

NMT(Neural Machine Translator) Decoder Sample Slides

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NMT:Encoder-Architecture:Decoder

- **Output:**

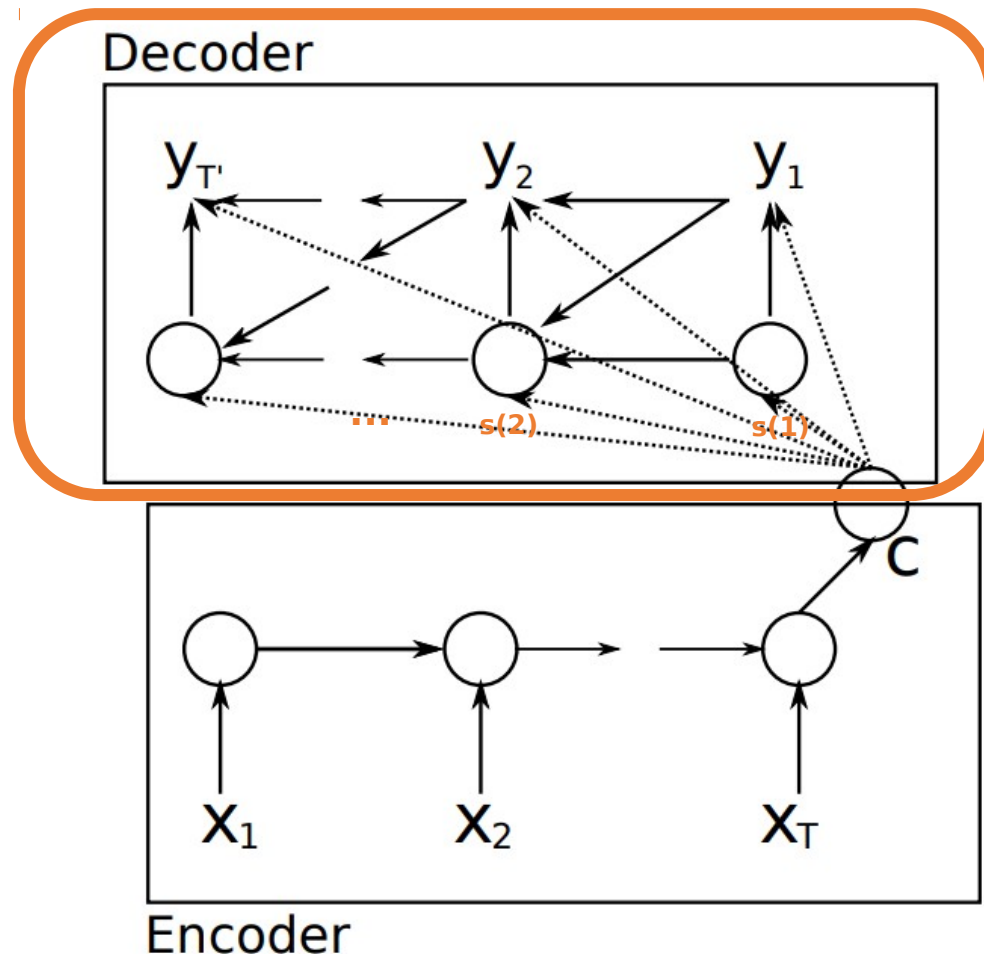
- word sequence (target) $y = (y_1, \dots, y_{T'})$

- **Decoder Transition:**

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

- **Conditional recurrent language model**

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



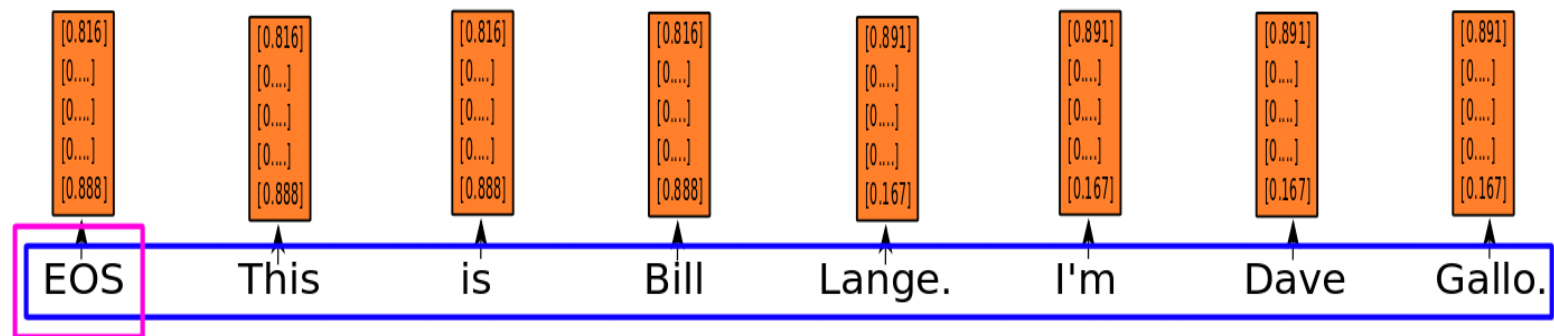
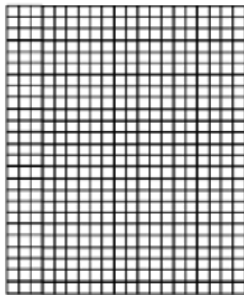
**$g(\cdot)$: Function that produce valid probs,
e.g. softmax**

