

Robust Coding Schemes for Distributed Sensor Networks with Unreliable Sensors

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Abstract — We consider a distributed sensor network in which several observations are communicated to the fusion center using limited transmission rate. The observations must be separately coded. We introduce a class of robust distributed coding schemes which flexibly trade off between system robustness and compression efficiency.

I. INTRODUCTION

Consider the distributed sensor network shown in Fig. 1. $\{X(t)\}_{t=1}^{\infty}$ is the target data sequence which the fusion center tries to recover. This data sequence may not be observed directly. Corrupted versions of $\{X(t)\}_{t=1}^{\infty}$ are separately coded by 2 sensors. The data rate at which Sensor i ($i = 1, 2$) may transmit information about its observations is limited to R_i bits per second. The sensors are not permitted to communicate with each other; i.e., Sensor i has to send data based solely on its own noisy observations $\{Y_i(t)\}_{t=1}^{\infty}$.

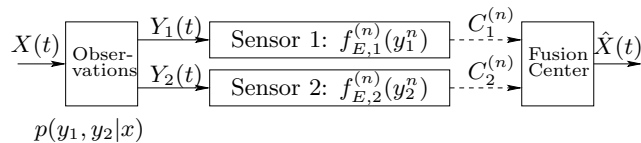


Figure 1: Model of distributed sensor network with unreliable sensors

Let $\{X(t), Y_1(t), Y_2(t)\}_{t=1}^{\infty}$ be temporally memoryless source with instantaneous joint probability distribution $P(x, y_1, y_2)$ on $\mathcal{X} \times \mathcal{Y}_1 \times \mathcal{Y}_2$, where \mathcal{X} is the common alphabet of the random variables $X(t)$ for $t = 1, 2, \dots$, \mathcal{Y}_i ($i = 1, 2$) is the common alphabet of the random variables $Y_i(t)$ for $t = 1, 2, \dots$. If Sensor i is able to function, then it encodes a block $y_i^n = [y_i(1), \dots, y_i(n)]$ of length n from its observed data using a source code $c_i^{(n)} = f_{E,i}^{(n)}(y_i^n)$ of rate $\frac{1}{n} \log |\mathcal{C}_i^{(n)}|$. If the fusion center only receives the data from Sensor i , then it tries to recover the target sequence $x^n = [x(1), \dots, x(n)]$ by implementing a mapping $f_{D,i} : \mathcal{C}_i \rightarrow \mathcal{X}^n$ ($i = 1, 2$). If the fusion center receives the data from both sensors, then it tries to recover the target sequence by implementing a mapping $f_{D,3} : \mathcal{C}_1^{(n)} \times \mathcal{C}_2^{(n)} \rightarrow \mathcal{X}^n$.

This model subsumes the multiple description problem [1] and the CEO problem [2], and was first studied in [3].

Definition 1 The quintuple $(R_1, R_2, D_1, D_2, D_3)$ is called achievable, if $\forall \varepsilon > 0$, $\exists n_0$ such that $\forall n > n_0$ there exist encoding functions:

$$f_{E,i}^{(n)} : \mathcal{Y}_i^{(n)} \rightarrow \mathcal{C}_i^{(n)} \quad \log |\mathcal{C}_i^{(n)}| \leq n(R_i + \varepsilon) \quad i = 1, 2$$

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and decoding functions:

$$f_{D,i} : \mathcal{C}_i^{(n)} \rightarrow \mathcal{X}^n \quad i = 1, 2$$

$$f_{D,3} : \mathcal{C}_1^{(n)} \times \mathcal{C}_2^{(n)} \rightarrow \mathcal{X}^n$$

such that for $\hat{X}_i^n = f_{D,i}(f_{E,i}(Y_i^n))$ ($i = 1, 2$), and for $\hat{X}_3^n = f_{D,3}(f_{E,1}(Y_1^n), f_{E,2}(Y_2^n))$,

$$\frac{1}{n} E \sum_{t=1}^n d(X(t), \hat{X}_i(t)) < D_i + \varepsilon \quad i = 1, 2, 3.$$

Here $d(\cdot, \cdot) : \mathcal{X} \times \mathcal{X} \rightarrow [0, d_{max}]$ is a given distortion measure.

Let \mathcal{Q} denote the set of all achievable quintuples.

II. MAIN RESULTS

Theorem 1 $(R_1, R_2, D_1, D_2, D_3)$ is achievable, if there exist random variables $(W_{1,1}, W_{1,2}, W_{2,1}, W_{2,2})$ jointly distributed with the generic source variables (X, Y_1, Y_2) such that the following properties are satisfied:

$$(i) \quad \begin{aligned} & (W_{1,1}, W_{1,2}) \rightarrow Y_1 \rightarrow (X, Y_2, W_{2,1}, W_{2,2}), \\ & (W_{2,1}, W_{2,2}) \rightarrow Y_2 \rightarrow (X, Y_1, W_{1,1}, W_{1,2}); \end{aligned}$$

(ii)

$$\begin{aligned} R_1 & \geq I(Y_1; W_{1,1}) + I(Y_1; W_{1,2} | W_{1,1}, W_{2,1}, W_{2,2}), \\ R_2 & \geq I(Y_2; W_{2,1}) + I(Y_2; W_{2,2} | W_{1,1}, W_{2,1}, W_{1,2}), \\ R_1 + R_2 & \geq I(Y_1; W_{1,1}) + I(Y_2; W_{2,1}) \\ & \quad + I(Y_1, Y_2; W_{1,2}, W_{2,2} | W_{1,1}, W_{2,1}); \end{aligned}$$

(iii) There exist functions:

$$f_i : \mathcal{W}_{i,1} \rightarrow \mathcal{X} \quad i = 1, 2,$$

$$f_3 : \mathcal{W}_{1,1} \times \mathcal{W}_{1,2} \times \mathcal{W}_{2,1} \times \mathcal{W}_{2,2} \rightarrow \mathcal{X},$$

such that $E d(X, \hat{X}_i) \leq D_i$ ($i = 1, 2, 3$), where $\hat{X}_1 = f_1(W_{1,1})$, $\hat{X}_2 = f_2(W_{2,1})$ and $\hat{X}_3 = f_3(W_{1,1}, W_{1,2}, W_{2,1}, W_{2,2})$.

If \mathcal{C} denotes the set of these achievable quintuples, then time sharing yields that $\text{conv}(\mathcal{C})$ is also an achievable region.

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