

# NMT(Neural Machine Translator) Decoder Sample Slides

## By Mohit Kumar

# NMT:Encoder-Architecture:Decoder

- **Output:**

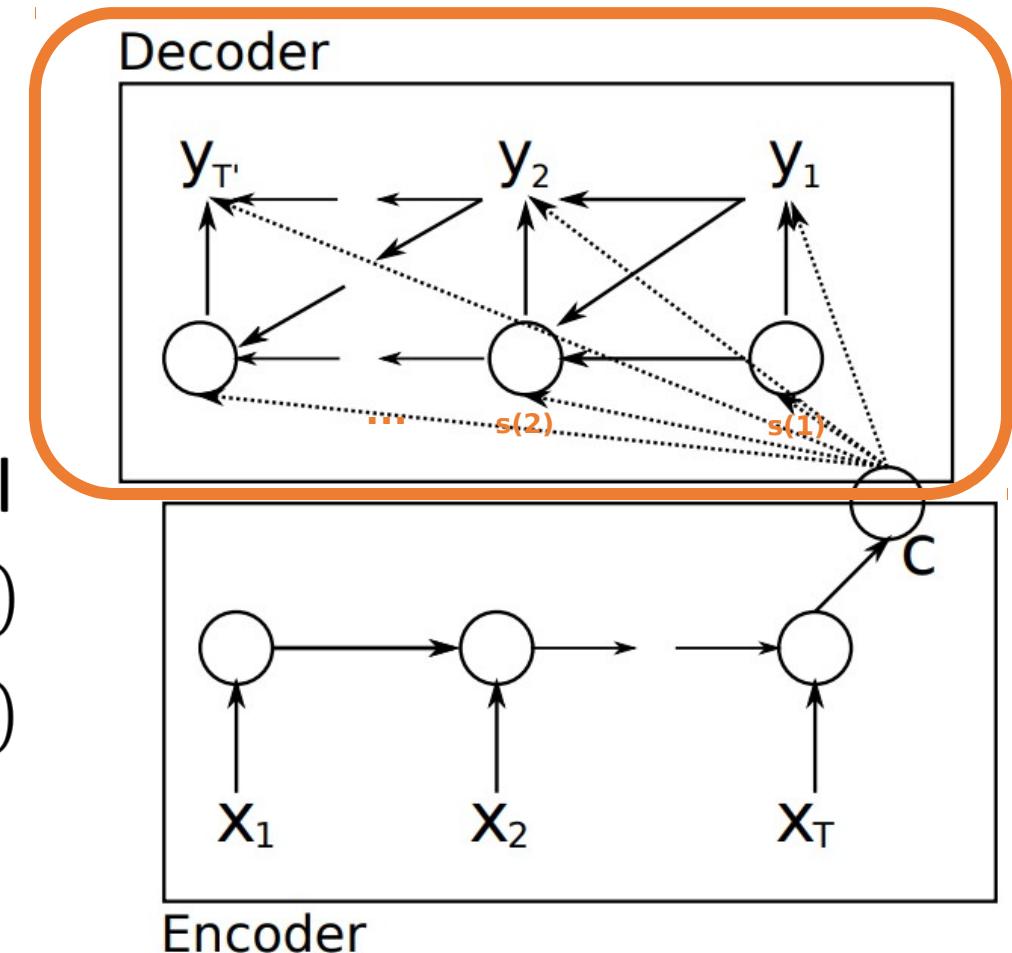
- word sequence (target)  $y = (y_1, \dots, y_{T'})$

- **Decoder Transition:**

- $s_t = RNN_{dec}(s_{t-1}, y_{t-1}, c)$

- **Conditional recurrent language model**

- $p(\vec{y}|\vec{x}) = \prod_{i=1}^{T'} p(y_i|y_1 \dots, y_{i-1}; c)$
- $p(y_i|y_1 \dots, y_{i-1}; c) = g(s_i, y_{i-1}, c)$



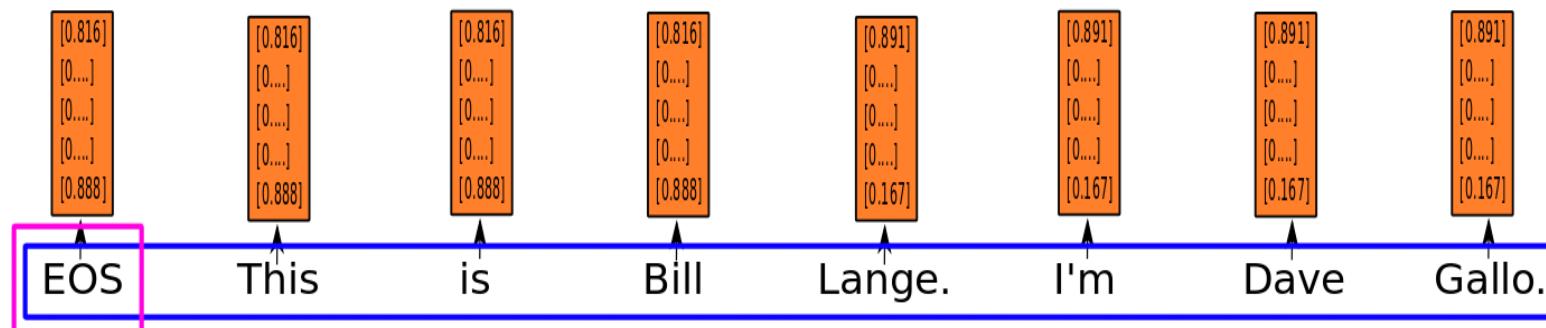
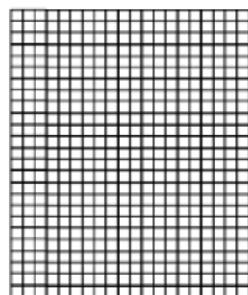
**g(·): Function that produce valid probs,  
e.g. softmax**

# NMT:Detailed:Encoder-Decoder:Decoder:Graph(step-3)

```
def process_encoding_input(target_data, word2int, batch_size):
    ending = tf.strided_slice(target_data, [0, 0], [batch_size, -1], [1, 1])
    decoding_input = tf.concat([tf.fill([batch_size, 1], word2int['TOKEN_GO']), ending], 1)
    return decoding_input
```

[[37 16 37 37 37 37 37 39 38 38 38 38 38 38 38]  
[40 37 16 37 37 37 37 39 38 38 38 38 38 38 38]]

- Pick up the corresponding English sentence.
- End of statement marked.
- GO marker added.