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 37 39 38 38 38 38 38 38]]

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

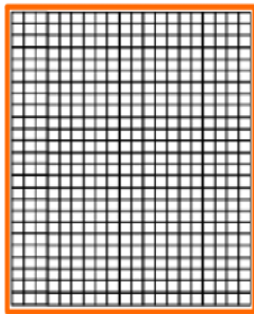


NMT:Detailed:Encoder-Decoder:Decoder:Graph(step-4)

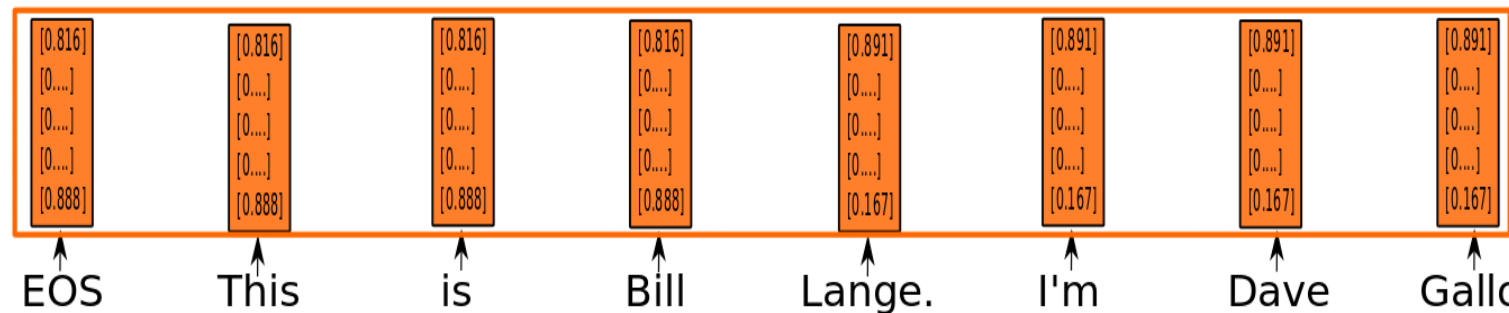
```
decoding_input = process_encoding_input(target_en_data, word2int_en, batch_size)
```

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

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



English Embeddings

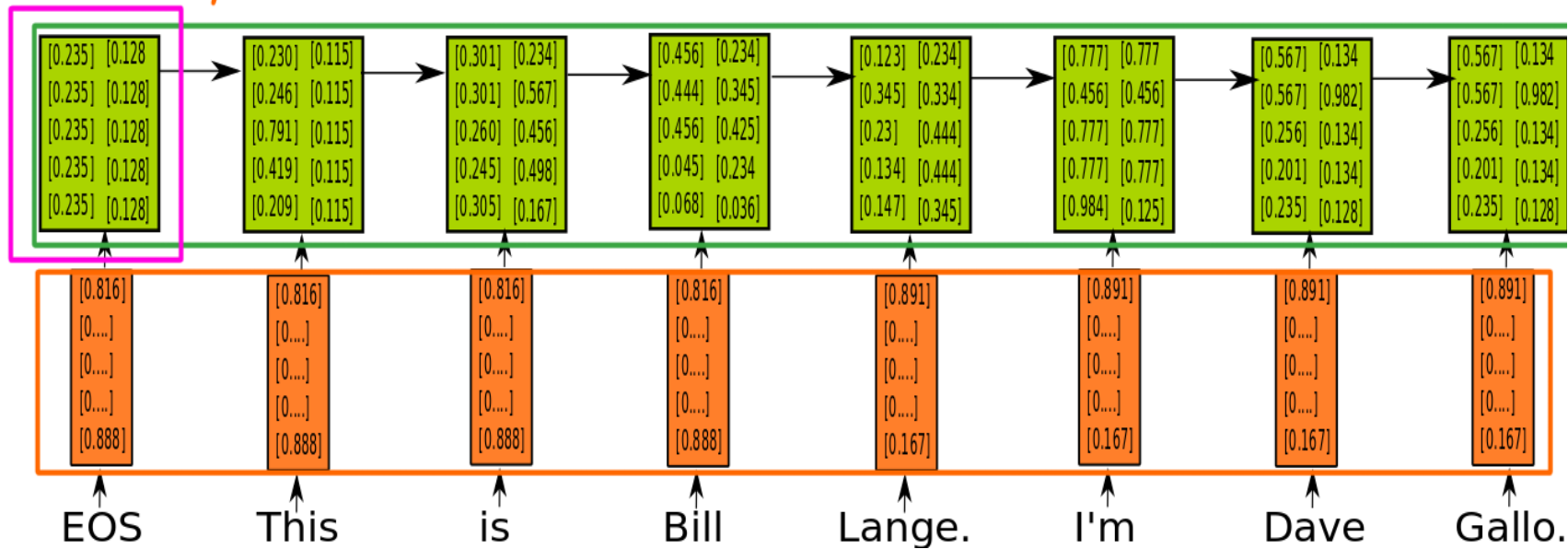
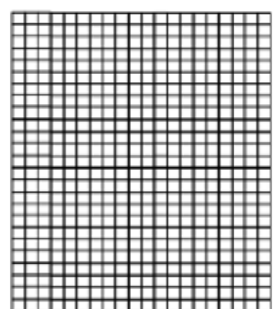


