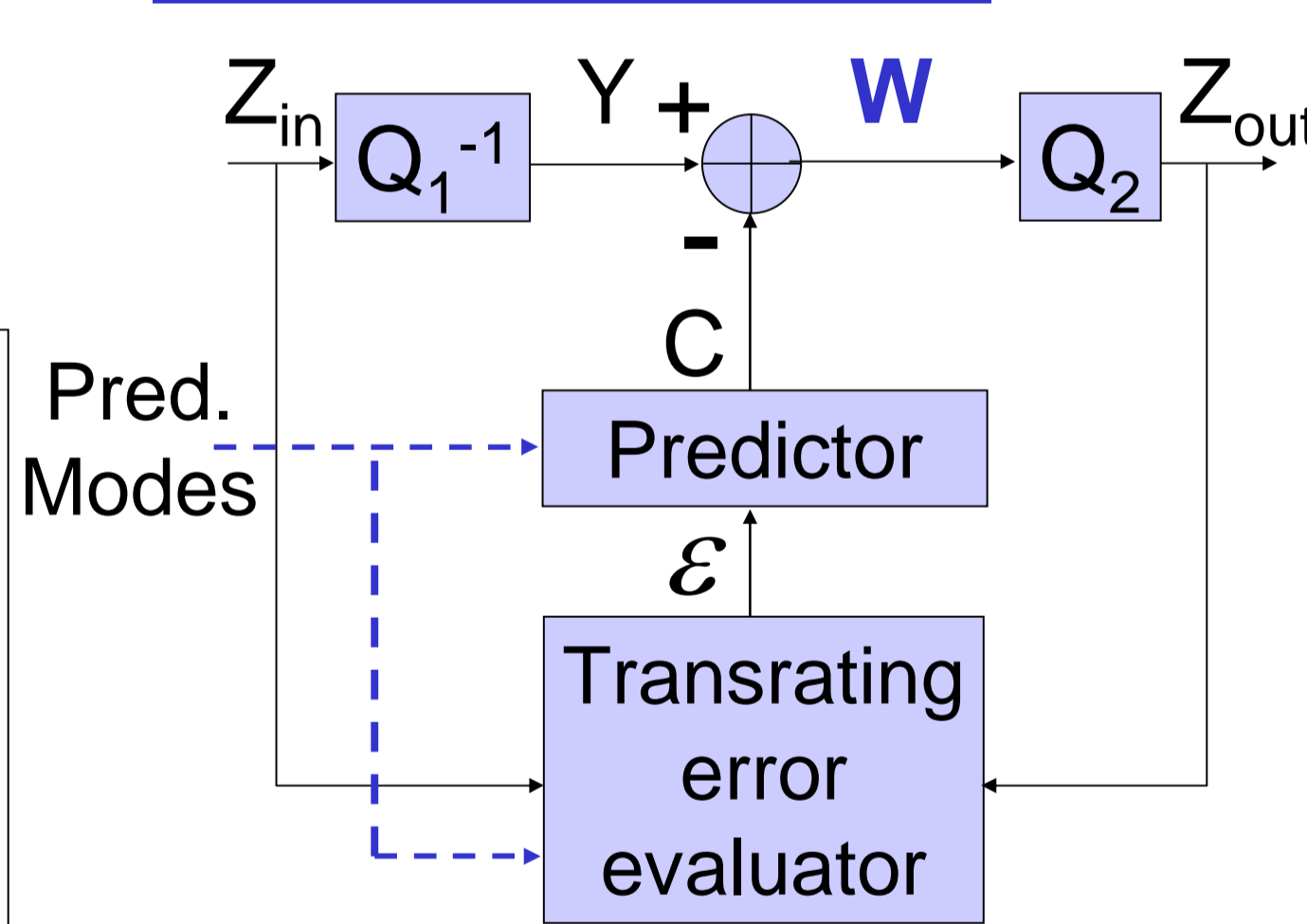


Open-loop vs. closed-loop ρ(Q₂) estimator

Open-loop estimator

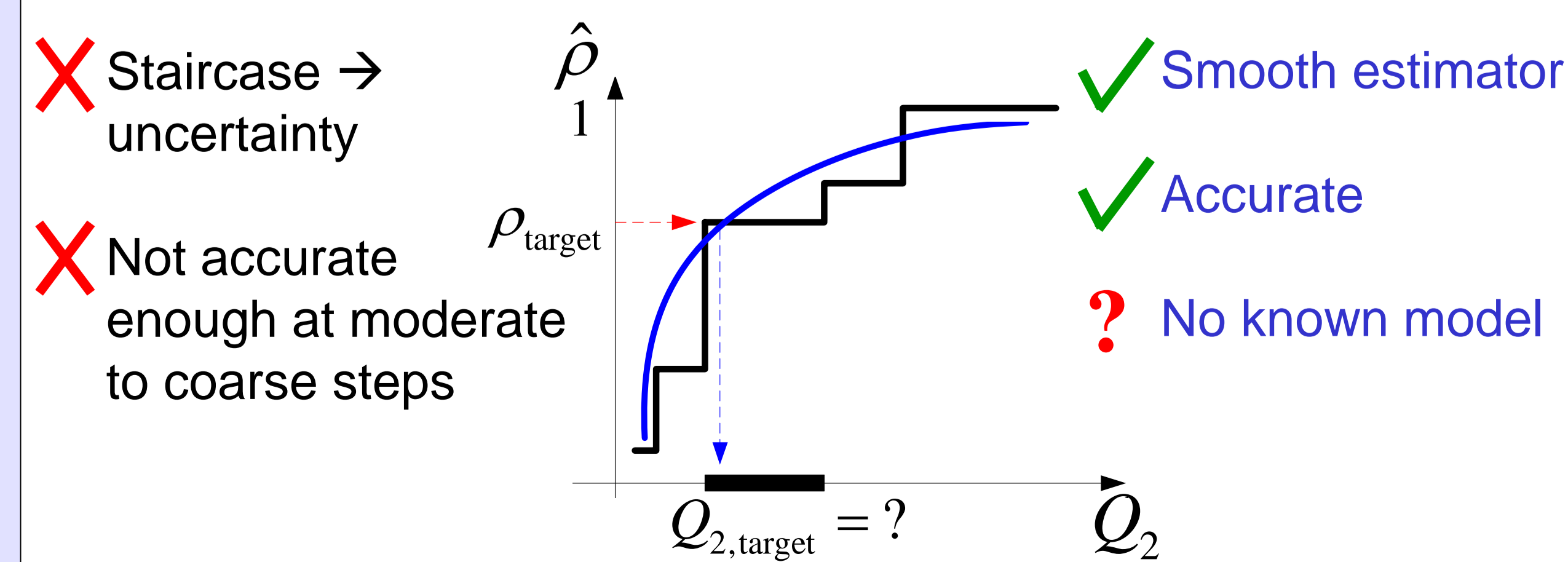


Closed-loop estimator



$$\hat{\rho}(Q_2) = \Pr.(|Y| \leq Th(Q_2))$$

$$\hat{\rho}(Q_2) = \Pr.(|W| \leq Th(Q_2))$$



Statistical ρ(Q₂) estimator

Modeling stages (closed-loop scheme just for modeling)

1. Extract the distribution of Y from the input:

$$p_Y(y) = \sum_{m=-M}^M p_m \cdot \delta(y - mQ_1)$$

2. **Model the distribution of the correction signal C**

- I. Characterization: segment into homogenous data parts
- II. Γ probability distribution
- III. Estimation of Γ distribution parameter

3. Use 1.+2. to model the distribution of W.

$$\Pr.(W \leq w_0) = \sum_{m=-M}^M p_m \cdot \int_{-\infty}^{w_0 - mQ_1} p_{-c|Y}(c|y = mQ_1) dc$$

