

Deep Learning을 이용한 개체명 분석

정상근

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구현

개체명 분석

개체명 분석

한국소비자보호원은 19일 시판중인 선물세트의 상당수가 과대 포장된

ORGANIZATION

DATE

Class 설계 (5개 Class)

- ✓ OG : Organization
- ✓ DT : Date
- ✓ TI : Time
- ✓ LC : Location
- ✓ PS : Person

개체명 분석 데이터 (1)

- <https://github.com/krikit/annie>
- 2016년 한글 및 한국어 정보처리 학술대회 : "2016년 엑소브레인 V2.0 & 울산대학교 말뭉치"라는 명칭의 CD로 배포

감성 분석 데이터 (2)

N21 Dataset

아	B-LC
프	I-LC
가	I-LC
니	I-LC
스	I-LC
탄	I-LC
의	O
	O
...	...
	O
헤	B-OG
즈	I-OG



Tab 문자로 구분 (\t)

- Train Data : ./applications/sentiment_analysis/data/ner.n2n.txt

BIO Tagging scheme for Sequential Data

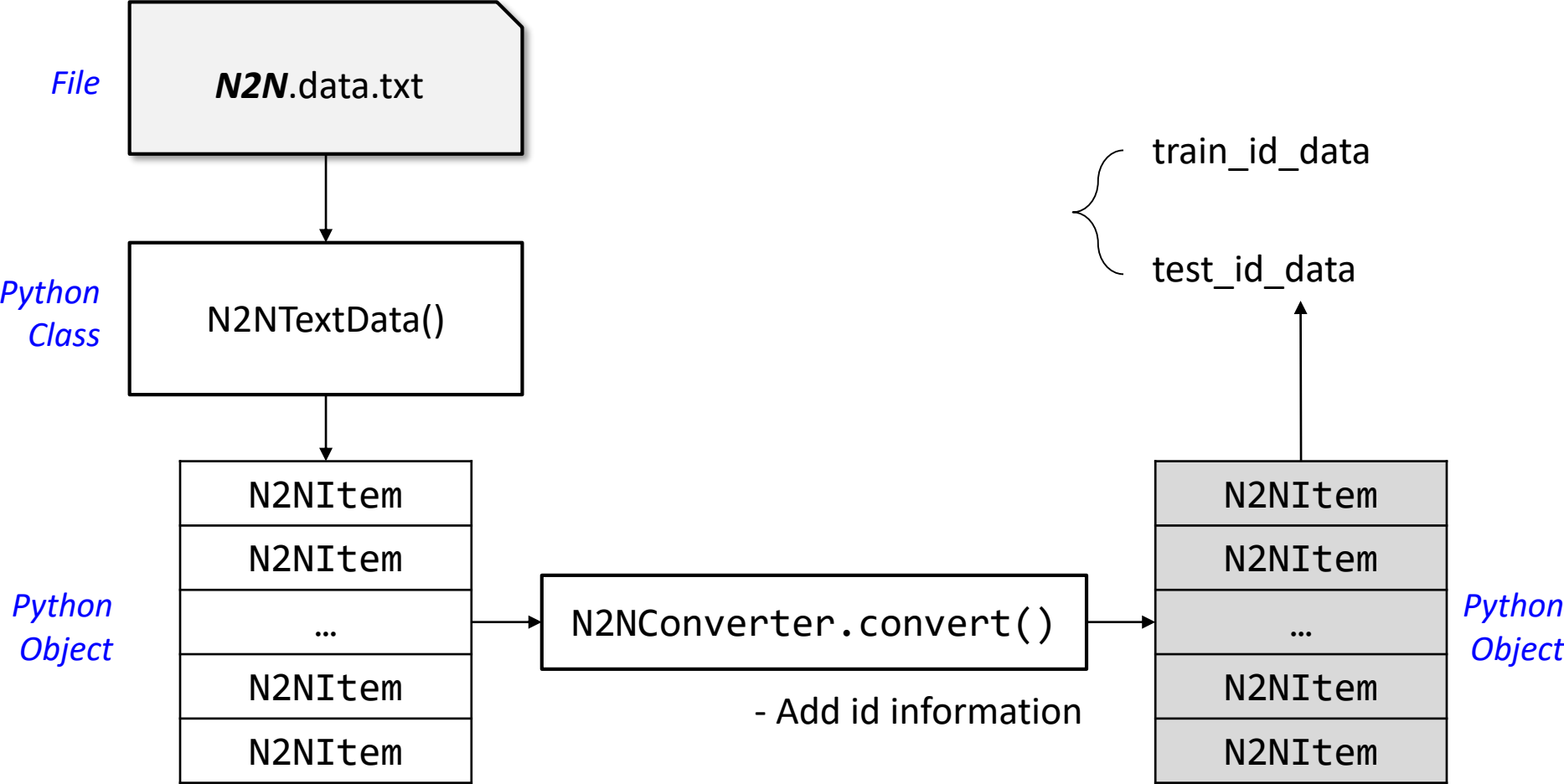
B-Tag	Beginning of the <i>tag</i>
I-Tag	Inside of the <i>tag</i>
O	Outside

지	난	달		2	7	일	부	터		매	일		오	후		4	시	에
B-DT	I-DT	I-DT	I-DT	I-DT	I-DT	I-DT	O	O	O	B-DT	I-DT	O	B-TI	I-TI	I-TI	I-TI	I-TI	O

You can use other sequential tagging schemes

- IO
- BMEWO
- BMEWO+
- BIOES
- ...