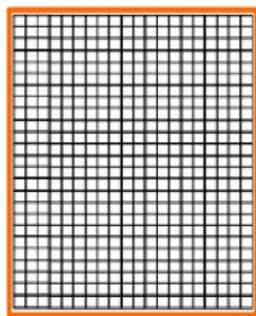


# NMT:Detailed:Encoder-Decoder:Decoder:Graph(step-4)

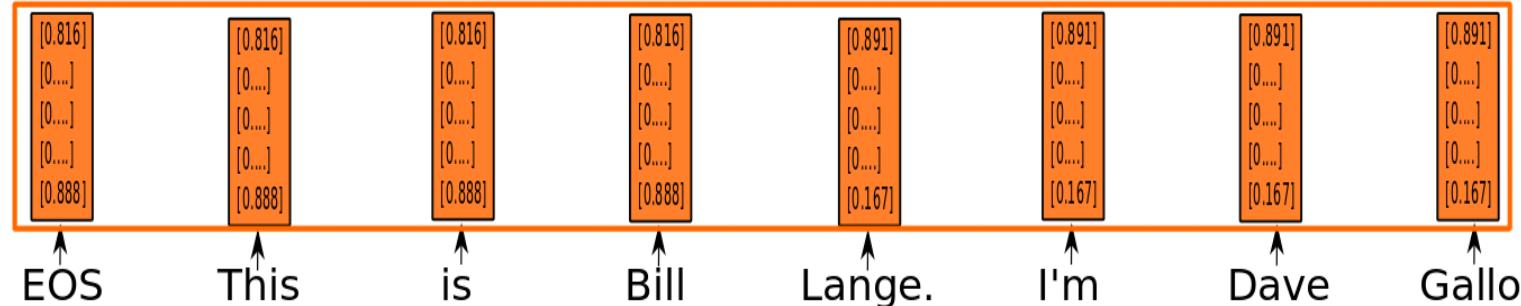
```
decoding_input = process_encoding_input(target_en_data, word2int_en, batch_size)
```

```
decoding_embed_input = tf.nn.embedding_lookup(en_embeddings_matrix, decoding_input)
```

- Pick up the embeddings from english embeddings.
- Not a "tf variable so no learning" english embeddings while training.



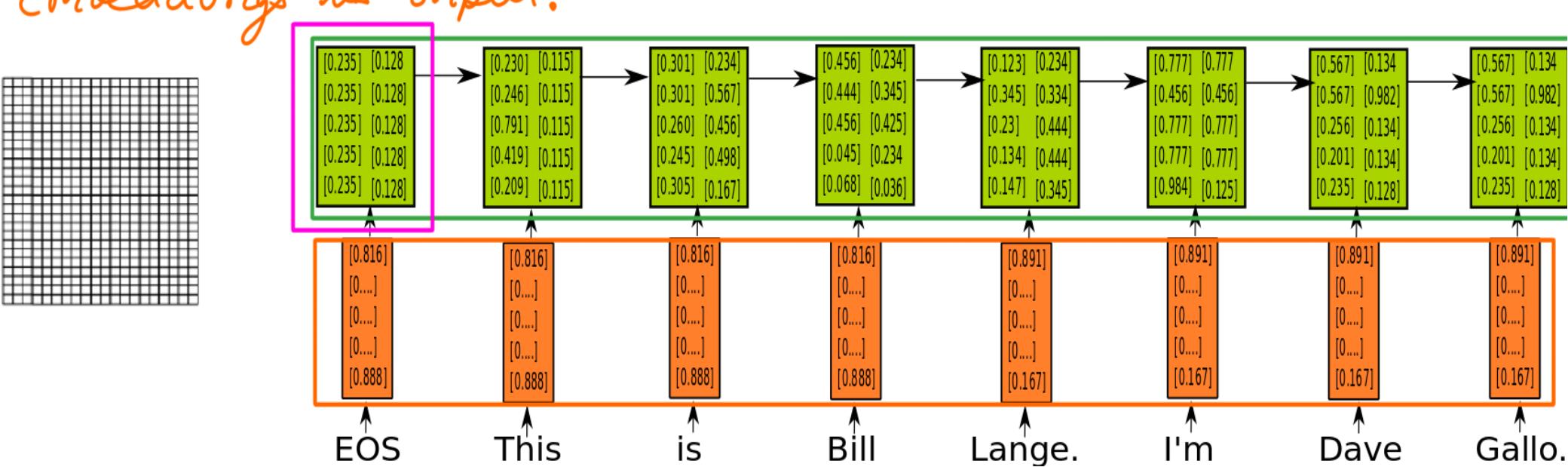
English Embeddings



# NMT:Detailed:Encoder-Decoder:Decoder:Graph(step-5)

```
with variable_scope.variable_scope(  
    "decoder", initializer=init_ops.constant_initializer(0.1)) as vs:  
    helper = TrainingHelper(inputs=decoding_embed_input, sequence_length=en_len, time_major=False)  
    dec = BasicDecoder(decoding_cell, helper, encoding_st op_layer )  
    logits, _, _ = dynamic_decode(dec, output_time_major=False, impute_finished=True,  
                                  maximum_iterations=max_en_len)  
    return logits
```

- Similar to dynamic\_rnn, except allows more control over  $c$  and  $h$  states after every step.
- Last  $c$  and  $h$  state of the encoder.
- Embeddings as input.



