## Report for Lab 1 [CS5340]

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Apart from functions listed in lab1 instruction pdf, some auxiliary functions were added for support some functions.

## compute\_marginals\_bp

For function compute\_marginals\_bp(V, factors, evidence), I added:

- collect(g,i,j,msg)
- sendmessage(g,j,i,msg)
- distribute(g,i,j,msg)
- computeMarginal(g,i,msg)

The design of these functions is according to Lecture 4 slides Page. 68.

## map\_eliminate

For function map\_eliminate, I added:

• map\_sendmsg(g,j,i,msg,cfg): In this function, the operation for passing messages is the same as operations in collect(g,i,j,msg), and original factor\_product and factor\_marginalize is replaced by factor\_sum and factor\_max\_marginalize to get  $m_{ji}^{max}(x_i)$ . As factor\_max\_marginalize return factor with argmax configuration on unmarginalized vars, we can then get the configuration  $\delta_{ii}(x_i)$  (marginalized vars) by replacing the keys in argmax dictionary:

```
cfg[j][i] = {}
for idx, dict_item in enumerate(msg[j][i].val_argmax):
    cfg[j][i][idx] = list(dict_item.values())[0]
```

It's very convenient.

- map\_collect(g,i,j,msg,cfg)
- map\_distribute(g, i, j, msg, cfg, max\_decoding)

These functions' pipelines have some similarities with the 4 functions I implemented above for <code>compute\_marginals\_bp</code>, and some modifications are made to make them work in log spaces and to find the configuration of max probability. The design of these functions is according to Lecture 5 slides Page. 48. But, there are some little changes. There is no <code>SETVALUE(i,j)</code> function like pseudo codes in slides. Considering it can be realized in one line code, and only be called in <code>map\_distribute</code> function, I realized it in <code>map\_distribute</code> function.