[N2N] Data Processing Flow



Code is available

https://github.com/hugman/deep_learning/tree/master/course/nlp


```
./applications./named_entity_recognition/dataset : load_data()
```

```
# vocab loader
```

```
token_vocab_fn = os.path.join( os.path.dirname(__file__), 'data',  
'token.vocab.txt')
```

```
token_vocab = Vocab(token_vocab_fn, mode='token')
```

```
target_vocab_fn = os.path.join( os.path.dirname(__file__), 'data',  
'target.vocab.txt')
```

```
target_vocab = Vocab(target_vocab_fn, mode='target')
```

```
# load train data
```

```
train_data_fn = os.path.join( os.path.dirname(__file__), 'data', 'ner.n2n.txt')
```

```
train_txt_data = N2NTextData(train_data_fn)
```

```
# convert text data to id data
```

```
train_id_data = N2NConverter.convert(train_txt_data, target_vocab, token_vocab)
```

Building Block

Step Design

**Reference
Design**

**Loss Function
Design**

Class Design

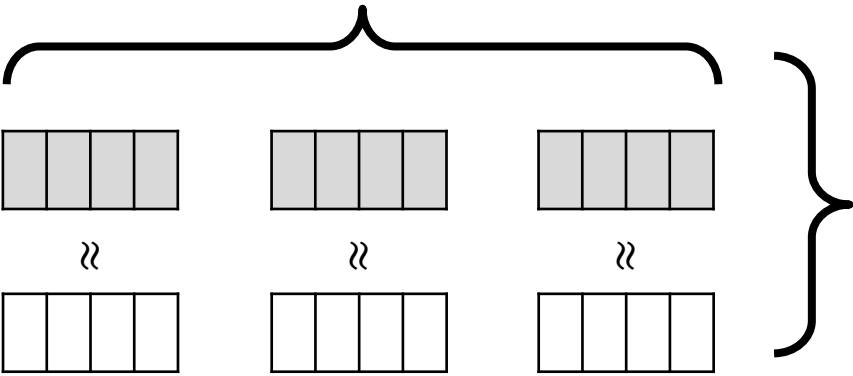
**Network
Design**

**Parameter
Updater
Design**

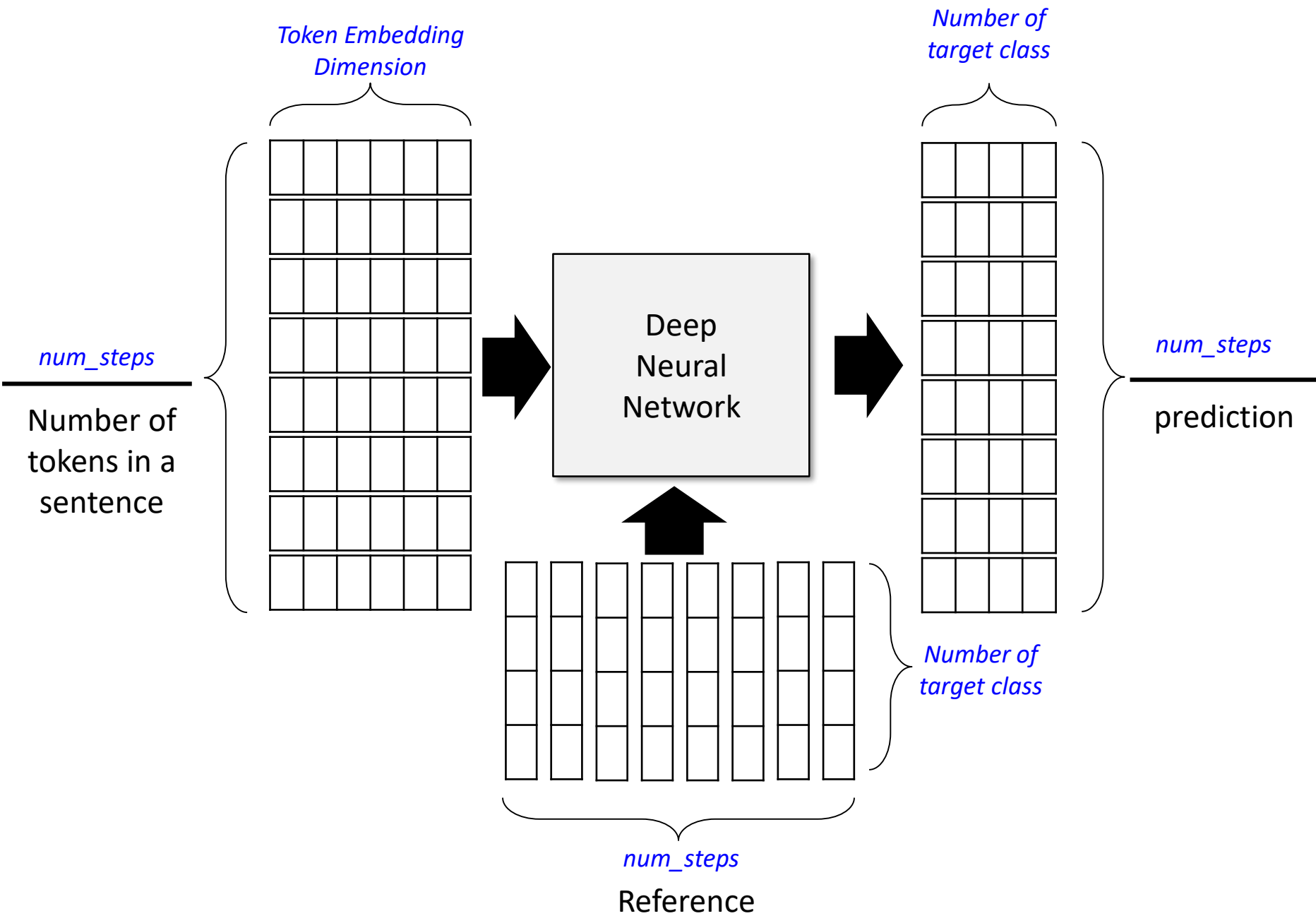
***Big & Deep
Network***

**Input
Design**

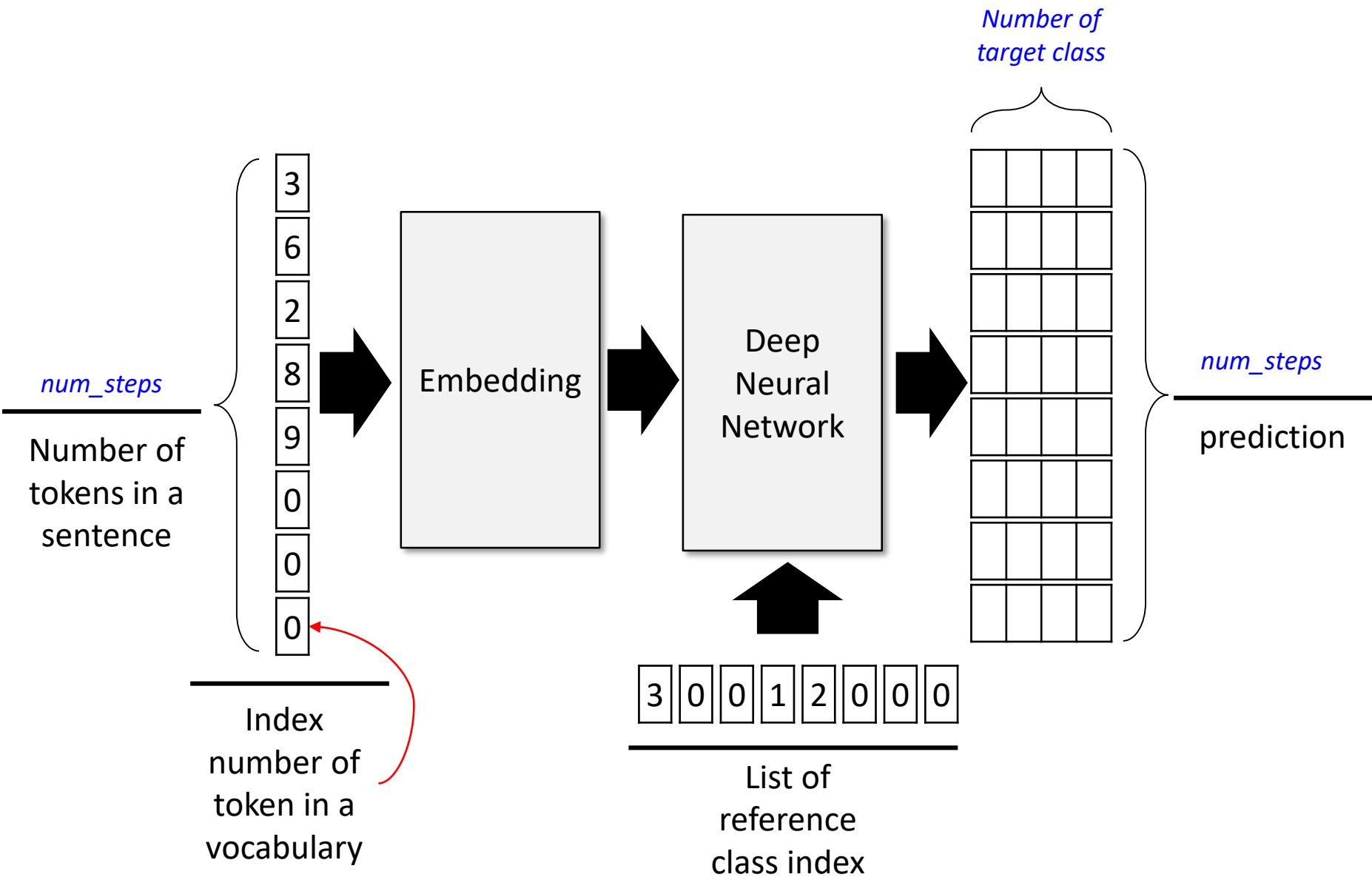
Input



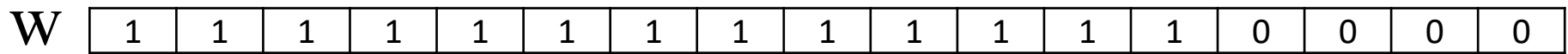
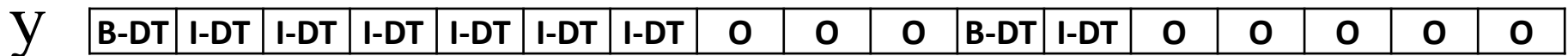
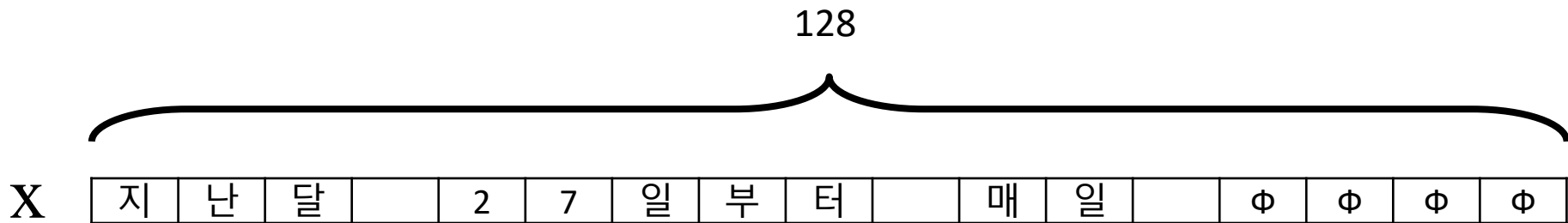
[N2N classification] DNN Interface