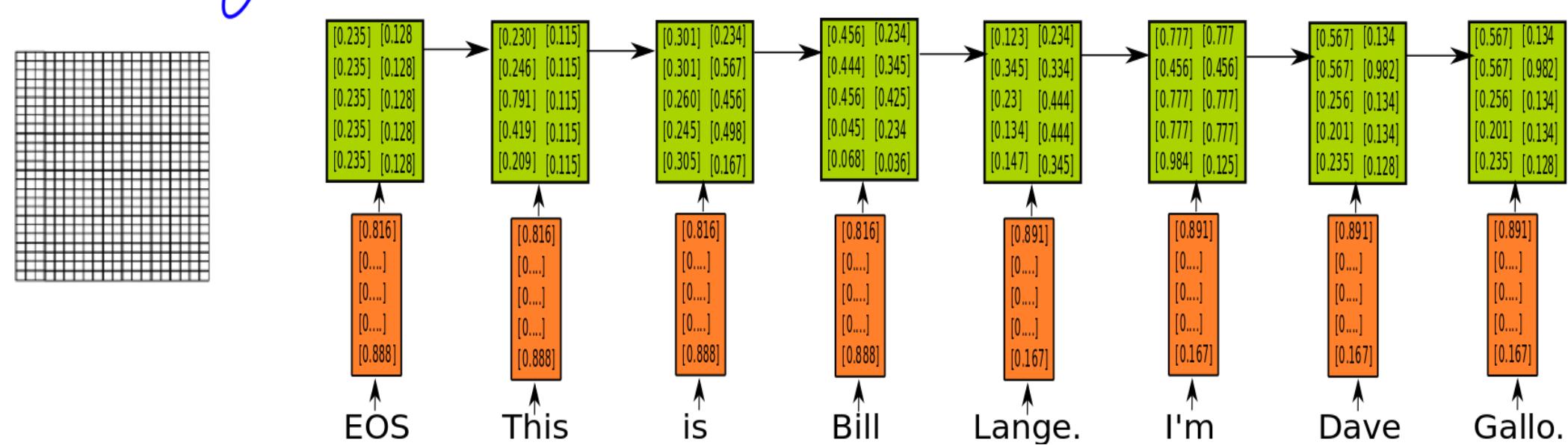
# NMT:Detailed(DB):Encoder-Decoder:Decoder:Graph(step-5)

```
with variable_scope.variable_scope(  
    "decoder", initializer=init_ops.constant_initializer(0.1)) as vs:  
    helper = TrainingHelper(inputs=decoding_embed_input, sequence_length=en_len, time_major=False)  
    dec = BasicDecoder(decoding_cell, helper, encoding_st, op_layer )  
    logits, _, _ = dynamic_decode(dec, output_time_major=False, impute_finished=True,  
                                  maximum_iterations=max_en_len)  
return logits
```

- Training helper's next input function returns the next input.
- For pre and post processing at every step.

• Will be more useful with attention.

```
def next_inputs(self, time, outputs, state, name=None):  
  
    self.time=time+1  
    finished=False  
    if(self.time == self.sequence):  
        finished=True  
    if(not finished):  
        next_inputs=self.inputs[0:self.batch,self.time].T  
    else:  
        next_inputs=None  
    return (finished, next_inputs, state)
```



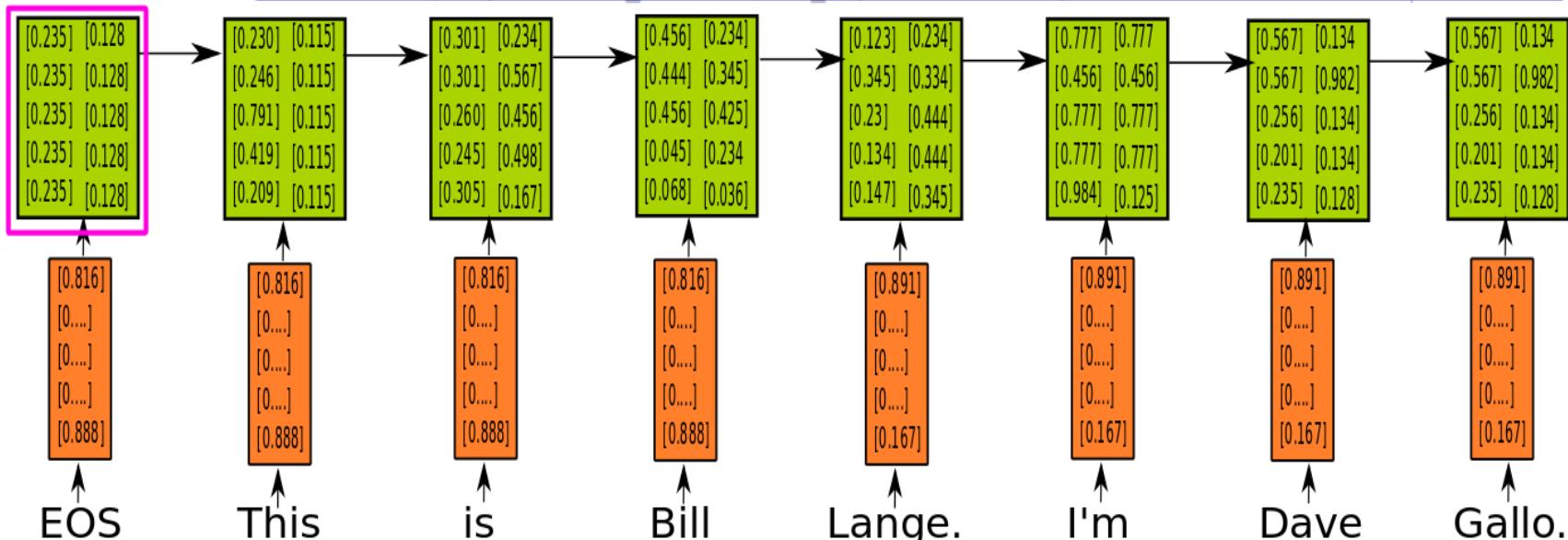
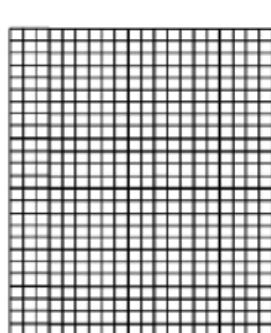
# NMT:Detailed(DB):Encoder-Decoder:Decoder:Graph(step-5)

```
with variable_scope.variable_scope(  
    "decoder", initializer=init_ops.constant_initializer(0.1)) as vs:  
    helper = TrainingHelper(inputs=decoding_embed_input, sequence_length=en_len, time_major=False)  
    dec = BasicDecoder(decoding_cell, helper, encoding_st, op_layer )  
logits, _, _ = dynamic_decode(dec, output_time_major=True,  
                           maximum_iterations=1000)  
return logits
```

