INVESTIGATIONS IN RELATIVE ENTROPY

1. Showing that one of the terms in $\partial_i I$ is zero

We are interested in a relative entropy I, which depends on the values f_i of the tuning curve at various points. We have that

 $I = E_{\mathbf{r}|\mathbf{s}}(\ln(S)),$

where

$$S(\mathbf{r}) = \sum_{k=0}^{M-1} Q_s^k(\mathbf{r})$$

and

$$Q_s^k(\mathbf{r}) = \prod_{j=0}^{M-1} \left(\frac{f_{j-k}(s)}{f_j(s)} \right)^{r_j} = \prod_{j=0}^{M-1} f_j(s)^{r_{j+k}-r_j}.$$

From now on we suppress the value of s (might as well be s = 0, by rotational symmetry) and write $f_i := f_i(s)$. The expectation is taken relative to the Poisson probability distribution

$$P(\mathbf{r}) = \prod_{i} \frac{f_i^{r_i}}{r_i!} e^{-f_i}.$$

That is, our relative entropy is

$$I = \sum_{\mathbf{r}} P(\mathbf{r}) \ln(S(\mathbf{r})).$$

We then have

$$\partial_i I = \sum_{\mathbf{r}} P(\mathbf{r}) \frac{\partial_i S}{S} + \sum_{\mathbf{r}} \partial_i P(\mathbf{r}) \ln(S).$$

In this section we show that the first term (or rather, f_i times the first term) is zero. We compute:

$$f_i \partial_i S = \sum_{k=0}^{M-1} (r_{i+k} - r_i) \prod_j f_j^{r_{j+k} - r_j},$$

SO

$$\frac{f_i \partial_i S}{S} = \frac{\sum_{k=0}^{M-1} (r_{i+k} - r_i) \prod_j f_j^{r_{j+k}}}{\sum_{k=0}^{M-1} \prod_j f_j^{r_{j+k}}},$$

where we have multiplied the numerator and denominator by $\prod_{j} f_{j}^{r_{j}}$. The extra factor of f_{i} is just a constant, and doesn't affect whether the total is zero or not. Now we take an

expectation. The probability of any particular value of \mathbf{r} is proportional to $\frac{\prod_j f_j^{r_j}}{\prod_j r_j!}$, so our expectation is proportional to

$$\sum_{\mathbf{r}} \frac{\sum_{k} (f_{i+k} - f_i) \prod_{j} f_j^{r_j + r_{j+k}}}{\sum_{k} \prod_{j} f_j^{r_j + k} r_j!}.$$

The denominator is invariant under cyclic permutations of the entries of \mathbf{r} , so we can average the numerator over those cyclic permutations, getting

$$\frac{1}{M} \sum_{k=0}^{M-1} \sum_{\ell=0}^{M-1} (r_{i+\ell+k} - r_{i+\ell}) \prod_{j} f_j^{r_{j+\ell} + r_{j+\ell+k}}.$$

Every term in this expansion appears exactly twice, with opposite signs, so the sum is zero. For instance, the term that contains a factor of $f_i^{r_i+r_{m+i}}$ comes once from $\ell=0$ and k=m, with coefficient $r_{i+m}-r_i$, and once from $\ell=m$ and k=-m, with coefficient r_i-r_{i+m} .

2. Computing the Hessian

We have shown that

$$\partial_i I = \sum (\partial_i P) \ln(S).$$

This implies that

$$\partial_{ij}^2 I = \sum (\partial_{ij}^2 P) \ln(S) + \sum (\partial_i P) \frac{\partial_j S}{S}.$$

We will investigate these terms one at a time.

The Poisson law has the convenient property that $\partial_i P(\mathbf{r}) = \left(\frac{r_i}{f_i} - 1\right) P(\mathbf{r}) = P(\mathbf{r} - e_i) - P(\mathbf{r})$, where e_i is a unit vector in the *i*th direction. Similarly,

$$\partial_{ij}^2 P(\mathbf{r}) = P(\mathbf{r} - e_i - e_j) - P(\mathbf{r} - e_i) - P(\mathbf{r} - e_j) + P(\mathbf{r}),$$

SO

$$\sum_{\mathbf{r}} \partial_{ij}^{2} P(\mathbf{r}) \ln(S) = \sum_{\mathbf{r}} [P(\mathbf{r} - e_{i} - e_{j}) - P(\mathbf{r} - e_{i}) - P(\mathbf{r} - e_{j}) + P(\mathbf{r})] \ln(S(\mathbf{r}))$$

$$= \sum_{\mathbf{r}} P(\mathbf{r}) [\ln(S(\mathbf{r})) - \ln(S(\mathbf{r} + e_{i})) - \ln(S(\mathbf{r} + e_{j})) + \ln(S(\mathbf{r} + e_{i} + e_{j}))].$$

This term can be evaluated numerically by Monte Carlo.

Unfortunately, the second term is not zero. Instead, we have

$$f_{i}f_{j}(\partial_{i}P)(\partial_{j}S)/S = \frac{P(\mathbf{r})(r_{i}-f_{i})\sum_{k}(r_{j+k}-r_{j})\prod_{\ell}f_{\ell}^{r_{\ell+k}}}{\sum_{k}\prod_{\ell}f_{\ell}^{r_{\ell+k}}}$$
$$= \frac{\sum_{k}r_{i}(r_{j+k}-r_{j})\prod_{\ell}f_{\ell}^{r_{\ell+k}+r_{\ell}}}{\sum_{k}\prod_{\ell}f_{\ell}^{r_{k+\ell}}e^{f_{\ell}}r_{\ell}!},$$

where we used the results of the previous section to eliminate $f_i(r_j - r_{j+k})$. Unfortunately, the cyclic permutation trick of the previous section does not cancel $r_i(r_{j+k} - r_j)$ against

 $r_i(r_j - r_{j+k})$. Rather, we wind up averaging $r_i(r_{j+k} - r_j)$ and $r_{i+k}(r_j - r_{j+k})$ to get an expression that is manifestly symmetric in i and j, namely

$$\sum_{\mathbf{r}} 2f_i f_j(\partial_i P)(\partial_j S)/S = \sum_{\mathbf{r}} \frac{-\sum_k (r_i - r_{i+k})(r_j - r_{j+k}) \prod_\ell f_\ell^{r_{\ell+k} + r_\ell}}{\sum_k \prod_\ell f_\ell^{r_{k+\ell}} e^{f_\ell} r_\ell!}.$$

This can also be rewritten as the expectation of

$$\frac{-\sum_{k} (r_{i} - r_{i+k})(r_{j} - r_{j+k}) \prod_{\ell} f_{\ell}^{r_{\ell+k}}}{\sum_{k} \prod_{\ell} f_{\ell}^{r_{k+\ell}}}.$$

I don't see any way to simplify this. However, it doesn't look like it should be too hard to sample numerically.

I'll play around with some simple cases (M = 2 or 3) to see if I'm overlooking any simplification, but I doubt it.