[N2N classification] DNN Interface with Embedding Library

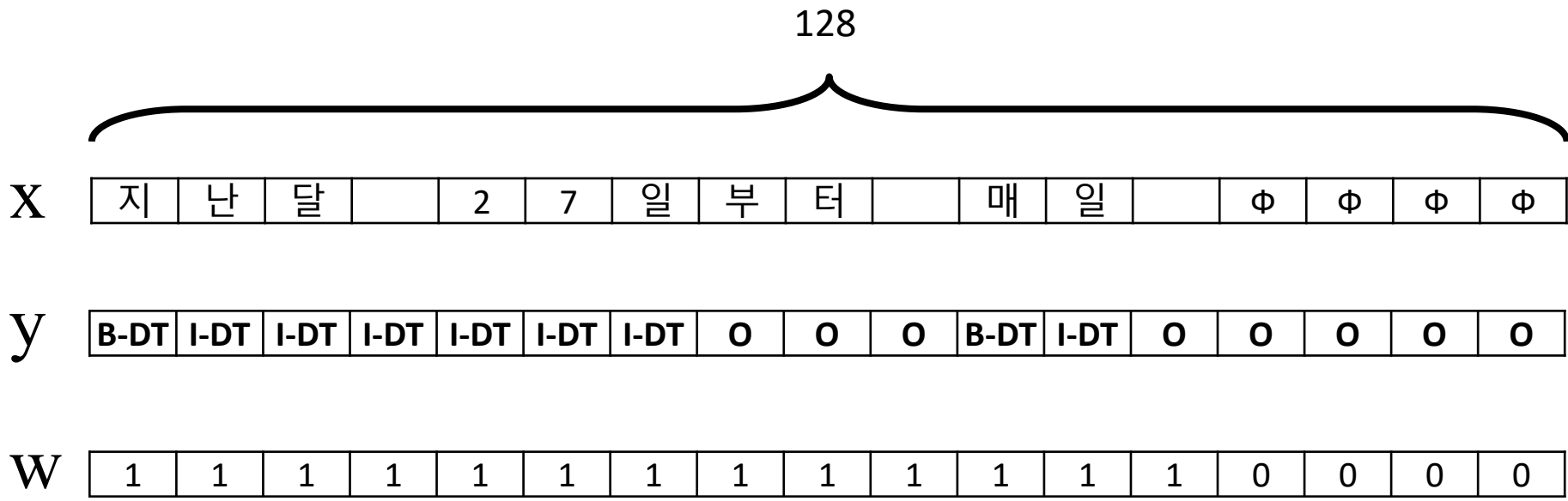


Input Design



Φ : padding symbol
W : to mark padding positions

Tensorflow Interface Implementation

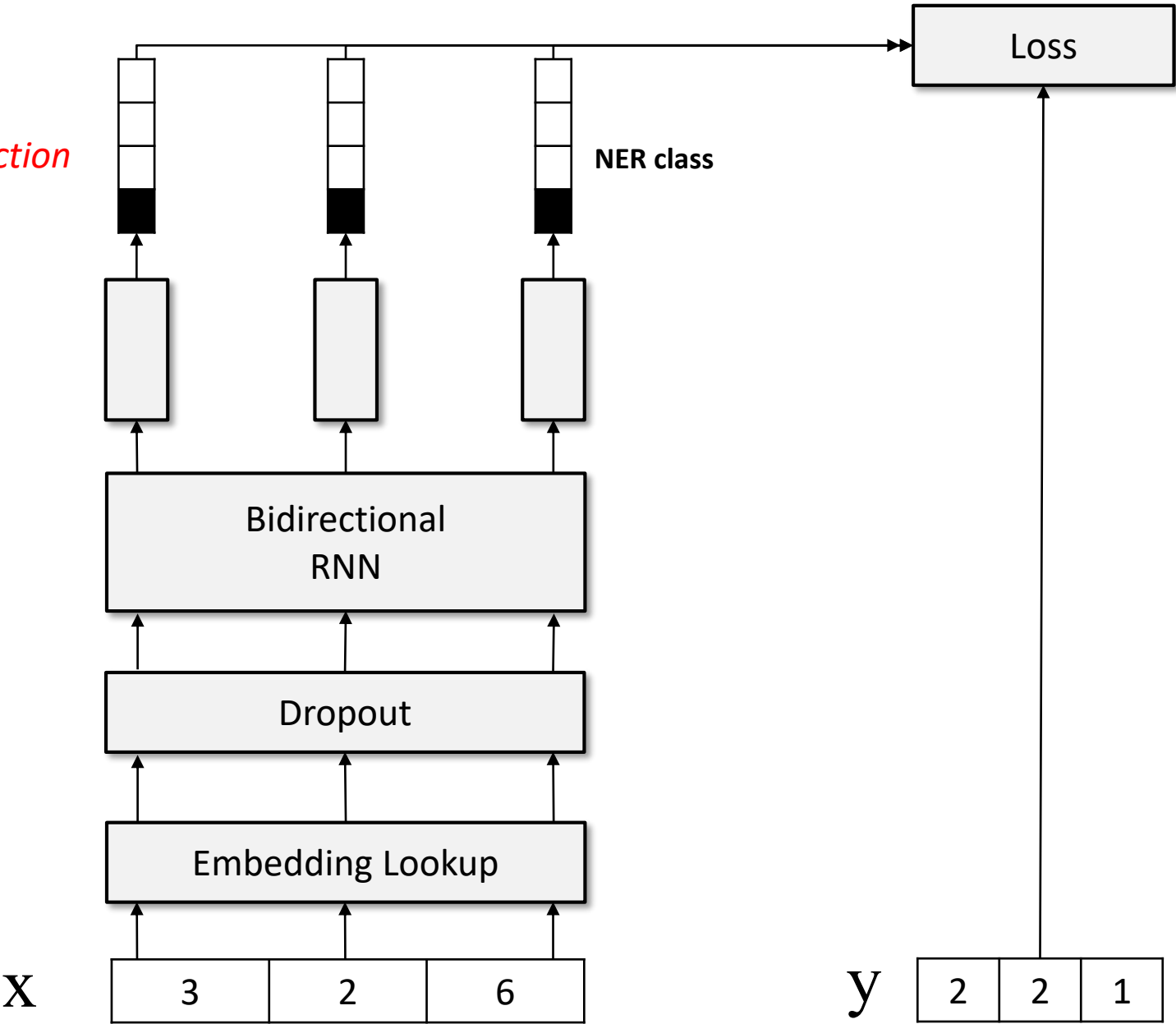


```
x = tf.placeholder(tf.int32, [None, hps.num_steps], name="pl_tokens")
y = tf.placeholder(tf.int32, [None, hps.num_steps], name="pl_target")
w = tf.placeholder(tf.int32, [None, hps.num_steps], name="pl_weight")
```