• BasicDecoder's step function

1. accepts current input and initial state.

```
def step(self, time, inputs, initial_state, name=None):  
    #print("Decoder.step.initial_state:",initial_state,self._cell)  
    if(initial_state is None):  
        output,newstate = self._cell(inputs)  
    else:  
        output,newstate = self._cell(inputs,initial_state)  
  
    if (self.output_layer is not None):  
        output=self.output_layer(output[-1])  
    else:  
        output=output[-1]  
    self.result[time]=output  
  
    finished, next_input, next_state=self._helper.next_inputs(time, output, newstate)  
    if self.attention is False:  
        next_state=None  
    return (output, next_state, next_input, finished)
```



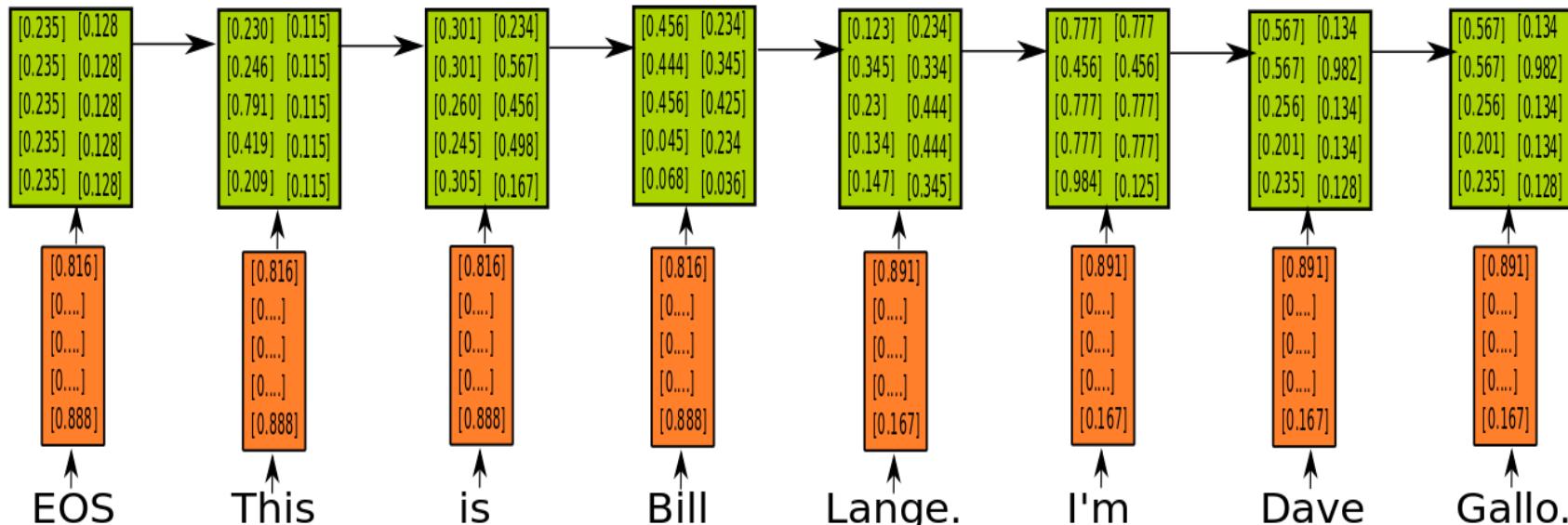
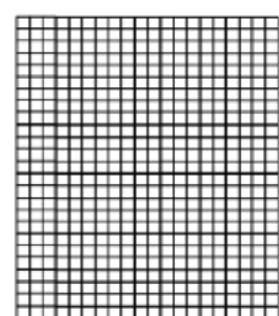
# NMT:Detailed(DB):Encoder-Decoder:Decoder:Graph(step-5)

```
with variable_scope.variable_scope(  
    "decoder", initializer=init_ops.constant_initializer(0.1)) as vs:  
    helper = TrainingHelper(inputs=decoding_embed_input, sequence_length=en_len, time_major=False)  
    dec = BasicDecoder(decoding_cell, helper, encoding_st, op_layer )  
    logits, _, _ = dynamic_decode(dec, output_time_major=True,  
                                maximum_iterations=en_len)  
    return logits
```

## • BasicDecoder's step function

2. Transform the RNN output with a FF multiplier, to "Y" shape.

```
def step(self, time, inputs, initial_state, name=None):  
    #print("Decoder.step.initial_state:",initial_state,self._cell)  
    if(initial_state is None):  
        output,newstate = self._cell(inputs)  
    else:  
        output,newstate = self._cell(inputs,initial_state)  
  
    if (self.output_layer is not None):  
        output=self.output_layer(output[-1])  
    else:  
        output=output[-1]  
    self.result[time]=output  
  
    finished, next_input, next_state=self._helper.next_inputs(time, output, newstate)  
    if self.attention is False:  
        next_state=None  
    return (output, next_state, next_input, finished)
```