NMT:Detailed:Encoder-Decoder:Decoder:Graph(step-5)

```
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_state, 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.



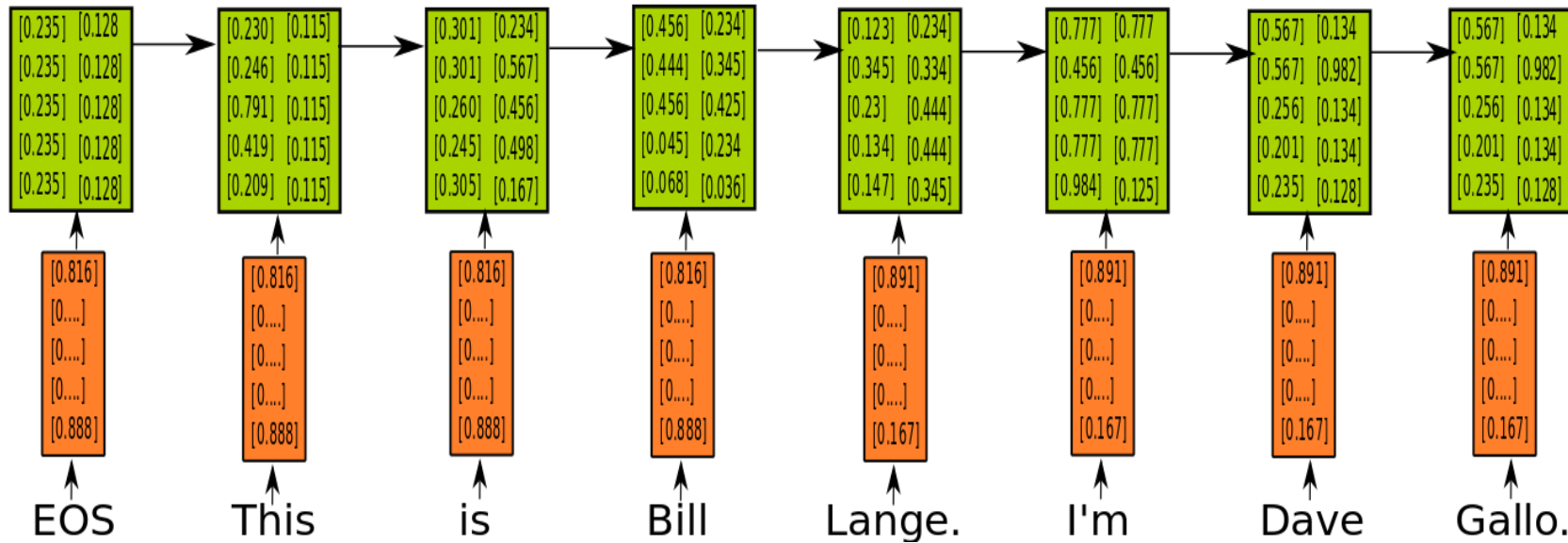
NMT:Detailed(DB):Encoder-Decoder:Decoder:Graph(step-5)

```
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.

```
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)
```

• Will be more useful with attention.



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_step, maximum_iterations)
    return logits
```

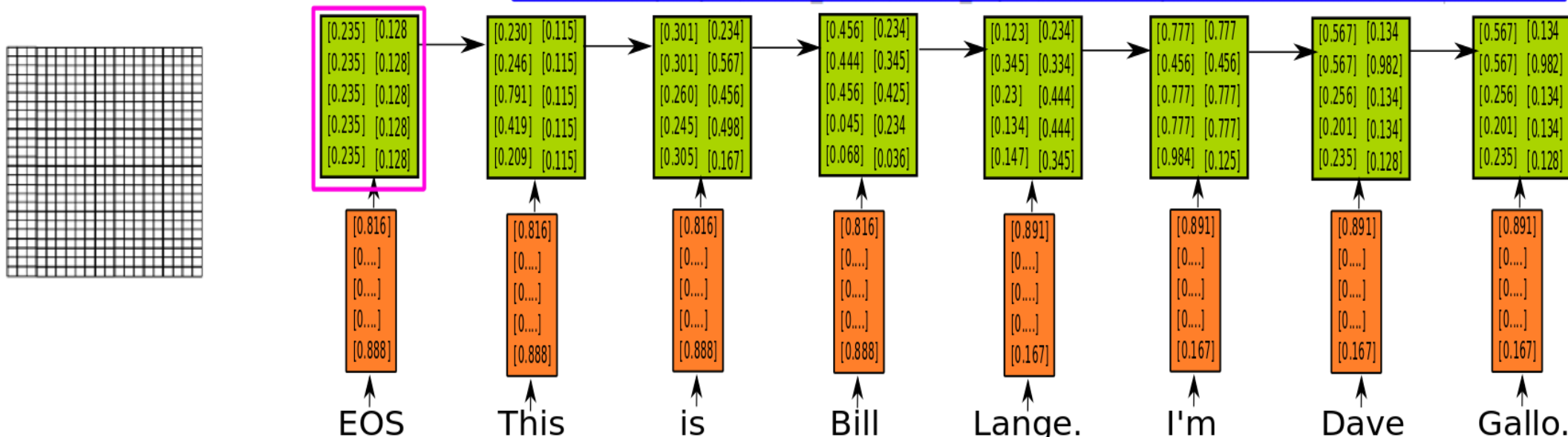
BasicDecoder's step function

- 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)
```



