



Sticky-favoring model over hidden state vocab {0,1}. Factor scores are “goodness points” are in log-scale additive form. (They’re positive, though for an HMM they would all be negative.)

$\log P(y | w) = (\text{constant}) + G(y_1, y_2, y_3)$
 $G(y_1, y_2, y_3) = A(y_1, y_2) + A(y_2, y_3) + B_1(y_1) + B_2(y_2) + B_3(y_3)$
inference goal: $\text{argmax}_y G(y)$

Additive Viterbi

For $t=1..T$,
 For k in $\{0,1\}$,

$$V_t[k] := \max_j (V_{t-1}[j] + A(j, k) + B_t(k))$$

$$B[k] := \text{arg max}_j (\dots)$$

For $t=1$, set $A_0(\text{anything})=0$ and $V_0[\text{anything}]=0$
 Final backtrace step: take best-scoring from last V_T , follow the backpointers all the way back

Run Viterbi and fill out the trellis with arcs like in the textbook’s HMM example.

Most probable sequence $y^* = (\quad , \quad , \quad)$ $G(y^*) =$