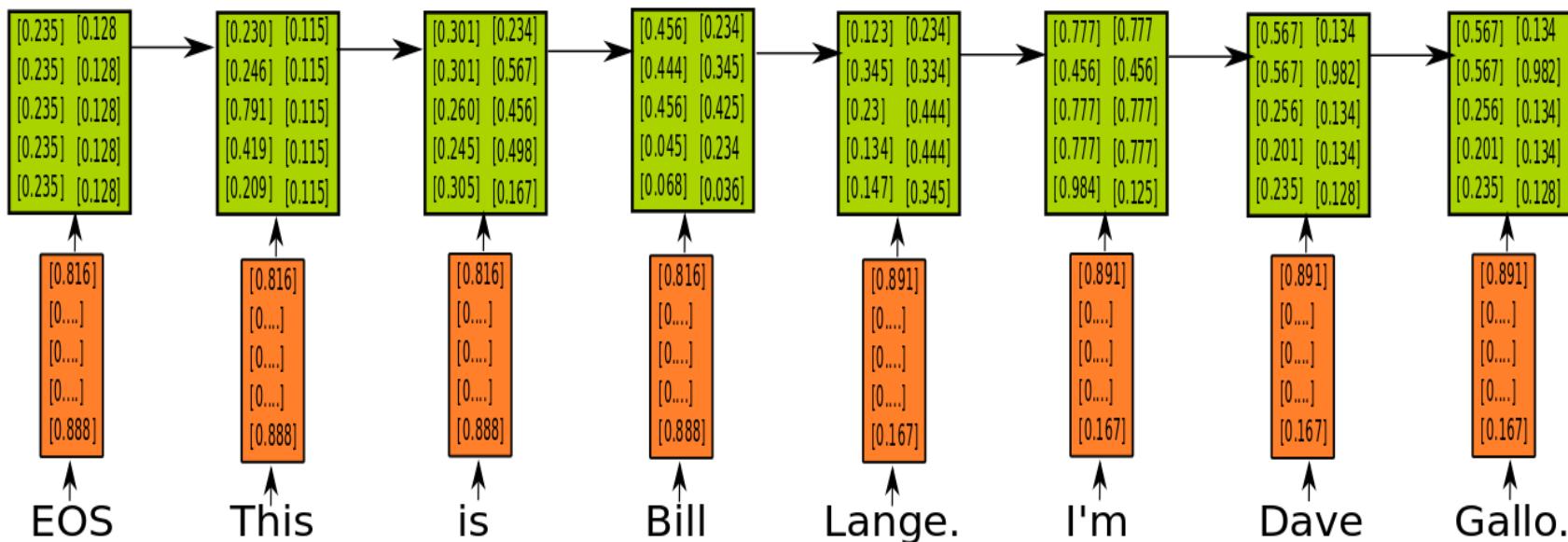
# NMT:Detailed(DB):Encoder-Decoder:Decoder:Graph(step-5)

```
with variable_scope.variable_scope(  
    "decoder", initializer=init_ops.constant_initializer(0.1)) as vs:  
    helper = TrainingHelper(inputs=decoding_embed_input, sequence_length=en_len, time_major=False)  
    dec = BasicDecoder(decoding_cell, helper, encoding_st, op_layer )  
logits, _, _ = dynamic_decode(dec, output_time_major=True,  
                               maximum_iterations=100)  
return logits
```

• BasicDecoder's step function

3. Request for the next input and return.

```
def step(self, time, inputs, initial_state, name=None):  
    #print("Decoder.step.initial_state:",initial_state,self._cell)  
    if(initial_state is None):  
        output,newstate = self._cell(inputs)  
    else:  
        output,newstate = self._cell(inputs,initial_state)  
  
    if (self.output_layer is not None):  
        output=self.output_layer(output[-1])  
    else:  
        output=output[-1]  
    self.result[time]=output  
  
    finished, next_input, next_state=self._helper.next_inputs(time, output, newstate)  
    if self.attention is False:  
        next_state=None  
    return (output, next_state, next_input, finished)
```



# NMT:Detailed(DB):Encoder-Decoder:Decoder:Graph(step-5)

```
with variable_scope.variable_scope(
    "decoder", initializer=init_ops.constant_initializer):
    helper = TrainingHelper(inputs=decoding_embed_input,
    dec = BasicDecoder(decoding_cell, helper, encoding_st
logits, _, _ = dynamic_decode(dec, output_time_major=
                                maximum_iterations=max_
return logits
```

1. Controls the loop

2. internally when the input has finished it terminates the loop.