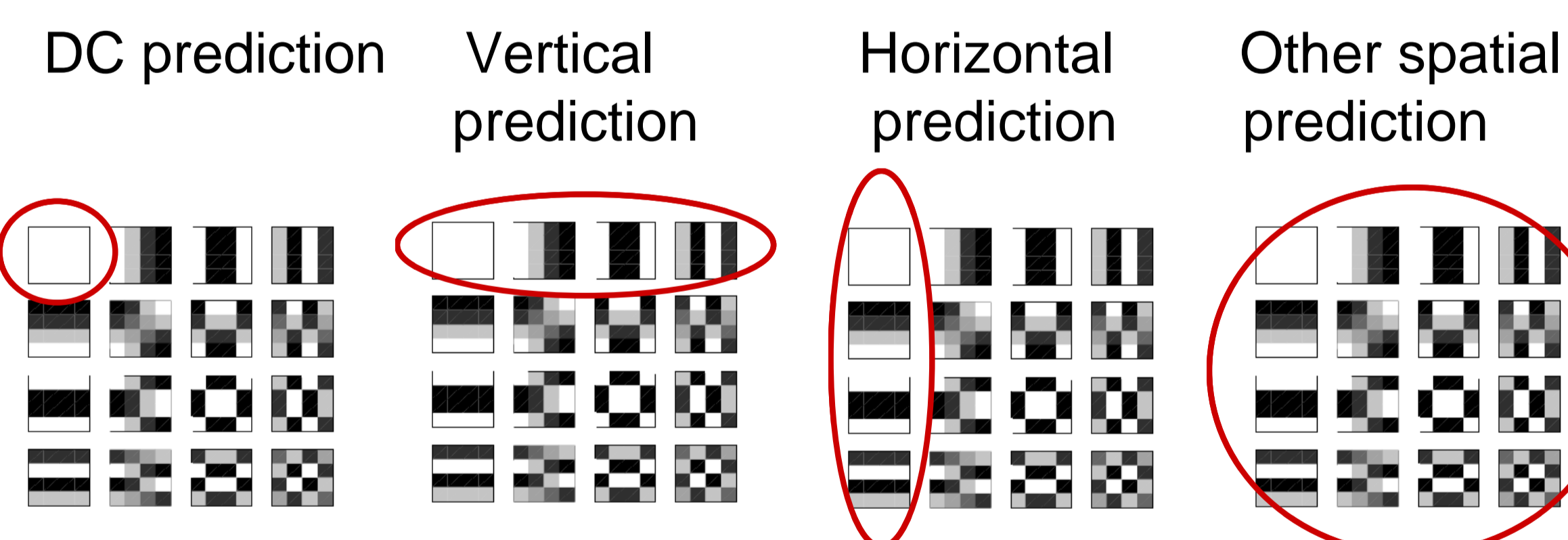
4. Estimate $\hat{\rho}(Q_2) = \Pr.(|W| \leq Th(Q_2))$

Correction signal modeling (I) Characterization of C

Segment the correction signal into groups according to the following criteria

1. **Spatial prediction modes** (e.g. DC, vertical, horizontal, various diagonal)

2. **Affected / unaffected transform coefficients:**



3. **Input quantization step-size (Q₁)**

Benefits

- ✓ Increased precision
- ✓ Complexity reduction – use open-loop estimator for unaffected coefficients.

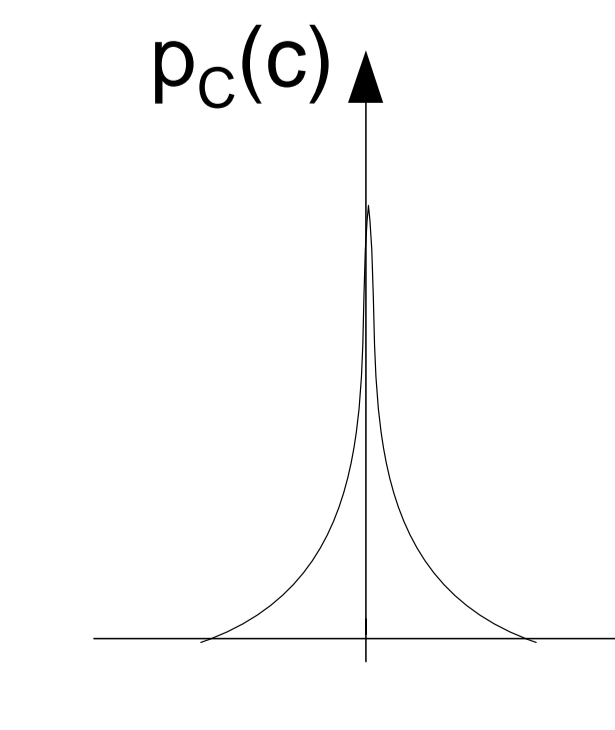
Correction signal modeling (II) Γ probability distribution