batch_size

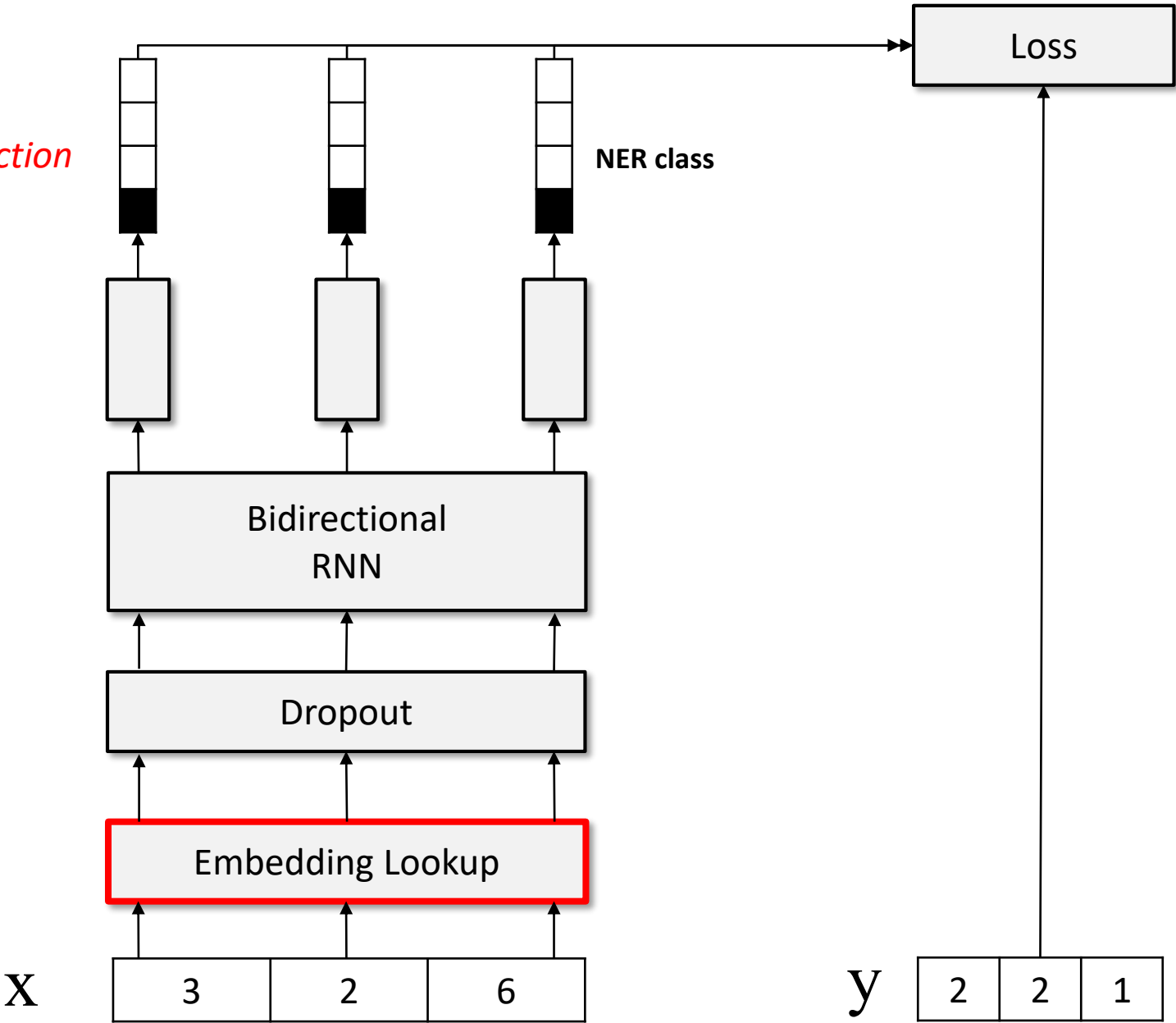
Neural Network Design

prediction



Neural Network Design | embedding

prediction



Tensorflow Implementation | Embedding

```
def _embedding(x)

    # character embedding
    shape      = [hps.vocab_size, hps.emb_size]
    initializer = tf.initializers.variance_scaling(distribution="uniform",
                                                    dtype=tf.float32)

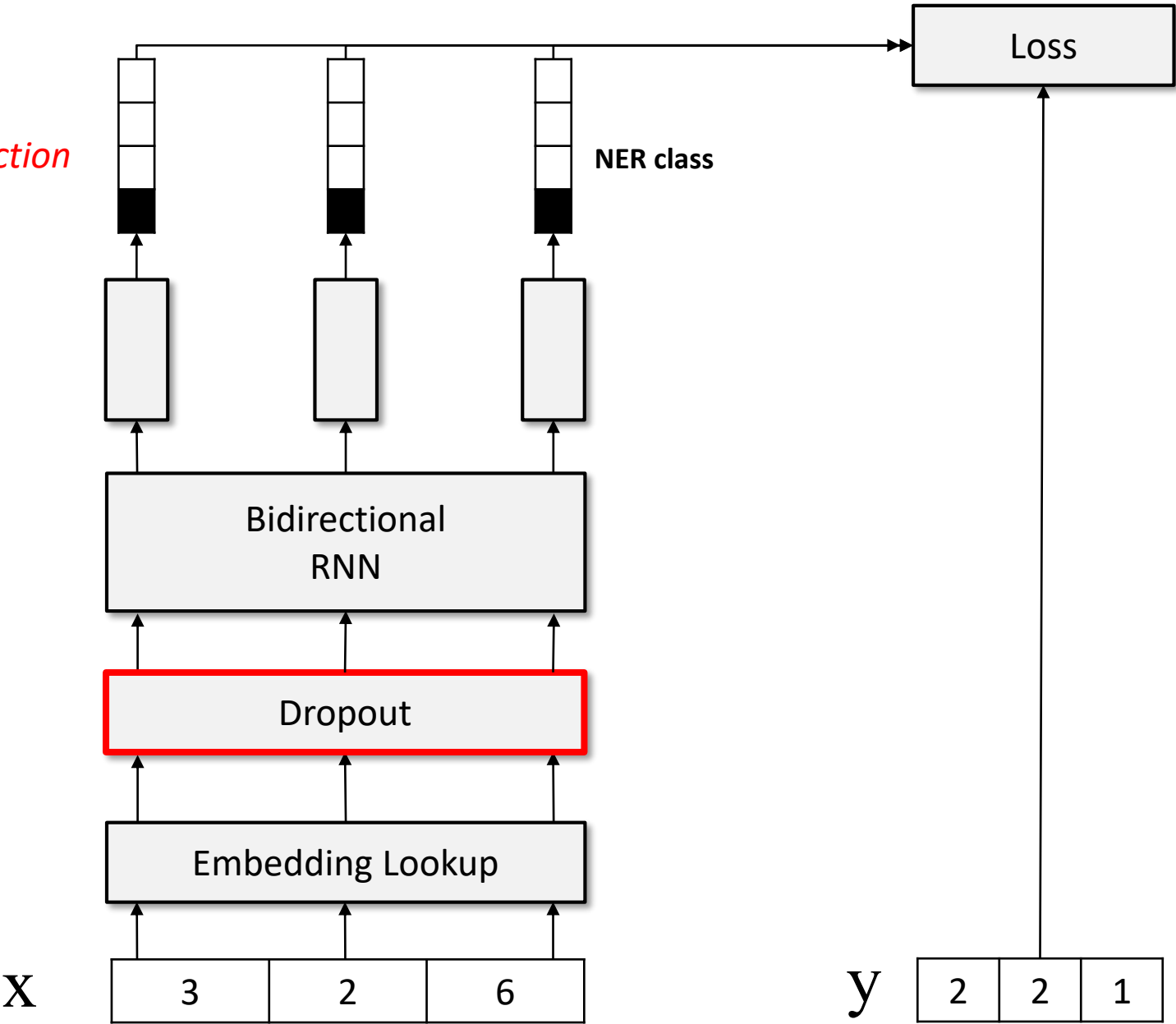
    emb_mat    = tf.get_variable("emb", shape,
                                   initializer=initializer,
                                   dtype=tf.float32)

    # [batch_size, sent_len, emb_dim]
    input_emb   = tf.nn.embedding_lookup(emb_mat, x)

    # split input_emb -> num_steps
    step_inputs = tf.unstack(input_emb, axis=1)
```