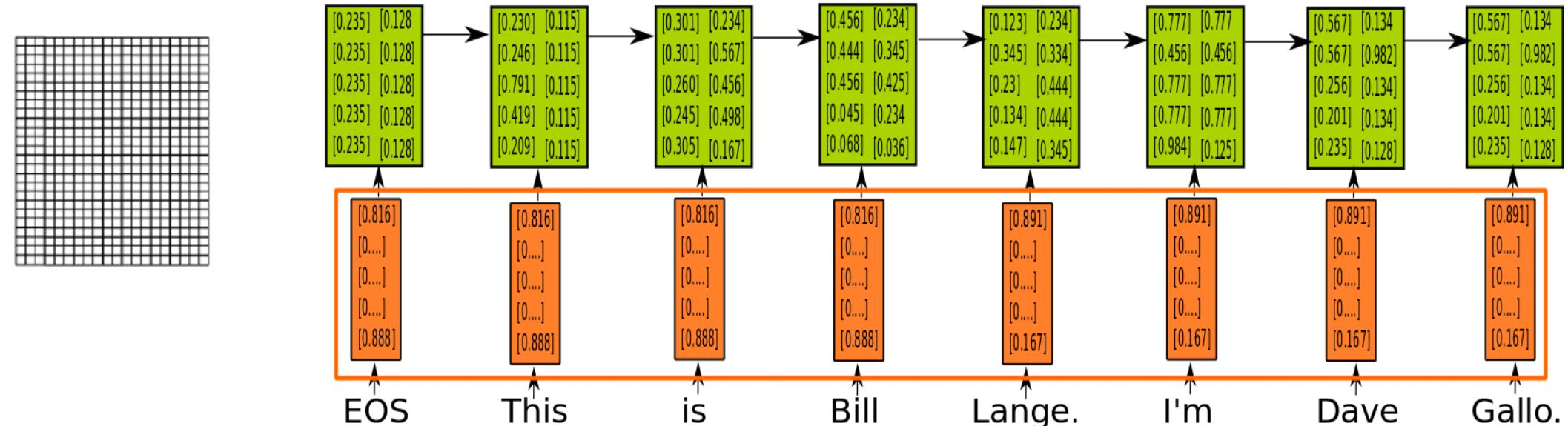
*3. Internally implemented using tf.while loop to optimize using GPUs.*

```
def dynamic_decode(decoder,
                   output_time_major=False,
                   impute_finished=False,
                   maximum_iterations=None,
                   parallel_iterations=32,
                   scope=None):
    finished, next_inputs, initial_state = decoder.initialize()
    time = 0

    next_state = initial_state
    while not finished:
        outputs, next_state, next_inputs, finished = decoder.step(
            time, next_inputs, next_state)
        time = time + 1

    decoder.finalize(outputs, next_state)

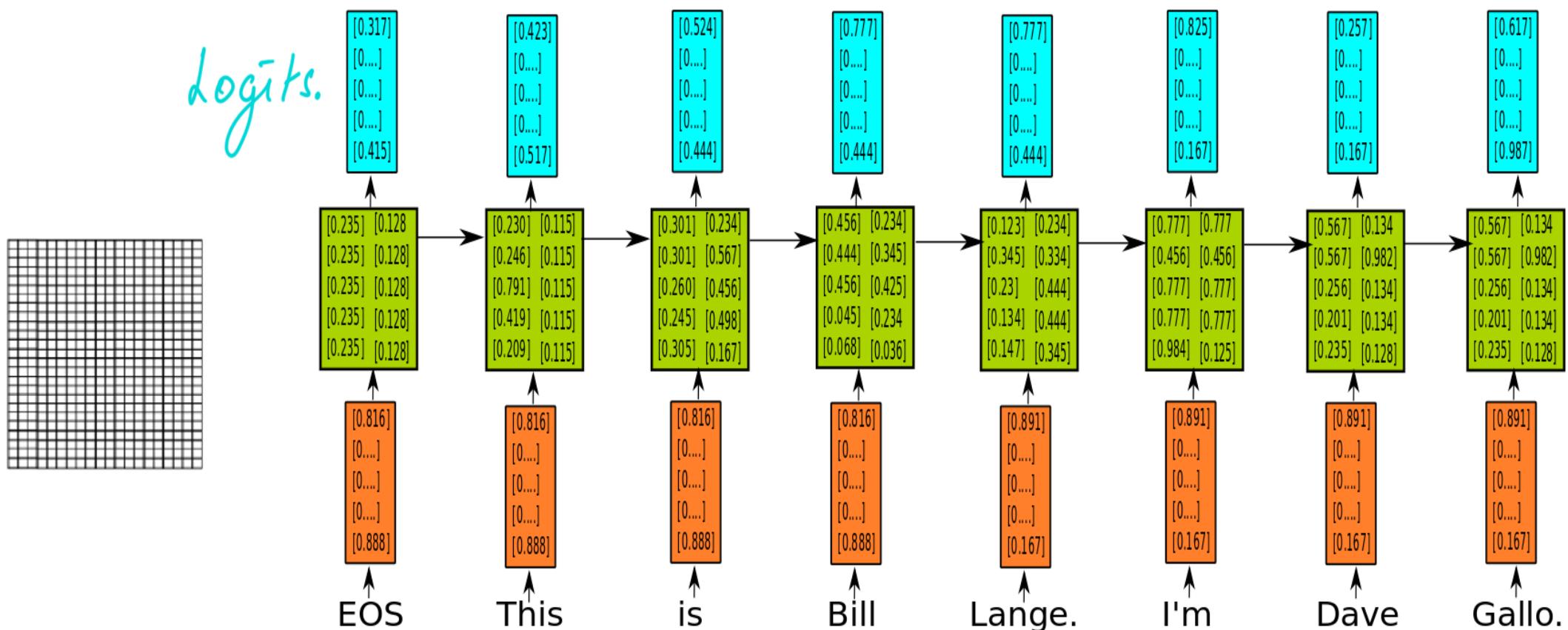
    return decoder.get_final_results(), next_state
```



# NMT:Detailed(DB):Encoder-Decoder:Decoder:Graph(step-5)

```
with variable_scope.variable_scope(  
    "decoder", initializer=init_ops.constant_initializer(0.1)) as vs:  
    helper = TrainingHelper(inputs=decoding_embed_input, sequence_length=en_len, time_major=False)  
    dec = BasicDecoder(decoding_cell, helper, encoding_st, op_layer )  
    logits, _, _ = dynamic_decode(dec, output_time_major=False, impute_finished=True,  
                                maximum_iterations=max_en_len)  
return logits
```

