

# Statistical Signal Processing

## Exercise 2

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### 1 KL Divergence

Wiki: In probability theory and information theory, the **Kullback-Leibler divergence** is a non-symmetric measure of the difference between two probability distribution  $P$  and  $Q$ . Specifically, the KL-divergence of  $Q$  from  $P$  is a measure of the information lost when  $Q$  is used to approximate  $P$ : The KL divergence measure the expected number of extra bits required to code [Huffman code] samples from  $P$  when using a code based on  $Q$ , rather than using the the true code based on  $P$ . Typically  $P$  represents the "true" distribution of data, observations, or a precisely calculated theoretical distribution. The measure  $Q$  typically represents a theory, mode, description or approximation of  $P$ .

In this question we're finding the KL divergence of  $f_2$  from  $f_1$ .

Using the Law of the Unconscious Statistician:

$$\mathcal{D}_{KL}(f_1||f_2) = \mathbb{E}_{f_1} \left[ \log \left( \frac{f_1(y)}{f_2(y)} \right) \right] = \int_{-\infty}^{\infty} \log \left( \frac{f_1(y)}{f_2(y)} \right) f_1(y) dy = \int_{-\infty}^{\infty} -\log \left( \frac{f_2(y)}{f_1(y)} \right) f_1(y) dy$$

Known inequality:  $\forall x \geq 0 : \log(x) \leq x - 1$