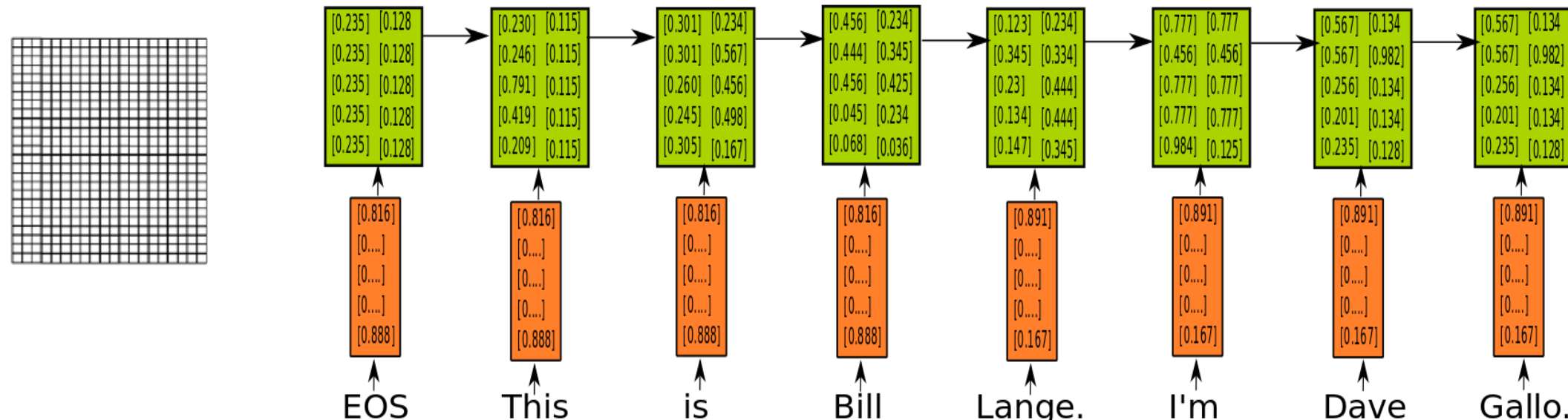
NMT:Detailed(DB):Encoder-Decoder:Decoder:Graph(step-5)

```
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_steps, maximum_iterations, helper.initial_state)  
    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_steps=
```

```
    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]
```

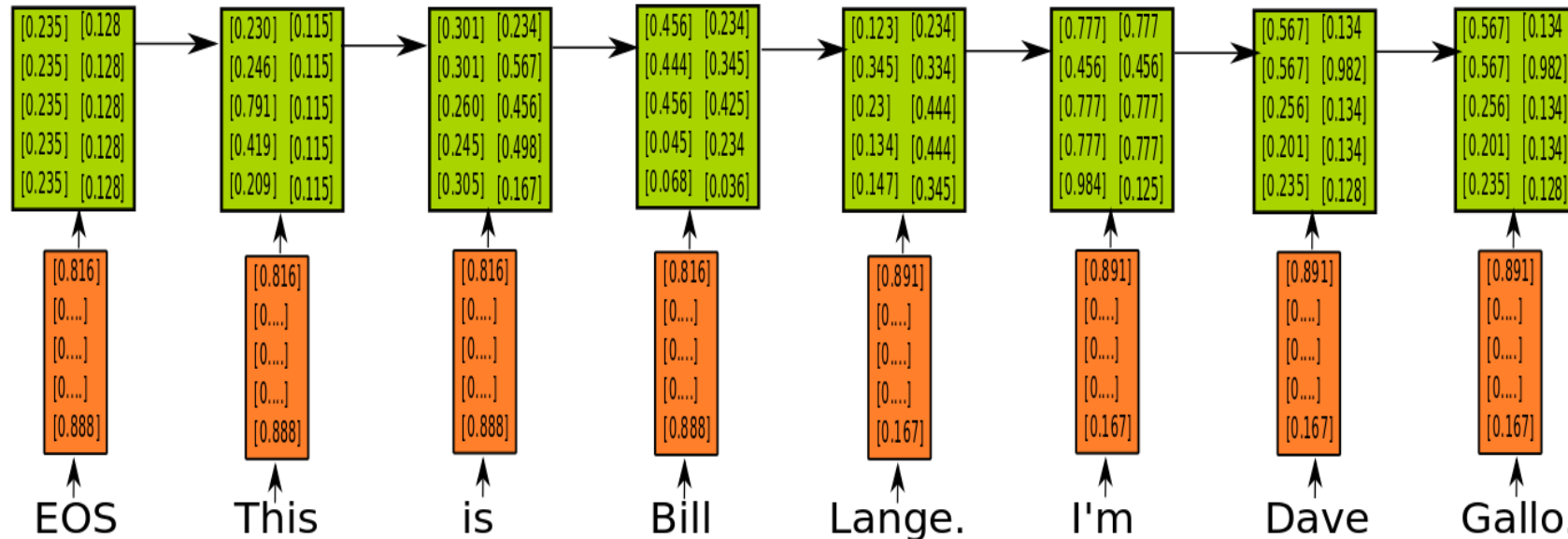
```
    self.result[time] = output
```

```
    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)
```



