

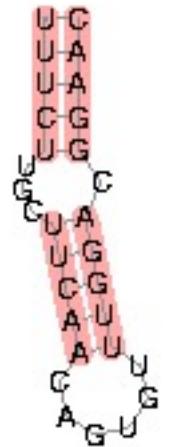
Faster Genome Annotation of Non-coding RNAs Without Loss of Accuracy

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Recomb '04

Rfam

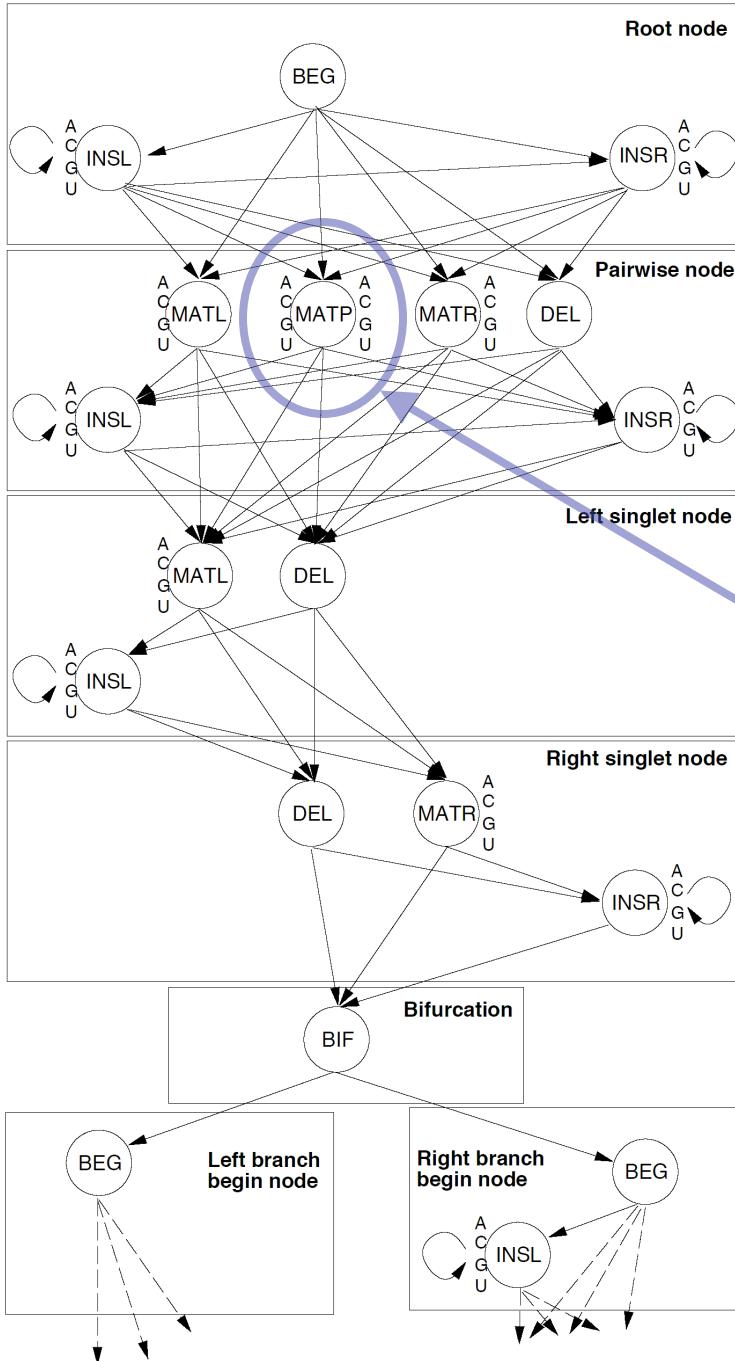


- Input (hand-tuned):
 - MSA
 - SS_cons
 - Score Thresh T
 - Window Len W
- Output:
 - CM
 - scan results

IRE (partial seed alignment):

