CS5340 Lab 1: Belief Propagation and Maximal Probability

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1. factor_product() and factor_sum()

Result is the element-wise multiplication (or addition in case of factor_sum()) of values at A.val[idxA] and B.val[idxB].

2. factor_marginalize() and factor_max_marginalize()

Marginalize every variable from the factor one-by-one. Remove the variable and its cardinality from the original factor to get the final factor variable and cardinality.

Sum (or take maximum in case of factor_max_marginalize()) values of all the rows in the original factor which equals one row in the final factor (without the marginalized variable value). E.g., X0 to be marginalized:

```
Original factor:
                                                  Final factor(sum): Final factor(max):
                              Marg assignment:
   X1 P(X0, X1)
                             X1 P(X0, X1)
                                                                     X1 P(X1)
                                                  X1 P(X1)
0
   0
        0.2
                             0
                                 0.2
                                                  0
                                                      0.5
                                                                         0.3 (argmax={X2=0, X0=1})
1
    0
        0.3 (argmax={X2=0})
                                 0.3
                                                  1
                                                      0.9
                                                                     1
                                                                         0.5 (argmax={X0=0})
0
   1
        0.5
                             1
                                 0.5
1
   1
        0.4 (argmax=None)
                             1
                                 0.4
```

observe_evidence()

Loop through every factor and select the ones where the evidence variable is present. Set probablity of all its assignments to 0 if it doesn't contain the evidence.

4. compute_joint_distribution()

Do a factor product of all factors in the list sequentially.

5. compute_marginals_naive()

Observe evidence for all factors. Compute its joint probability. Marginalize over all the variables minus the given variable. Normalize the probabilities.

6. compute_marginals_bp() Do a DFS on the graph and store messages such that:

```
messages[a][b] ==> message from node a to node b, followed by inference on node b
```

Collect all messages from leaf to root nodes. Distribute all messages from root to the leaf nodes. Do an inference for all variables in V .

- 7. map_eliminate()
- Observe evidence for all factors. Convert all proabilities into its log values.
- Generate graph from the factors. Select root=Node(0).
- Collect all messages from leaf to root nodes. Perform inference on the root node. These steps are enough to give us the:
 - Max probability of the entire graph as maximum probability remains the same no matter what the query node is.
 - Max probability configuration as val_argmax is populated while computing the marginals every time.
- Remove the evidence from the max decoding configuration and add for the root node.