

## 1 Intensity-level entropy

Given a discrete random variable  $X$  with support  $\mathcal{X}$ , the *Shannon entropy* is

$$H = \sum_{x \in \mathcal{X}} -P(x) \ln P(x).$$

The *intensity-level entropy* is the Shannon entropy of the empirical distribution of intensity values.

```
1 import numpy as np
2
3 def shannon_entropy(h):
4     """The Shannon entropy in bits"""
5     return -sum(p*np.log2(p) if p > 0 else 0 for p in h)
6
7 def intensity_distribution(data, upper=256):
8     """The intensity distribution of 8-bit `data`. """
9     hist, _ = np.histogram(data, bins=range(upper+1), density=True)
10    return hist
11
12 def intensity_entropy(data, upper=256):
13    """The intensity-level entropy of 8-bit image data"""
14    return shannon_entropy(intensity_distribution(data, upper))
15
16 def intensity_expected(f, data):
17    """The intensity-distribution expected value of `f`. """
18    return sum(p*f(p) for p in intensity_distribution(data))
```