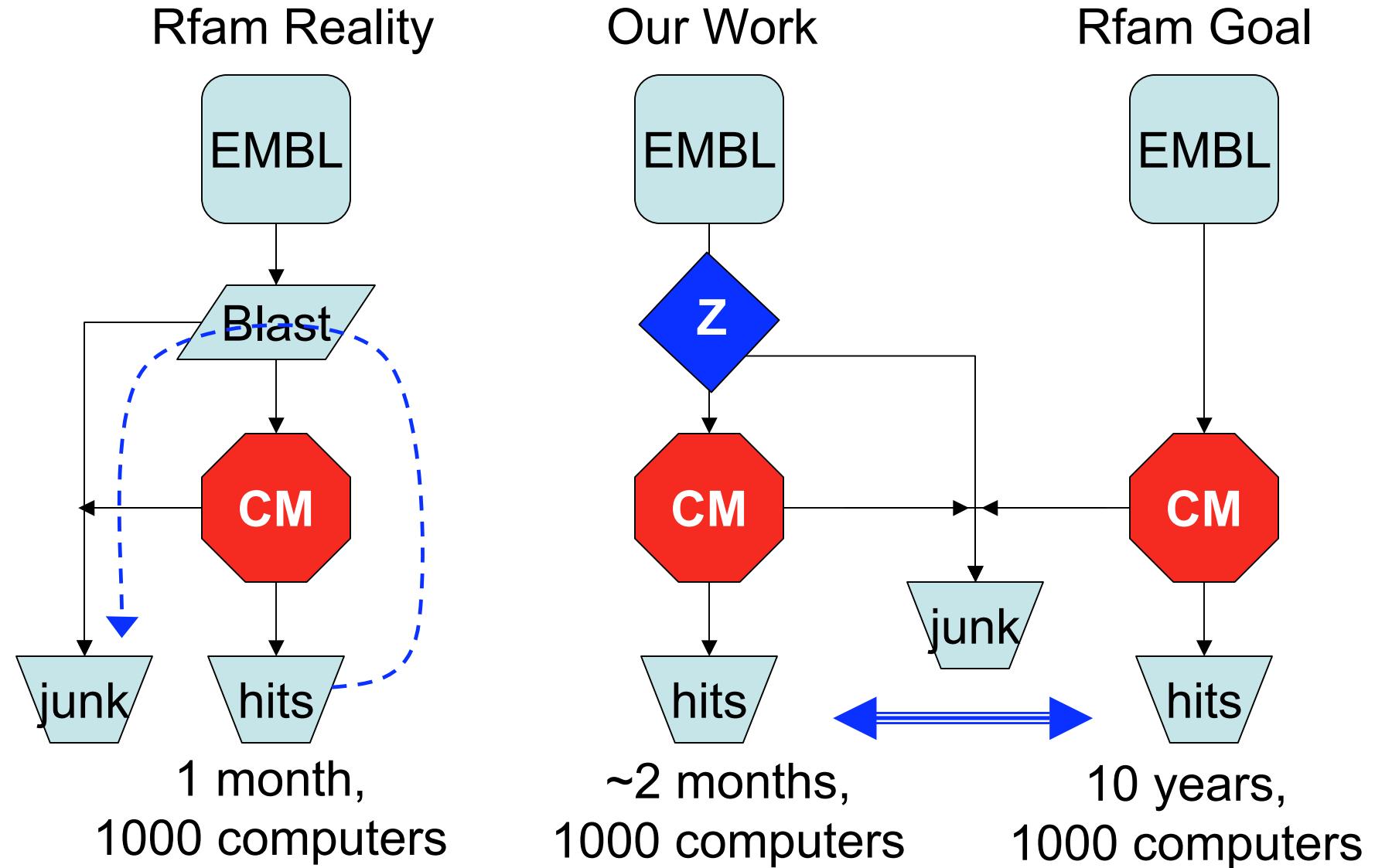
Hom. sap.	GUUCCUGCUUCAACAGUGUUUGGAUGGAAC
Hom. sap.	UUUCUUC. UUCAACAGUGUUUGGAUGGAAC
Hom. sap.	UUUCCUGUUUCAACAGUGCUUGGA. GGAAC
Hom. sap.	UUUAUC. . AGUGACAGAGUUCACU. AUAAA
Hom. sap.	UCUCUUGCUUCAACAGUGUUUGGAUGGAAC
Hom. sap.	AUUAUC. . GGGAACAGUGUUUCCC. AUAAU
Hom. sap.	UCUUGC. . UUCAACAGUGUUUGGACGGAAAG
Hom. sap.	UGUAUC. . GGAGACAGUGAUCUCC. AUAUG
Hom. sap.	AUUAUC. . GGAAGCAGUGCCUUCC. AUAAU
Cav. por.	UCUCCUGCUUCAACAGUGCUUGGACGGAGC
Mus. mus.	UAUAUC. . GGAGACAGUGAUCUCC. AUAUG
Mus. mus.	UUUCCUGCUUCAACAGUGCUUGAACCGGAAC
Mus. mus.	GUACUUGCUUCAACAGUGUUUGAACCGGAAC
Rat. nor.	UAUAUC. . GGAGACAGUGACCUC. AUAUG
Rat. nor.	UAUCUUGCUUCAACAGUGUUUGGACGGAAC
SS_cons	<<<<...<<<<.....>>>>. >>>>