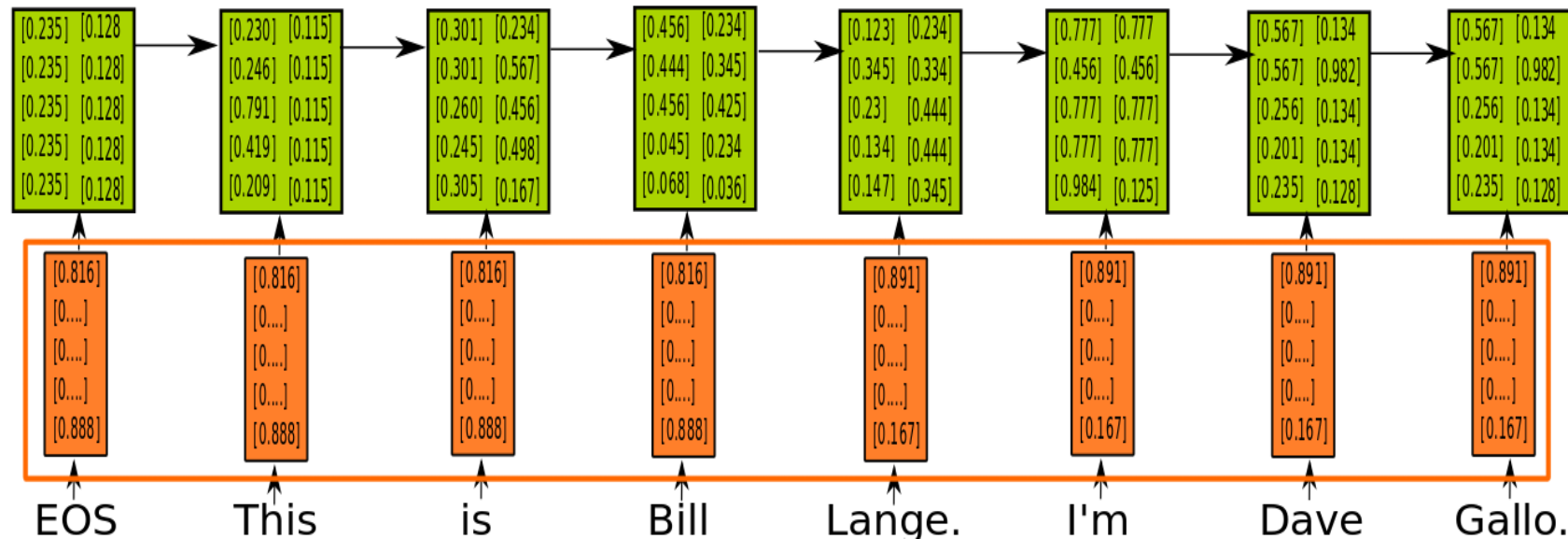
NMT:Detailed(DB):Encoder-Decoder:Decoder:Graph(step-5)

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

```
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
```

