

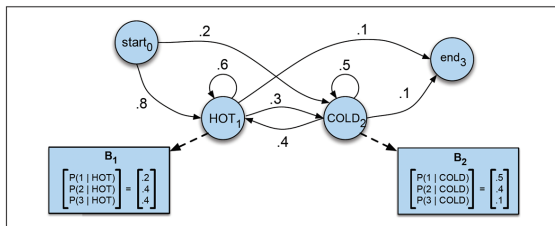
# Exercise: The Forward Algorithm & Viterbi Algorithm

Today's exercise is to:

1. Implement the Forward Algorithm for the Hidden Markov Model (shown on the next slide) to compute the probability of the observation sequence  $3\ 1\ 3$ .
2. Implement the Viterbi Algorithm to compute the most likely weather sequence for the observation sequence  $3\ 1\ 3$ .

Use the file `hmm_template.py` posted on blackboard, it contains the incomplete functions `compute_forward` and `compute_viterbi`.

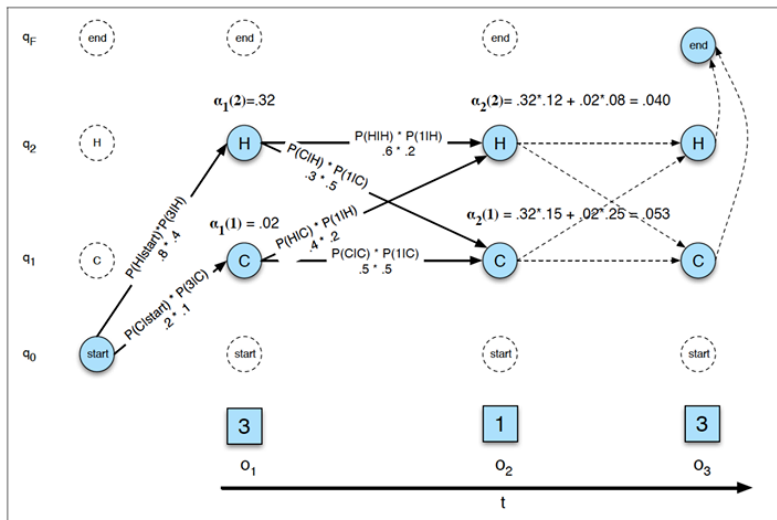
# The Hidden Markov Model Used in the Exercise



The above shows a Hidden Markov Model for relating numbers of ice creams eaten by Jason (the observations) to the weather (H or C, the hidden variables).

You can also check <https://web.stanford.edu/~jurafsky/slp3/9.pdf> for further details.

# Visual Representation of the Forward Algorithm.



# Pseudocode for the Forward Algorithm.

**function** FORWARD(*observations* of len  $T$ , *state-graph* of len  $N$ ) **returns** *forward-prob*

create a probability matrix *forward*[ $N+2, T$ ]

**for** each state  $s$  **from** 1 **to**  $N$  **do** ; initialization step

$forward[s, 1] \leftarrow a_{0,s} * b_s(o_1)$

**for** each time step  $t$  **from** 2 **to**  $T$  **do** ; recursion step

**for** each state  $s$  **from** 1 **to**  $N$  **do**

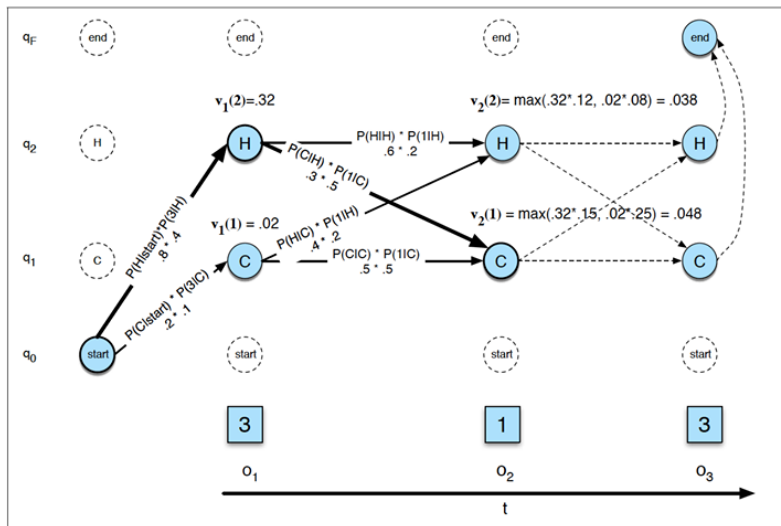
$$forward[s, t] \leftarrow \sum_{s'=1}^N forward[s', t-1] * a_{s',s} * b_s(o_t)$$

$forward[q_F, T] \leftarrow \sum_{s=1}^N forward[s, T] * a_{s,q_F}$  ; termination step

**return** *forward*[ $q_F, T$ ]

Note that in the code, the transition matrix corresponds to **a**, whereas the emissions matrix corresponds to **b**.

# Visual Representation of the Viterbi Algorithm.



# Pseudocode for the Viterbi Algorithm.

**function** VITERBI(*observations of len  $T$ , state-graph of len  $N$* ) **returns** *best-path*

create a path probability matrix  $viterbi[N+2, T]$

**for** each state  $s$  **from** 1 **to**  $N$  **do** ; initialization step

$viterbi[s, 1] \leftarrow a_{0,s} * b_s(o_1)$

$backpointer[s, 1] \leftarrow 0$

**for** each time step  $t$  **from** 2 **to**  $T$  **do** ; recursion step

**for** each state  $s$  **from** 1 **to**  $N$  **do**

$viterbi[s, t] \leftarrow \max_{s'=1}^N viterbi[s', t-1] * a_{s',s} * b_s(o_t)$

$backpointer[s, t] \leftarrow \operatorname{argmax}_{s'=1}^N viterbi[s', t-1] * a_{s',s}$

$viterbi[q_F, T] \leftarrow \max_{s=1}^N viterbi[s, T] * a_{s,q_F}$  ; termination step

$backpointer[q_F, T] \leftarrow \operatorname{argmax}_{s=1}^N viterbi[s, T] * a_{s,q_F}$  ; termination step

**return** the backtrace path by following backpointers to states back in time from  $backpointer[q_F, T]$

Note that in the code, the transition matrix corresponds to **a**, whereas the emissions matrix corresponds to **b**.

# Homework

Find the probability of the following observation sequences:

- ▶ 3, 3, 1, 1, 2, 2, 3, 1, 3.
- ▶ 3, 3, 1, 1, 2, 3, 3, 1, 2.

Also find the most likely weather sequences for the two observation sequences.