### Question 1.

As you can see from the equation, m is  $0.9 * m_{prev}$  and 0.1 \* gradient. This means that the amount it varries is mostly dependent on value of  $m_{prev}$ .

This is useful because there will probably be lots of noise in our minibatches,, so this approach will on average be less susceptible to noise

### Question 2.

Overall, v < 0,  $\sqrt(v) < 0$  for terms where the gradient is < 1. Thus, for those terms, there will be a larger update. Conversely, if the gradient > 1, then v > 0,  $\sqrt(v) > 0$  and those terms will get smaller update.

This means smaller gradients gets slightly larger updates and larger gradients get slightly smaller learning updates. This might be useful as all the terms are updated at a similar pace, meaning they all converge at the same time.

## Question 3.

We know that  $E[h_{drop}]_i = E[d \odot h]_i = (1 - p_{drop}) * h_i + p_{drop} * 0 = (1 - p_{drop}) * h_i$ Since we want  $E[\gamma d \odot h_i] = h_i$ , then  $h_i = E[\gamma d \odot h_i] = \gamma E[d \odot h_i] = \gamma (1 - p_{drop}) h_i$ 

$$h_i = \gamma (1 - p_{drop}) h_i$$
$$1 = \gamma (1 - p_{drop})$$
$$\frac{1}{(1 - p_{drop})} = \gamma$$

## Question 4.

We want to apply dropout during training so that some hidden units will learn similar information as other hidden laters. During evaluation, we want to use all the units to get the best results. Although some hidden units may have learned the same information, having all of them present is kind of similar to taking a "vote" amongst the hidden units and choosing the best results.

# Question 5.

| Stack                          | Buffer                             | New Dep            | Trans          |
|--------------------------------|------------------------------------|--------------------|----------------|
| (root)                         | (i parsed this sentence correctly) |                    | Initial Config |
| (root, i)                      | (parsed this sentence correctly)   |                    | shift          |
| (root, i, parsed)              | (this sentence correctly)          |                    | shift          |
| (root, parsed)                 | (this sentence correctly)          | parsed -> I        | left-arc       |
| (root, parsed, this)           | (sentence correctly)               |                    | shift          |
| (root, parsed, this, sentence) | (correctly)                        |                    | shift          |
| (root, parsed, sentence)       | (correctly)                        | sentence -> this   | left-arc       |
| (root, parsed)                 | (correctly)                        | parsed -> sentence | right-arc      |
| (root, parsed, correctly)      | ()                                 |                    | shift          |
| (root, parsed)                 | ()                                 | parsed->correctly  | right-arc      |
| (root)                         | ()                                 | root->parsed       | right-arc      |

#### Question 6.

We know that we must call SHIFT once for every single word to move the word from buffer to stack (n SHIFT steps).

We also know that in a tree w/n nodes, there are n-1 verticies. In our final parse tree, we have exactly n+1 nodes (n words +1 ROOT). In addition, each LEFT-ARC or RIGHT-ARC step creates exactly 1 edge. Hence, there are exactly n ARC steps.

In total, there will be n+n=2n parse steps. If you count the initial config as a step too, then there will be 2n+1 parse steps.

### Question 7.

Error type: Verb Phrase Attachment Error Incorrect Dependency: wedding -> fearing Correct Dependency: heading -> fearing cs224n a3 (arthurg)

# Question 8.

Error type: Coordination Attachment Error Incorrect Dependency: makes -> rescue Correct Dependency: rush -> rescue

## Question 9.

Error Type: Prepositional Phrase Attachment Error

Incorrect Dependency: named -> Midland Correct Dependency: guy -> Midland

## Question 10.

Error Type: Modifier Attachment Error Incorrect Dependency: elements -> most Correct Dependency: crucial -> most