1. Feed forward layer transforms  
the h-state to "Y" shape.

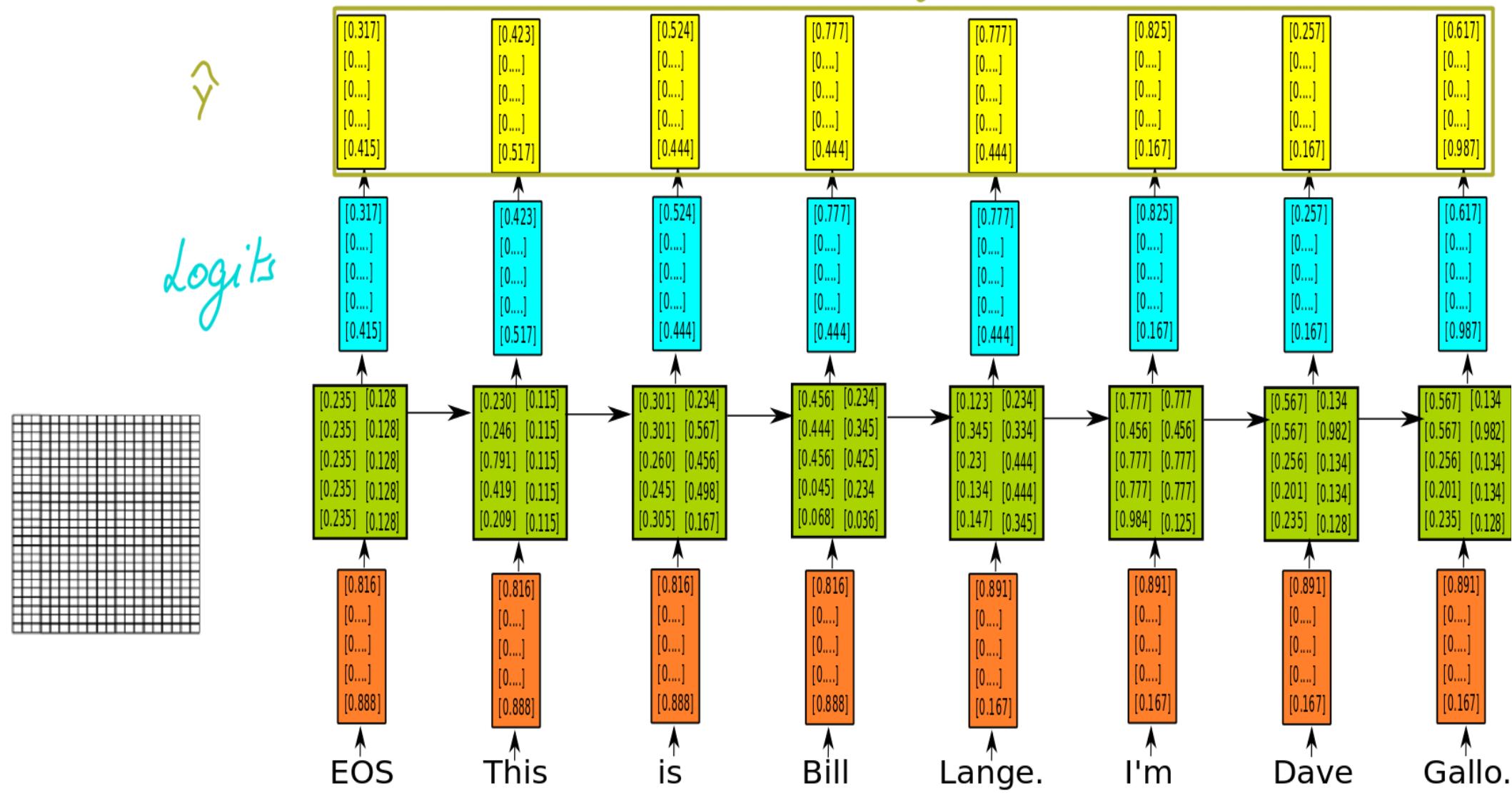


# NMT:Detailed:Encoder-Decoder:Decoder:Graph(step-6)

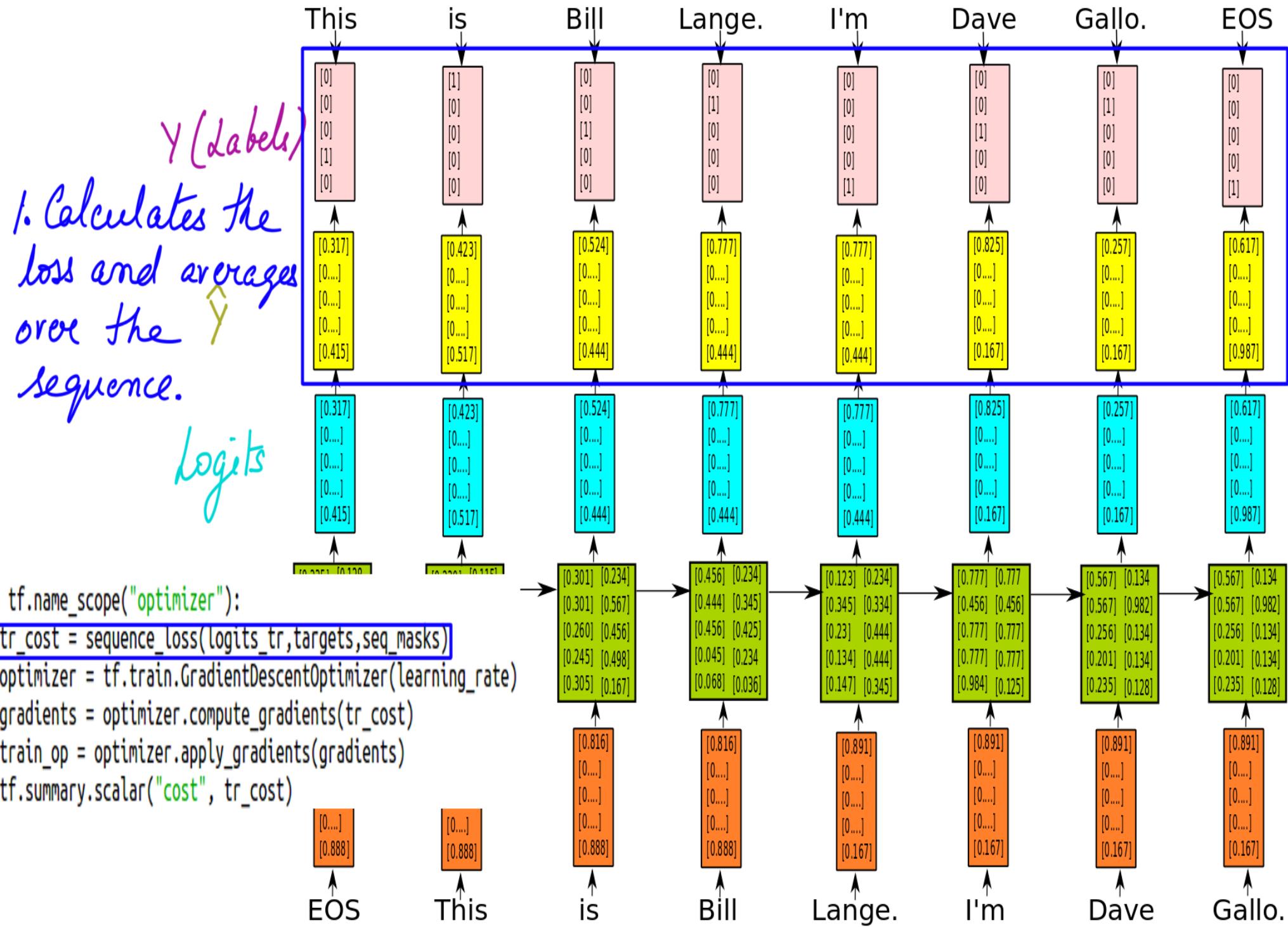
```
with tf.name_scope("optimizer"):
```

```
    tr_cost = sequence_loss(logits_tr, targets, seq_masks)
    optimizer = tf.train.GradientDescentOptimizer(learning_rate)
    gradients = optimizer.compute_gradients(tr_cost)
    train_op = optimizer.apply_gradients(gradients)
    tf.summary.scalar("cost", tr_cost)
```