Covariance Model



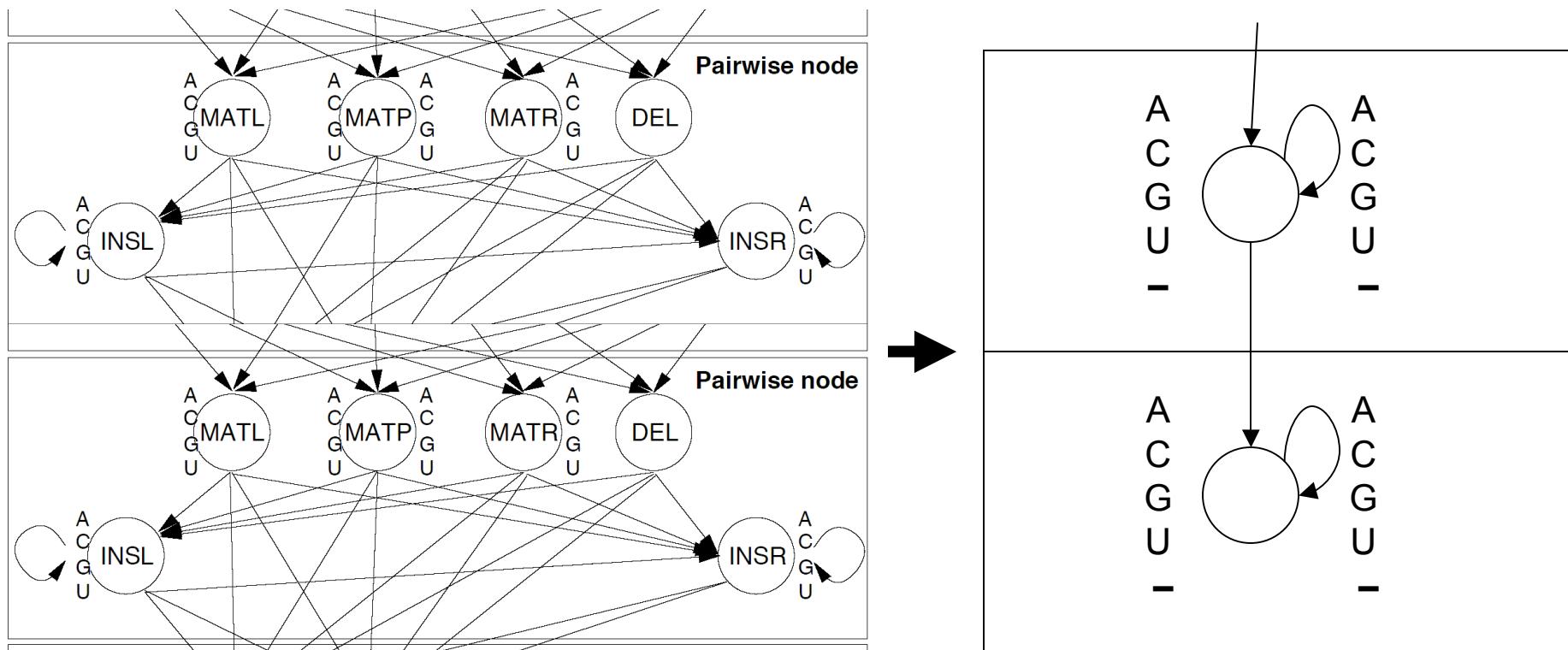
Key difference of CM vs HMM:
Pair states emit paired symbols,
corresponding to base-paired
nucleotides; 16 emission
probabilities here.