3. Internally implemented using a while loop to optimize using GPUs.

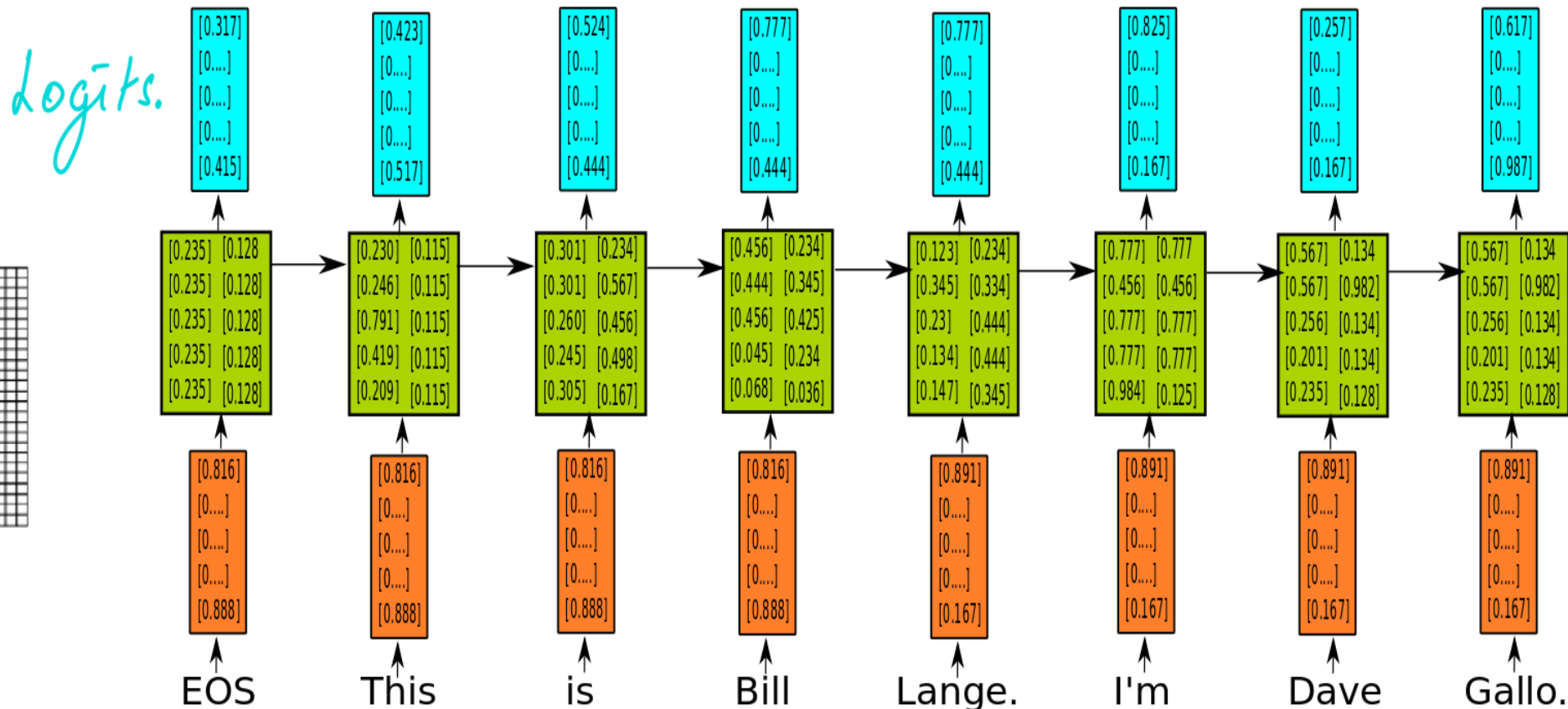
1. Controls the loop
2. internally when the input has finished it terminates the loop.



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.