- Definition:

$$p_C(c) = \frac{1}{2\sqrt{\pi}} \sqrt{\frac{\beta}{|c|}} \cdot \exp(-\beta|c|)$$

where smaller β → wider distribution

- ML estimator: $\hat{\beta} = \frac{0.5N}{\sum_{i=1}^N |c_i|}$

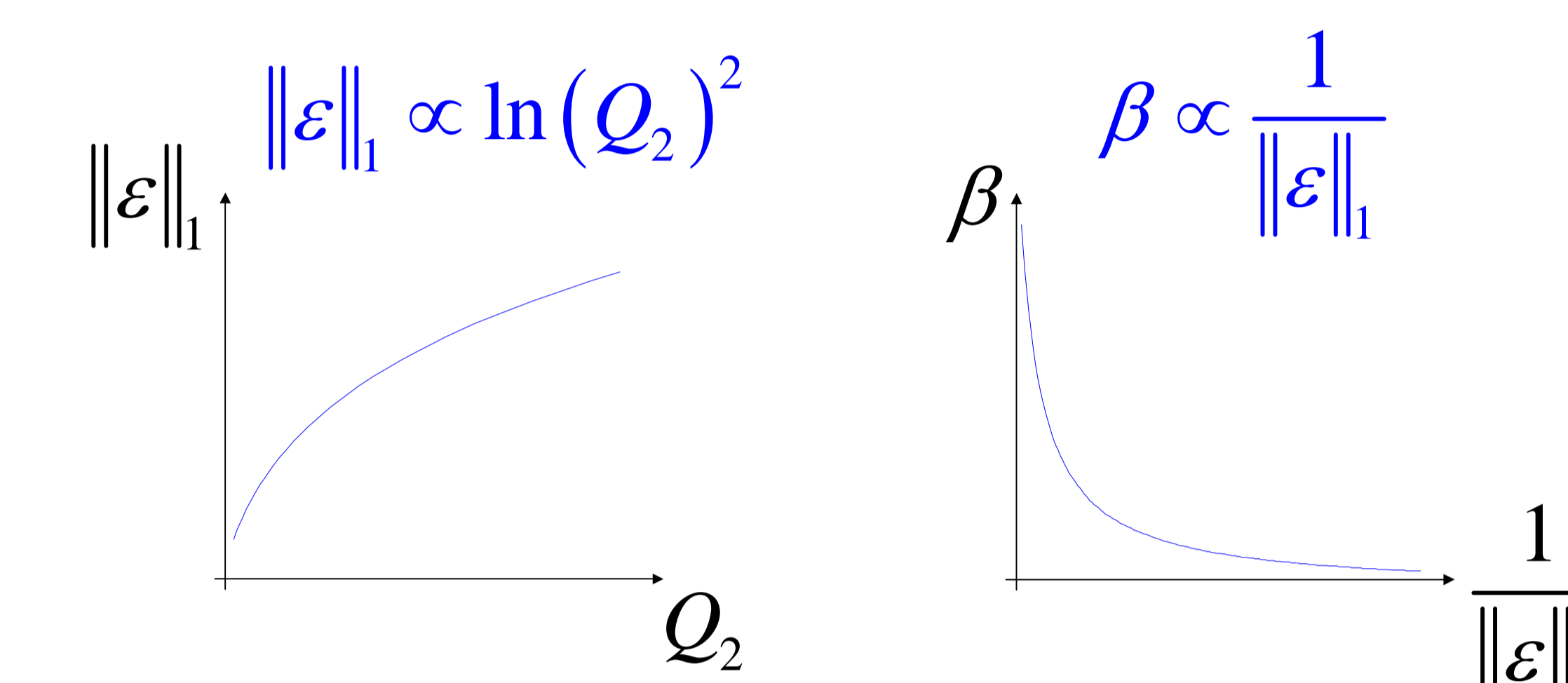


Correction signal modeling (III) Estimation of Γ distribution parameter

Problems:

1. C is not evaluated → online β estimation required
2. Great variability in the β-Q₂ relation over different data groups complicates its modeling.

