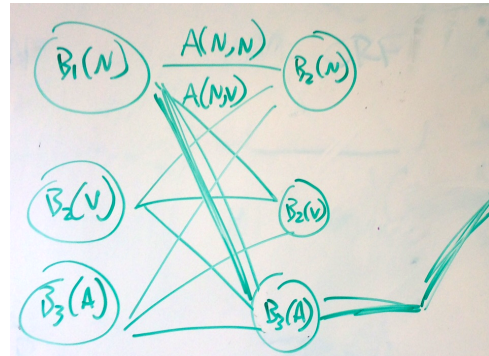
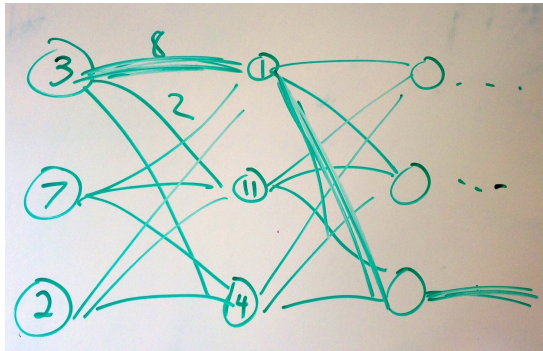


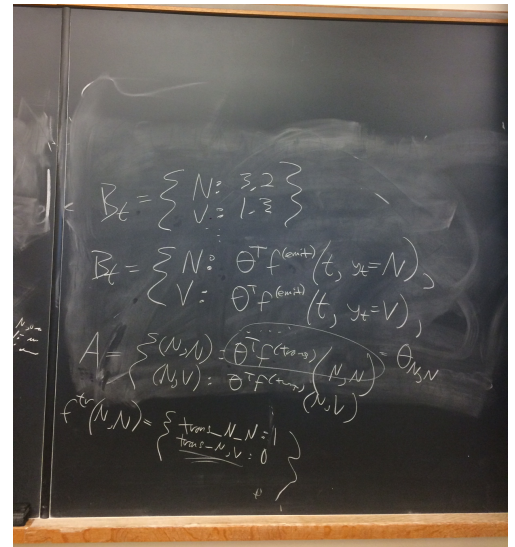
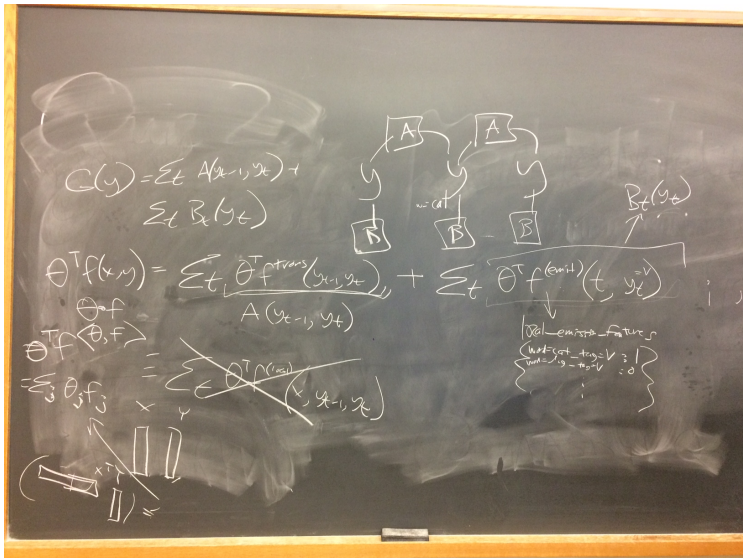
Office hours:
alternate explanation of Viterbi and calc_factor_scores.

LEFT: Think of Viterbi as an algorithm that takes in as input a graph where there are weights on each node and edge. The graph has a lattice structure. It returns the highest-scoring path through that lattice. (where path score = sum of all node and edge weights on path.)

RIGHT: calc_factor_scores builds this graph, represented as (Ascores, Bscores). The B_t scores are node weights and A scores are edge weights (same pattern of edge weights at each timestep).



Blackboard at the end of 10/8/15 lecture: defining local factor scores from local CRF features



$$\text{def } \bar{\theta}_t = \frac{1}{t} \sum_{s=1}^t \theta_s$$

$$\text{Asymptotic } \begin{cases} \theta_t = \theta_1 + g t \\ S_t = S_{t-1} + (t-1) g t \end{cases}$$

$$\text{Asymptotic } \bar{\theta}_t = \theta_1 - \frac{1}{t} S_t$$

$\bar{\theta}_1 = g_1$	$\bar{\theta}_1 = \theta_1 = g_1$
$\bar{\theta}_2 = g_1 + g_2$	$\bar{\theta}_2 = \frac{1}{2}(\theta_1 + \theta_2)$
$\bar{\theta}_3 = g_1 + g_2$	$= \frac{1}{2}(g_1 + g_1 + g_2)$
$= g_1 + g_2 + g_3$	$= g_1 + \frac{1}{2} g_2$

Office hours:
starting Problem 3, averaged perc. weightsums trick.

Some intuition:
bar{theta} overcounts the gradients near the start. they get summed in for all the later thetas. so theta undercounts early gradients relative to bar{theta}. thus we make S, which highly weights gradients near the end. subtracting S out of theta reduces weights of gradients near the end.