Neural Network Design

prediction



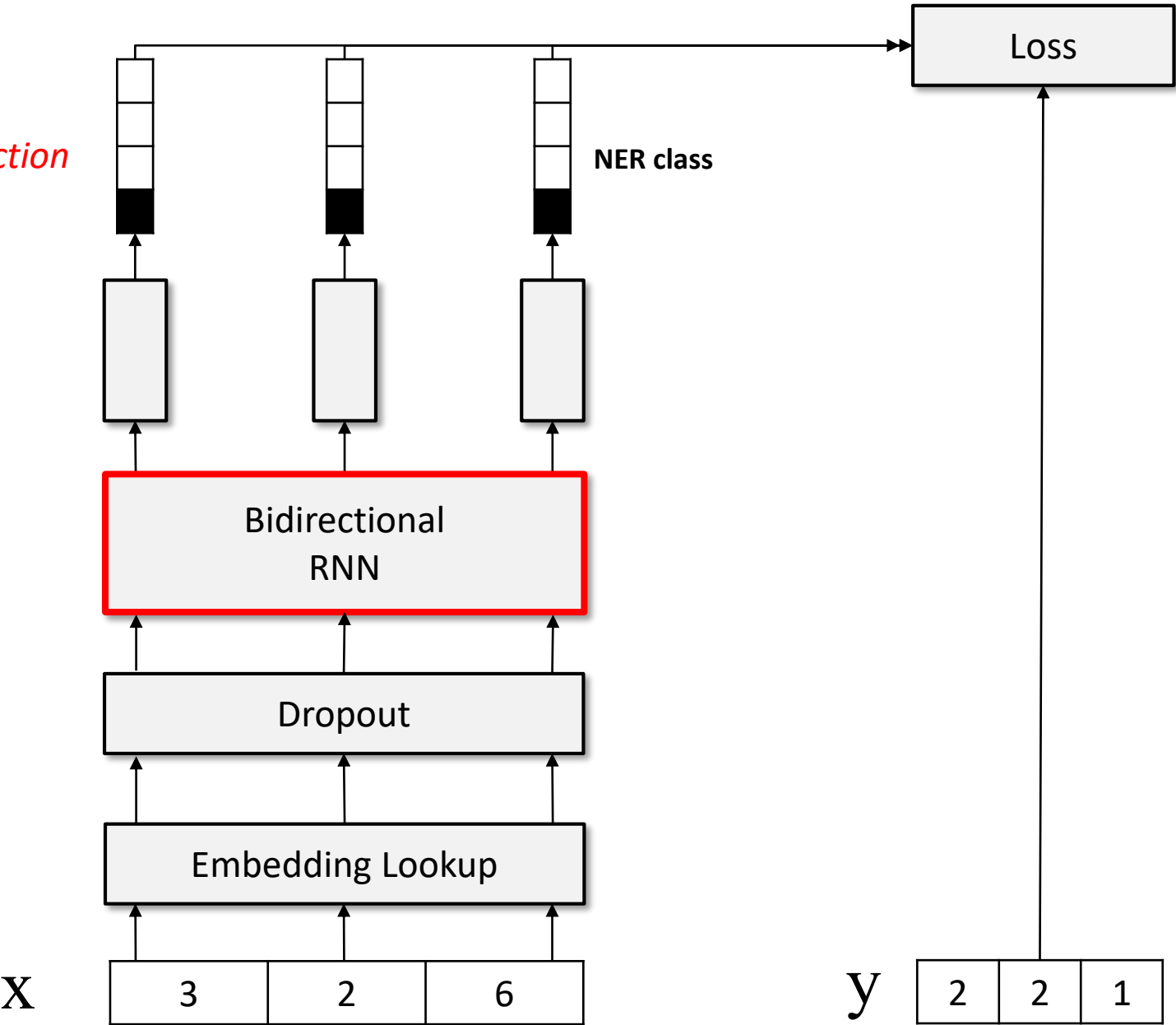
Tensorflow Implementation | Dropout

```
def _sequence_dropout(step_inputs, keep_prob)

    # apply dropout to each input
    # input : a list of input tensor which shape is [None, input_dim]
    with tf.name_scope('sequence_dropout') as scope:
        step_outputs = []
        for t, input in enumerate(step_inputs):
            step_outputs.append( tf.nn.dropout(input, keep_prob) )
    return step_outputs
```

Neural Network Design | BiRNN based N2N encoding

prediction



Tensorflow Implementation | RNN based N2N encoding | Exercise

```
def sequence_encoding_n2n(step_inputs, seq_length, cell_size)
```