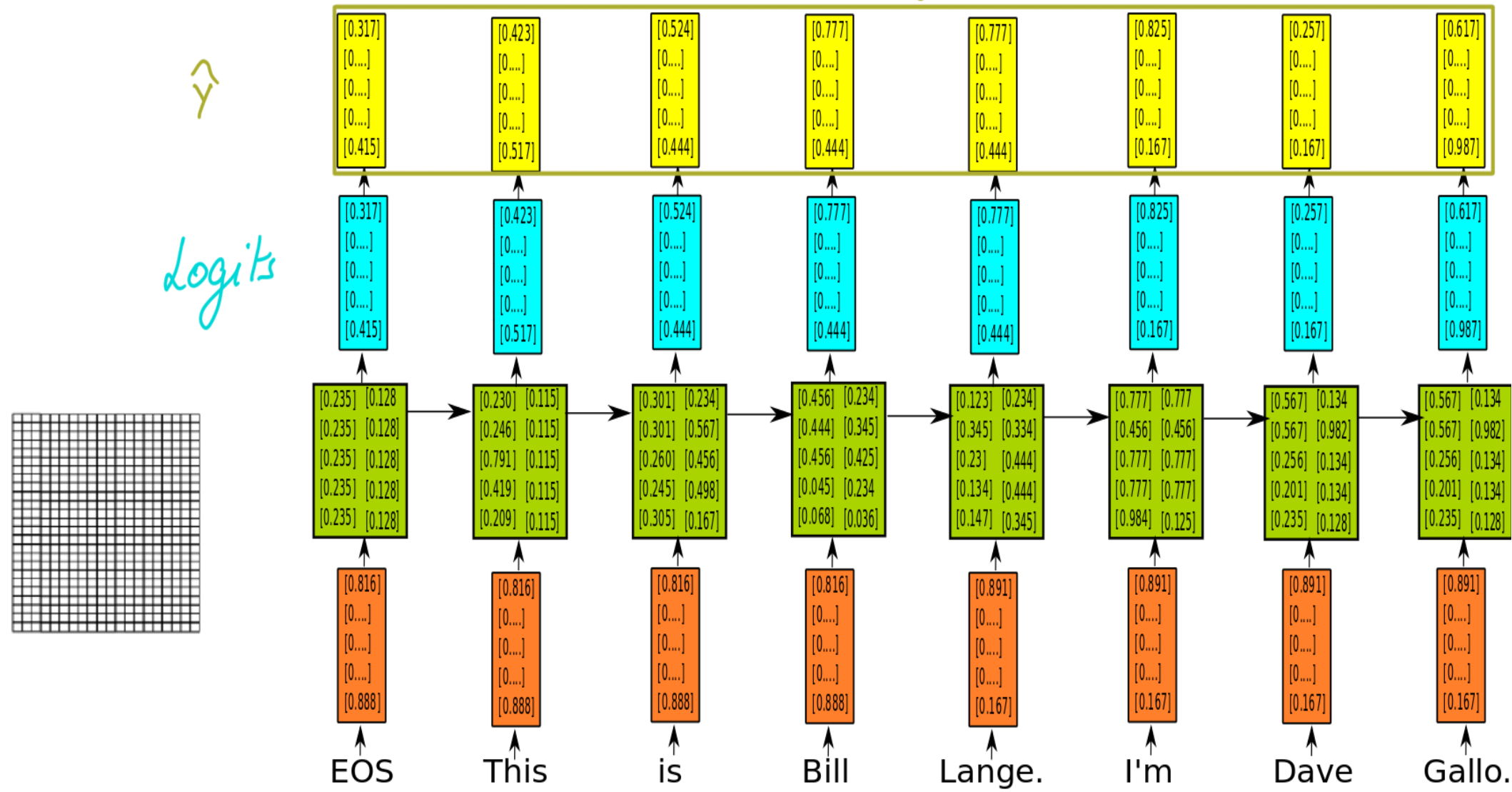


NMT:Detailed:Encoder-Decoder:Decoder:Graph(step-6)

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)
```

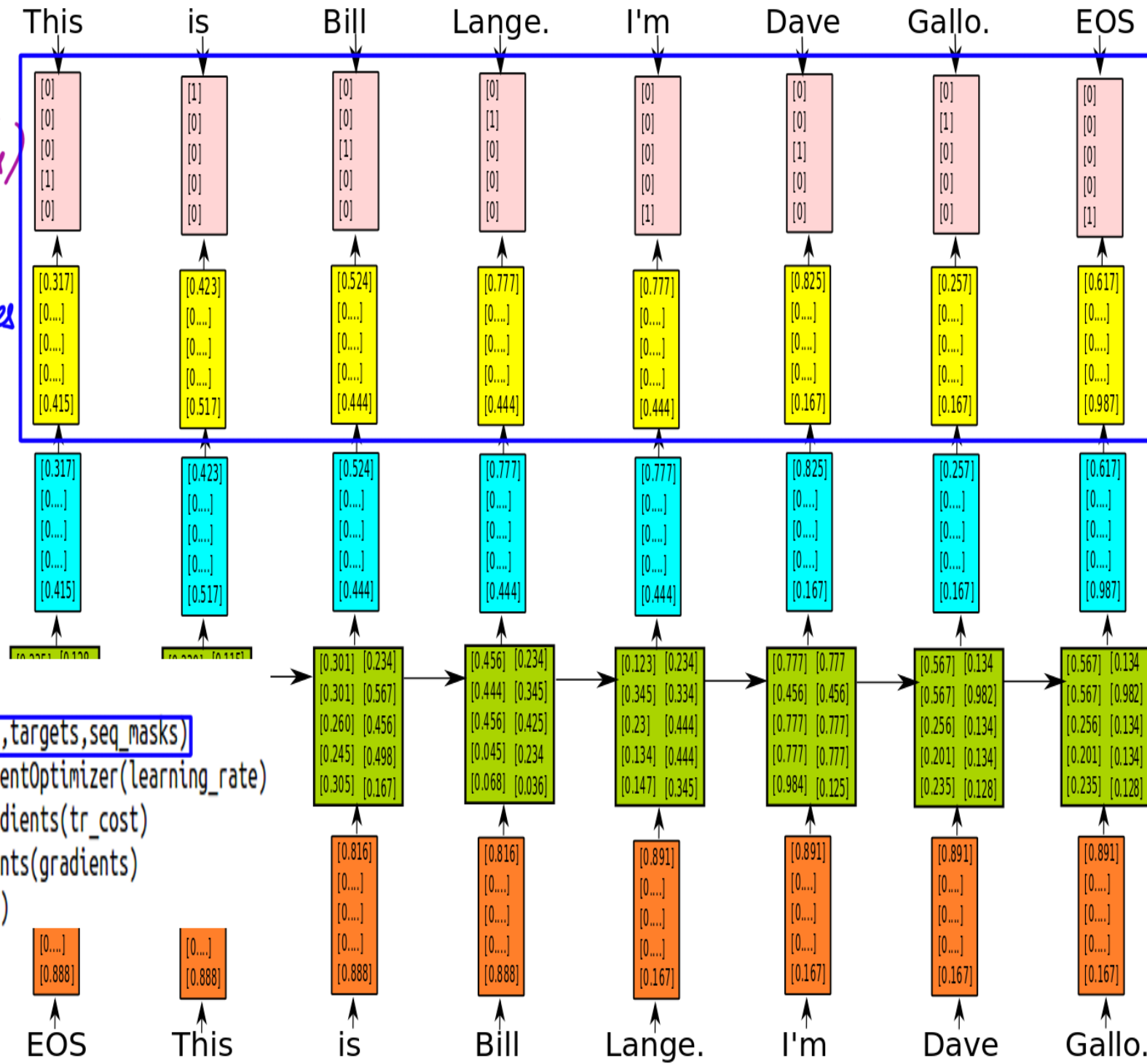
1. *Softmax ($\hat{Y}$) and Loss (E) calculation are done internally. In fact this internally calls softmax loss-function.*



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

Y (Labels)
1. Calculates the loss and averages over the sequence.

Logits



with tf.name_scope("optimizer"):

