CSE 527 Lecture 10

More on the Gibbs Sampler

AlignAce (Roth, et al. 1998)

- Lawrence et al.: protein motifs
- Roth et al.: DNA regulatory motifs
- Differences:
 - Genomic background model,
 e.g. yeast Saccharomyces cerevisiae is 62% A-T
 - both strands used
 - overlapping sites prohibited
 - Multiple motifs: find best & mask
 - "MAP" scoring; "specificity" scoring

Projects – see web

- Implementation or literature review
- Small (interdisciplinary) groups preferred
- Suggestion:
 - make a schedule
 - bite-size-pieces
- Some ideas on web/by email & I'm happy to talk/listen/give (bad?) advice - send email

Rocke & Tompa (Recomb '98)

- Gibbs, adapted for gapped motifs
- single "genomic" DNA sequence

Why Gaps

- Biology often tolerates diversity
- 2 similar TFs bind 2 similar sites
- Same TF binds 2 sites (perhaps one better than the other)
- Dimeric TFs often "don't care" in middle & flexible
- TF and/or DNA may twist/bulge

A Gapped Motif

```
O TAT < CCCCCCTCA C CTTCG G CAGCTCCCCCATAA

1 ATC < CCCCCCTCA C TTCG G CAGCTCCCCCCATAA

2 GTA < CCCCCTCAGTCACTTCGCG CAGCTCCCCCCATAA

3 AAT < CCCCCCTCAGTC TTCGCG CAGCTCCCCC TAA
```

Why gaps are hard

- Alignment
 - Pairwise -- O(n²)
 Multiple -- O(n^k)
 - Gibbs/MEME/... require many alignments
- Scoring

R/T Approach - Scores

- WMM
- Relative entropy, aka expected LLR
- Score gaps like background, "minus a small penalty"

R/T Approach - Alignment

- Gibbs replaces 1 string per iteration
- Use pairwise alignment between new string and previously computed alignment of remaining k-1
- Actually align motif against whole genome -Time O(genome length x motif width)

R/T Approach-"Gibbs"

- discard 0-2 random strings at each iteration
- pick replacement greedily, not by sampling; avoid local max by random restarts (see Rocke's thesis for more on this)

Test Data

- Haemophilus influenzae
- ~1.8 megabases
- Delete all protein-coding, leaves ~ 350 kb
- Concatenate, separated with markers
- Plus reverse complement, total ~ 700 kb

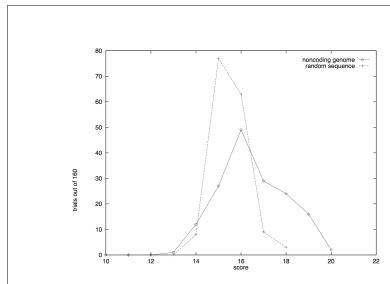


Figure 2: 160 trials of the basic algorithm on the noncoding genome vs. a random sequence

Motif width=10

A Motif + Context

```
< CGCCCTTTCA >
                           at position 118666.
      < CGCCCTTTCA >
                           at position 642660.
2 AAT < CGCCTTTTCA > AAA
                           at position 425287.
3 ATC < CGCCC-TTCA > TGA
                           at position 330462.
4 TTG < CGCCC-TTCA > CTA
                           at position 558509.
5 AAC < CGCCCATTCA > ATC
                           at position 237890.
      < CGCCC-TTCA > CGT
                           at position 495353.
7 TCT < CGCCTTTTCA > TTG
                           at position 34553.
      < CGCCCTTTCA >
                           at position 677174.
      < CGCCC-TTCA > GGG
                           at position 222102.
```

Figure 1: A sample motif (score 16.6) produced by the basic algorit

Rewindowing

- After convergence, "rewindow" -- choose subset of rows and adjust left/right boundaries to maximize score.
- NP-hard? Use another greedy heuristic

Rewindowing

```
O GGA <
           CGCCCTTTCA
                          > CGG
                                  at position 118663.
1 GGA <
           CGCCCTTTCA
                          > CGG
                                  at position 642657.
2 GCT <
           CGCCC-TTCAGGG > TTC
                                  at position 222099.
3 GGA <
           CGCCCTTTCA
                          > CGG
                                  at position 677171.
           CGCCC-TTCACGT > AAT
4 \text{ AAA} <
                                  at position 495350.
```

Figure 3: The motif of Figure 1 after rewindowing (score 20.8)

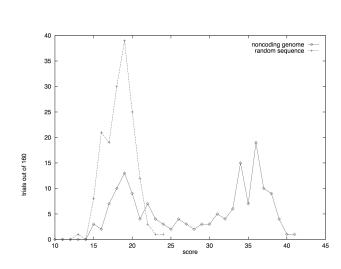


Figure 4: 160 trials of the two-phase algorithm on the noncoding genome vs. a random sequence

A closer look at 35

- 6 almost perfectly identical regions of 5.3 kb, each 3 rRNA genes plus some tRNA genes
- 9% of genome but 50% of high-scoring motifs
- removed 80kb containing them & re-ran

Figure 5: 160 trials of the two-phase algorithm on the noncoding genome with long repeats removed vs. a random sequence

genome w/ long repeats removed

After Removal

```
O TCG < GCAGCTCCCCCCATAAATGG > GTG at position 449120.

1 TCG < GCAGCTCCCCCCATAAATGG > GTG at position 448927.

2 GCG < ACAGCTCCCCCCATAAATGG > GTG at position 232857.

3 GCG < CCAGCTCCC-CCGTAAACGG > GTG at position 88280.
```

Figure 6: A sample motif (score 25) produced by two phases

More rewindowing

```
O TCG < GCAGCTCCCCCCATAAATGG > GTG at position 449120.

1 TCG < GCAGCTCCCCCCATAAATGG > GTG at position 448927.

2 GCG < ACAGCTCCCCCCATAAATGG > GTG at position 232857.

3 GCG < CCAGCTCCC TAAACGG > GTG at position 88280.

O TAT < CCCCCCTCA--C-CTTCG-G-CAGCTCCCCCCATAAATGGGTGGAGCCAAGAT > TAG at position 449105.

1 ATC < CCCCCCTCA--C-TTCG-G-CAGCTCCCCCCATAAATGGGTGGAGCCAAGAT > TAG at position 44913.

2 GTA < TCCCCCTCAGTCACTCGCGCACACCTCCCCCATAAATGGGTGGAGCCAAGGT > AAT at position 232837.

3 AAT < CCCCCCTCAGTC--TTCGCGCCAGCTCCCC TAAACGGGTGGAGCCAAGGG > ATC at position 88262.
```

Figure 7: The motif of Figure 6 after seven phases (score 62)

0 & I identical for another 55 bases;5 differences in next 44.Probably not a TFBS, but not "random"

Summary

- handles gaps
- greedy "sampling" / random restarts
- avoids full multiple alignment by exploiting good partial alignment
- validation null model for comparison
- look at data -
 - rewindowing
 - rRNA cluster