Input : a list of <tf.Tensor shape=(?, 50), dtype=float32>

Return : a list of <tf.Tensor, shape=(?, 100)>

Keywords

tf.contrib.rnn.GRUCell

tf.nn.bidirectional_dynamic_rnn

tf.stack

Tensorflow Implementation | RNN based N2N encoding

```
def sequence_encoding_n2n(step_inputs, seq_length, cell_size)
    # birnn based N2N encoding
    f_rnn_cell = tf.contrib.rnn.GRUCell(cell_size, reuse=False)
    b_rnn_cell = tf.contrib.rnn.GRUCell(cell_size, reuse=False)
    _inputs     = tf.stack(step_inputs, axis=1)
    outputs, states, = tf.nn.bidirectional_dynamic_rnn(
        f_rnn_cell,
        b_rnn_cell,
        _inputs,
        sequence_length=tf.cast(seq_length, tf.int64),
        time_major=False,
        dtype=tf.float32,
        scope='birnn',
    )

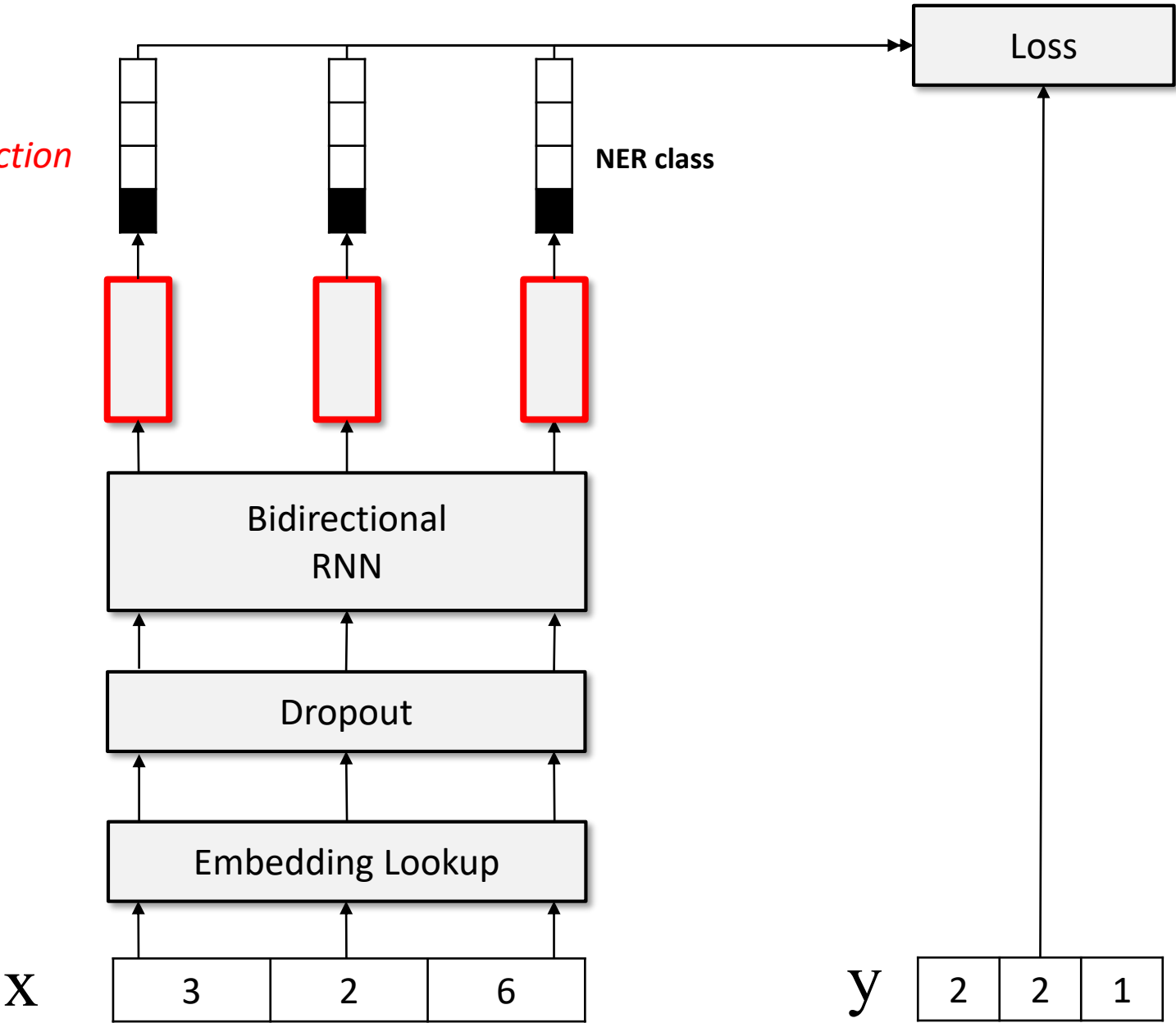
    output_fw, output_bw = outputs
    states_fw, states_bw = states

    output          = tf.concat([output_fw, output_bw], 2)
    step_outputs    = tf.unstack(output, axis=1)

    final_state     = tf.concat([states_fw, states_bw], 1)
    return step_outputs
```

Neural Network Design | to class N2N

prediction



Tensorflow Implementation | to class | Exercise

```
def _to_class(step_inputs, num_class)
```