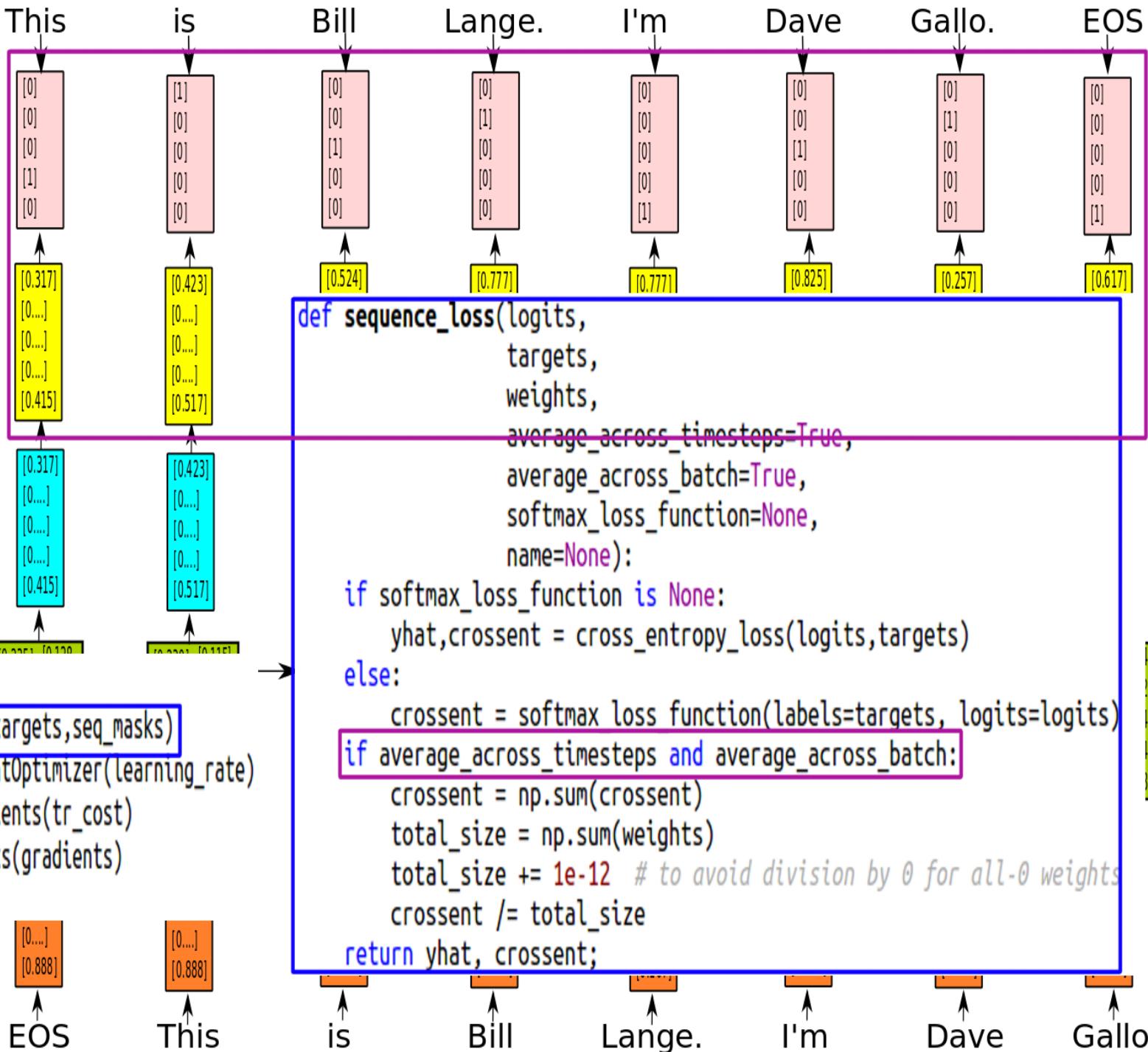
1. Softmax ( $\hat{Y}$ ) and Loss ( $E$ ) calculation are done internally. In fact this internally calls softmax\_loss function.



# NMT:Detailed:Encoder-Decoder:Decoder:Graph(step-6)



# NMT:Detailed(DB):Encoder-Decoder:Decoder:Graph(step-6)



# NMT:Detailed:Encoder-Decoder:Decoder:Graph(step-7)

```
fr_embeddings_matrix,en_embeddings_matrix,fr_word2int,en_word2int,fr_filtered,en_filtered,args=get_nmt_data()
set_modelparams(args)
train_graph = tf.Graph()
with train_graph.as_default():
    input_data, targets, learning_rate, dropout_probs, en_len, max_en_len, fr_len = model_inputs()
    encoding_embed_input, encoding_op, encoding_st, rnn_inputs, decoding_input,decoding_embed_input, logits_tr = seq2seq_model(
        tf.reverse(input_data, [-1]),
        targets,
        dropout_probs,
        fr_len,
        en_len,
        max_en_len,
        len(en_word2int) + 1,
        hidden_size,
        n_layers,
        en_word2int,
        batch_size,
        encoder_type,fr_embeddings_matrix,en_embeddings_matrix)
```

- The graph defined
- The input when served reverse increases the accuracy.