```
tr_cost = sequence_loss(logits_tr, targets, seq_masks)
```

```
optimizer = tf.train.GradientDescentOptimizer(learning_rate)
```

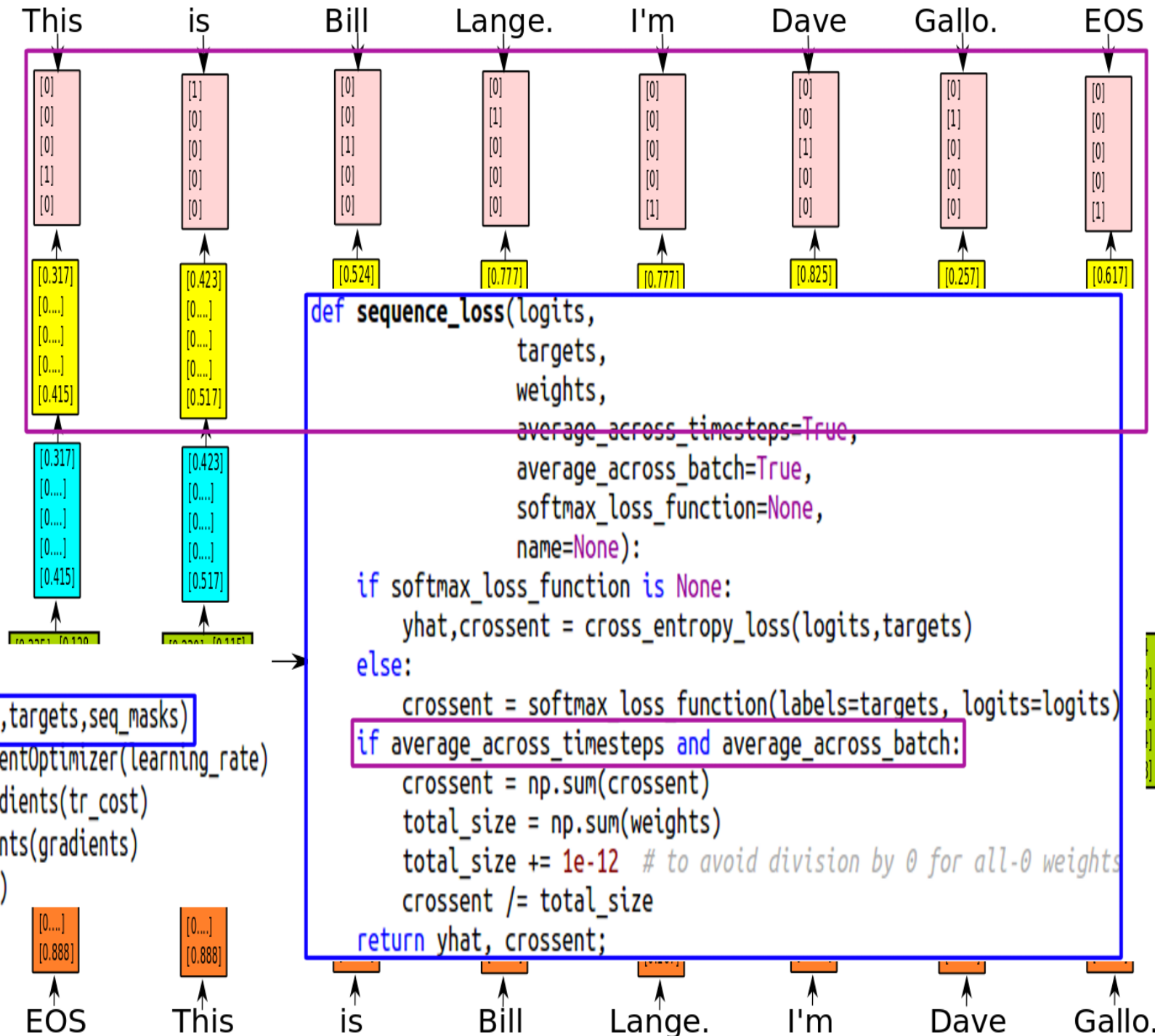
```
gradients = optimizer.compute_gradients(tr_cost)
```

```
train_op = optimizer.apply_gradients(gradients)
```

```
tf.summary.scalar("cost", tr_cost)
```


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

• Average across time steps and batch.



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)
```

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.