Input : a list of <tf.Tensor shape=(?, 200) dtype=float32>

Return : a list of <tf.Tensor, shape=(?, 11), dtype=float32)>

Keywords

tensorflow.contrib.layers.python.layers.linear

Tensorflow Implementation | to class

```
def _to_class(step_inputs, num_class)
```

```
    T = len(step_inputs)
    step_output_logits = []
    for t in range(T):
        # encoder to linear(map)
        out = step_inputs[t]
        if t==0: out = linear(out, num_class, scope="Rnn2Target")
        else:    out = linear(out, num_class, scope="Rnn2Target", reuse=True)
        step_output_logits.append(out)
    return step_output_logits
```

```
# output of the neural network
```

```
# step_preds and step_out_probs
step_out_probs = []
step_out_preds = []
for _output in step_outputs:
    _out_probs = tf.nn.softmax(_output)
    _out_pred  = tf.argmax(_out_probs, 1)

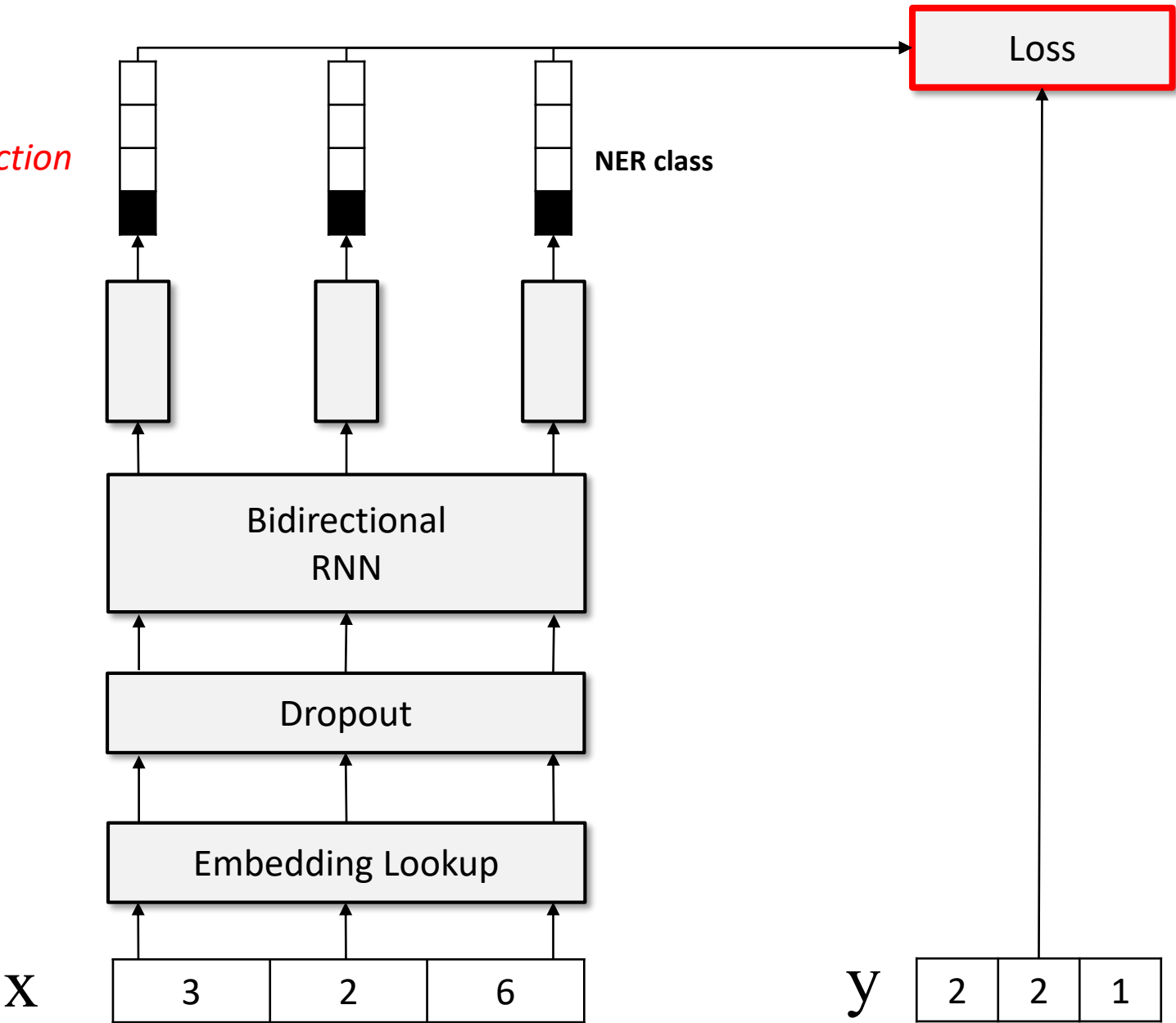
    step_out_probs.append(_out_probs)
    step_out_preds.append(_out_pred)
```

```
# stack for interface
```

```
self.step_out_probs = tf.stack(step_out_probs, axis=1, name="step_out_probs")
self.step_out_preds = tf.stack(step_out_preds, axis=1, name="step_out_preds")
```

Neural Network Design | Loss calculation

prediction



Tensorflow Implementation | sequencewise loss calculation | Exercise

```
def _loss(step_outputs, step_refs, weights)
```

Input step_outputs:

a list of <tf.tensor, shape=(?, 11), dtype=float32>

Input step_refs:

a list of <tf.tensor, shape=(?, 128), dtype=int32>

Return : tf.Tensor, shape=()

Keywords

tensorflow.contrib.seq2seq.sequence_loss

Tensorflow Implementation | sequencewise loss calculation

```
def _to_class_n2n(step_inputs, num_class)
    # step_outputs : a list of [batch_size, num_class] float32 - unscaled logits
    # step_refs     : [batch_size, num_steps] int32
    # weights       : [batch_size, num_steps] float32
    # calculate sequence wise loss function using cross-entropy
    _batch_output_logits = tf.stack(step_outputs, axis=1)
    loss = sequence_loss(
        logits=_batch_output_logits,
        targets=step_refs,
        weights=weights
    )
    return loss
```

➤ `from tensorflow.contrib.seq2seq import sequence_loss`

Tensorflow Implementation | parameter update

```
# optimizer settings
```

```
optimizer      = tf.train.AdamOptimizer(hps.learning_rate)
self.train_op  = optimizer.minimize(self.loss,
                                   global_step=self.global_step)
```

➤ ***train_op*** should be called outside of the network to update parameters

Training Process

```
# training process
while not sv.should_stop():
    fetches = [model.global_step, model.loss, model.train_op]
    a_batch_data = next( train_data_set.iterator )
    y, x, w = a_batch_data
    fetched = sess.run(fetches, {
                                model.x: x,
                                model.y: y,
                                model.w: w,
                                model.keep_prob: hps.keep_prob,
                                }
    )

    local_step += 1

    _global_step = fetched[0]
    _loss        = fetched[1]
```

Q/A

감사합니다.

Lecture Blog: www.hugman.re.kr

Lecture Video: <https://goo.gl/7NL5hV>

Lecture Slides: <https://goo.gl/6NfR1V>

Code Share: <https://github.com/hugman>

Facebook: <https://goo.gl/1RML3C>

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