CM's are good, but slow

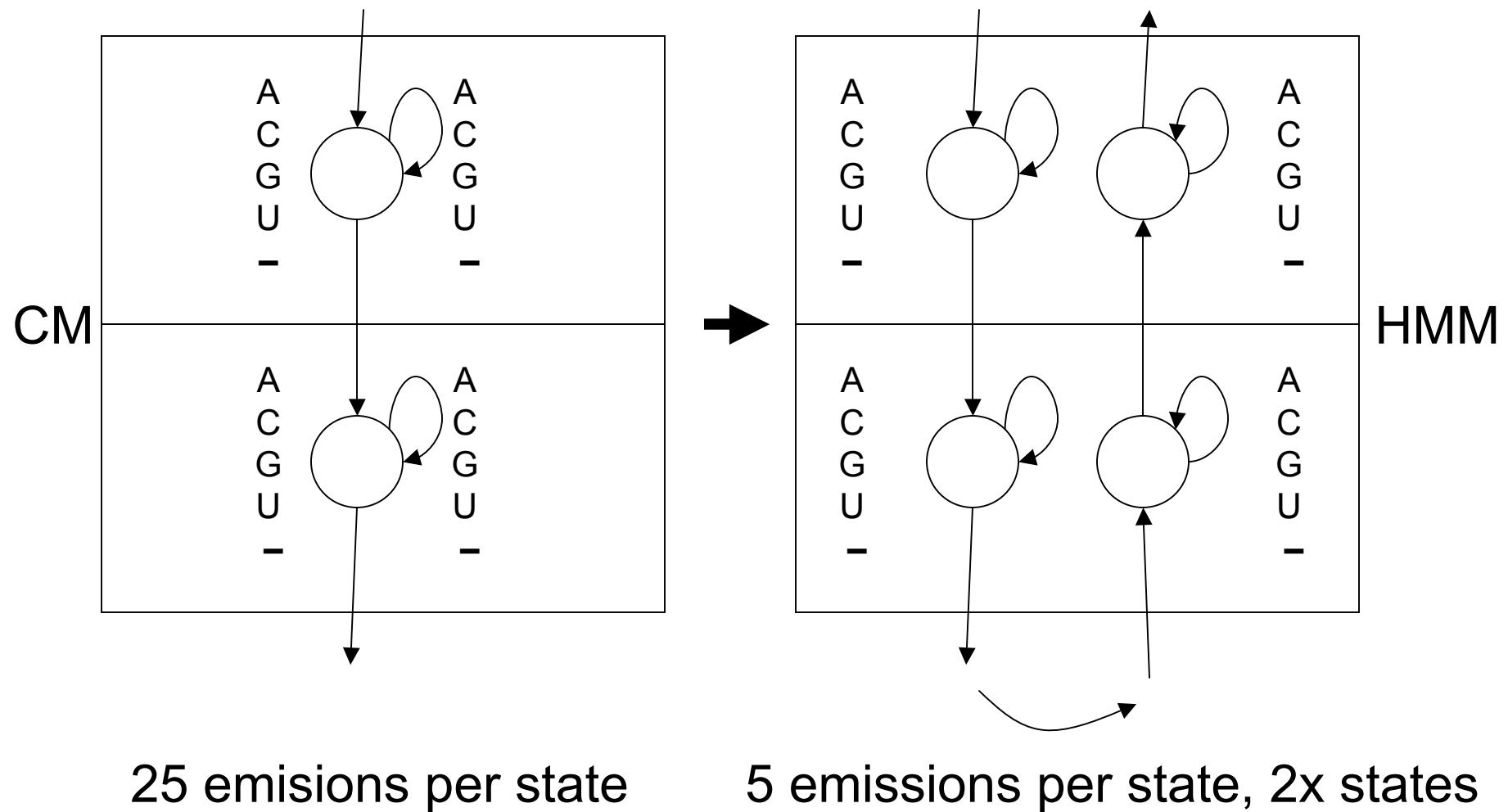


Oversimplified CM

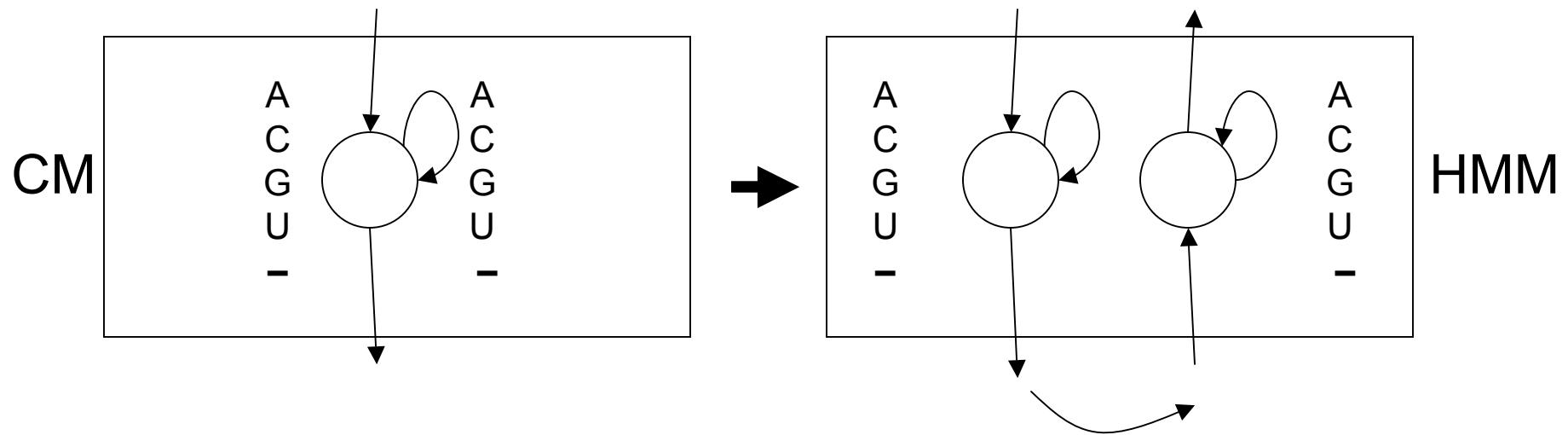
(for pedagogical purposes only)



CM to HMM



Key Issue: 25 scores → 10

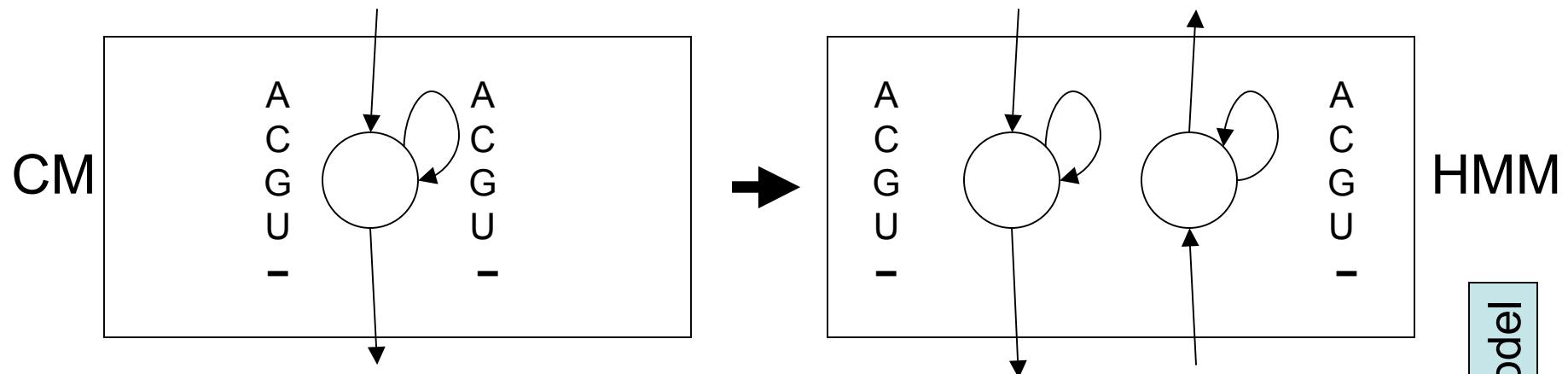


- Need: $\log \text{Viterbi scores } \text{CM} \leq \text{HMM}$

Viterbi/Forward Scoring

- Path π defines transitions/emissions
- $\text{Score}(\pi) = \text{product of “probabilities” on } \pi$
- NB: ok if “probs” aren’t, e.g. $\sum \neq 1$
- E.g. in CM, emissions are odds ratios vs 0th-order background
- For any nucleotide sequence x :
 - Viterbi-score(x) = $\max\{ \text{score}(\pi) \mid \pi \text{ emits } x\}$
 - Forward-score(x) = $\sum\{ \text{score}(\pi) \mid \pi \text{ emits } x\}$