Solution: **Decompose β-Q₂ model**



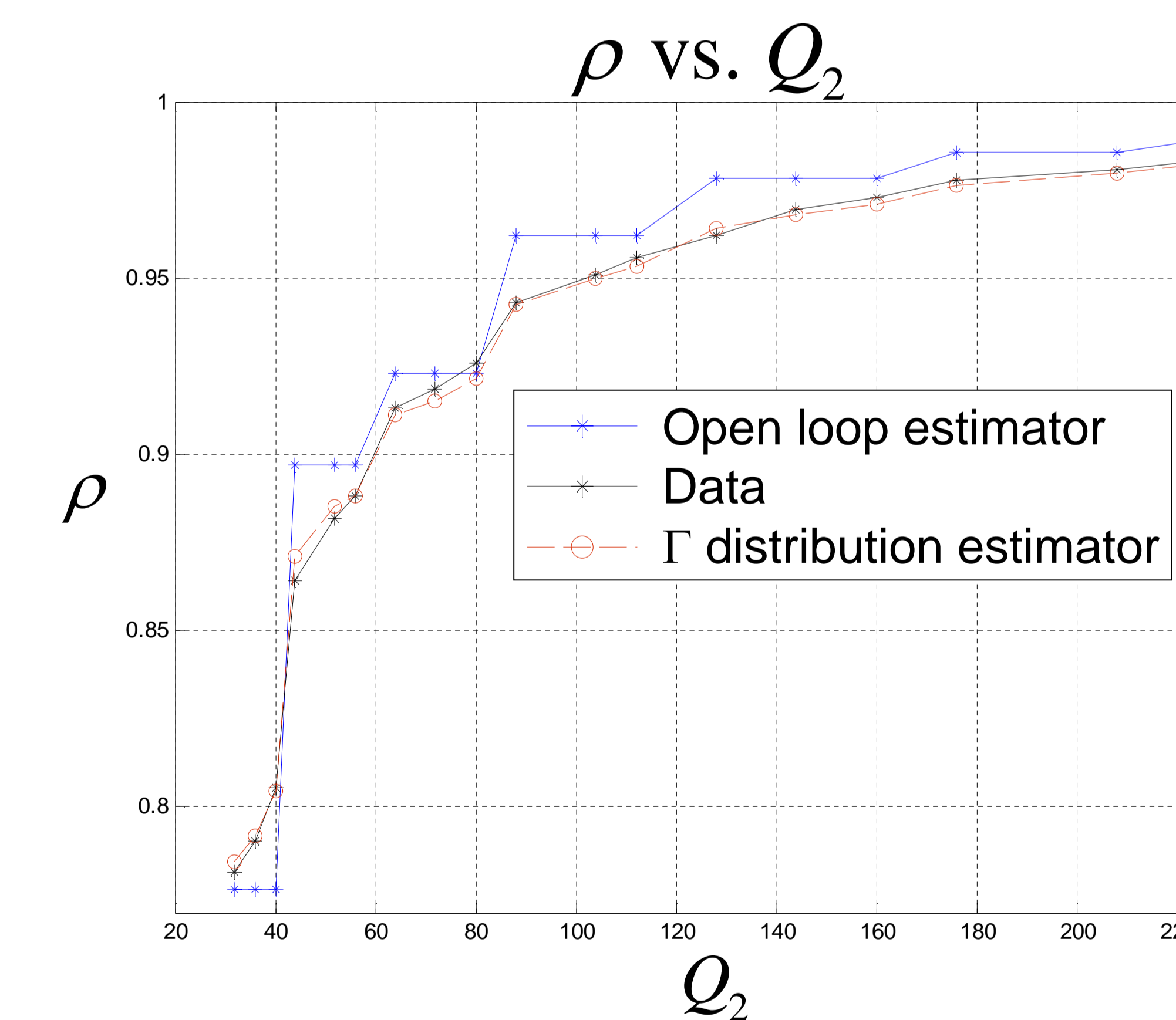
Parameters depend on input: Q₁, ||Y||₂

Parameter changes with prediction mode

Results

ρ-Q₂ estimator performance

The proposed Γ distribution estimator has an average relative error of less than 1.7%



Rate-Q₂ estimator performance

ρ-Q ₂ estimator	Mean relative rate deviation [%]
Open loop estimator	10.8
Data	2.5
Γ distribution parameter	3.0

Conclusion

- Novel **statistical-based ρ-Q₂ model** for transrating of H.264 intra-coded frames.
- The model provides **average rate deviation of 3%**, as compared to 10.8% average deviation, obtained using an open-loop estimator.