Key Issue: 25 scores → 10



- Need: log Viterbi scores CM ≤ HMM

$$P_{AA} \leq L_A + R_A$$

$$P_{AC} \leq L_A + R_C$$

$$P_{AG} \leq L_A + R_G$$

$$P_{AU} \leq L_A + R_U$$

$$P_{A-} \leq L_A + R_-$$

$$P_{CA} \leq L_C + R_A \quad \dots$$

$$P_{CC} \leq L_C + R_C \quad \dots$$

$$P_{CG} \leq L_C + R_G \quad \dots$$

$$P_{CU} \leq L_C + R_U \quad \dots$$

$$P_{C-} \leq L_C + R_- \quad \dots$$

NB:HMM not a prob. model

Rigorous Filtering

$$\begin{aligned} P_{AA} &\leq L_A + R_A \\ P_{AC} &\leq L_A + R_C \\ P_{AG} &\leq L_A + R_G \\ P_{AU} &\leq L_A + R_U \\ P_{A-} &\leq L_A + R_- \\ \dots \end{aligned}$$

- Any scores satisfying the linear inequalities give rigorous filtering

Proof:

CM Viterbi path score
≤ “corresponding” HMM path score
≤ Viterbi HMM path score
(even if it does not correspond to *any* CM path)

Some scores filter better

$$P_{UA} = 1 \leq L_U + R_A$$

$$P_{UG} = 4 \leq L_U + R_G$$

Option 1:

$$L_U = R_A = R_G = 2$$

Option 2:

$$L_U = 0, R_A = 1, R_G = 4$$

Assuming ACGU ≈ 25%

Opt 1:

$$L_U + (R_A + R_G)/2 = 4$$

Opt 2:

$$L_U + (R_A + R_G)/2 = 2.5$$

Optimizing filtering

- For any nucleotide sequence x :
 $\text{Viterbi-score}(x) = \max\{ \text{score}(\pi) \mid \pi \text{ emits } x \}$
 $\text{Forward-score}(x) = \sum\{ \text{score}(\pi) \mid \pi \text{ emits } x \}$
- Expected Forward Score
 $E(L_i, R_i) = \sum_x \text{Forward-score}(x) * \Pr(x)$
 - NB: E is a function of L_i, R_i only
- Optimization:
Minimize $E(L_i, R_i)$ subject to score L.I.s
 - This is heuristic (“forward $\downarrow \Rightarrow$ Viterbi $\downarrow \Rightarrow$ filter \downarrow ”)
 - But still rigorous because “subject to score L.I.s”

Under 0th-order background model

Calculating $E(L_i, R_i)$

$$E(L_i, R_i) = \sum_x \text{Forward-score}(x) * \text{Pr}(x)$$

- Forward-like: for every state, calculate expected score for all paths ending there, easily calculated from expected scores of predecessors & transition/ emission probabilities/scores

Minimizing $E(L_i, R_i)$

- Calculate $E(L_i, R_i)$ *symbolically*, in terms of emission scores, so we can do partial derivatives for numerical convex optimization algorithm

$$\frac{\partial E(L_1, L_2, \dots)}{\partial L_i}$$

Estimated Filtering Efficiency

(139 Rfam 4.0 families)

Filtering fraction	# families (compact)	# families (expanded)
$< 10^{-4}$	105	110
$10^{-4} - 10^{-2}$	8	17
.01 - .10	11	3
.10 - .25	2	2
.25 - .99	6	4
.99 - 1.0	7	3

Results: buried treasures

Name	# found BLAST + CM	# found rigorous filter + CM	# new
<i>Pyrococcus</i> snoRNA	57	180	123
Iron response element	201	322	121
Histone 3' element	1004	1106	102
Purine riboswitch	69	123	54
Retron msr	11	59	48
Hammerhead I	167	193	26
Hammerhead III	251	264	13
U4 snRNA	283	290	7
S-box	128	131	3
U6 snRNA	1462	1464	2
U5 snRNA	199	200	1
U7 snRNA	312	313	1

Results: With additional work

	# with BLAST+CM	# with rigorous filter series + CM	# new
Rfam tRNA	58609	63767	5158
Group II intron	5708	6039	331
tRNAscan- SE (human)	608	729	121
tmRNA	226	247	21
Lysine riboswitch	60	71	11
And more...			