

# 法律声明

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## LessonTwo:自然语言处理 主讲: 赵英俊

# 课时大纲

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- 自然语言处理理论基础
- 用到的高级API详解
- 项目和代码设计思路
- 手把手开始编程

# 自然语言处理理论基础

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计算  $P(w_1, w_2, \dots, w_n)$

马尔可夫(Markov)假设：

“无记忆性”：未来的事件，只取决于有限的历史

$$P(w_5 | w_4, w_3, w_2, w_1)$$

$$P(w_5)$$

unigram

$$P(w_5 | w_4)$$

bigram

$$P(w_5 | w_4, w_3)$$

trigram

$$P(w_1, w_2, \dots, w_n) = P(w_1)P(w_2 | w_1) \dots P(w_n | w_{n-1})$$

$$P(w_1, w_2, \dots, w_n) = P(w_1 | START)P(w_2 | w_1) \dots P(w_n | w_{n-1})P(EOS | w_n)$$

# 自然语言处理理论基础

## □ 优化目标函数

- Batch Gradient Descent
  - 必须遍历所有的训练数据才更新一次
- Stochastic Gradient Descent
  - 每看见一个数据点就更新，非常不稳定
- Mini-batch Gradient Descent Mini-batch梯度下降法
  1. 初始化参数
    - $\Theta = \text{uniform}(d)$
  2. 随机抽取m个数据点  $T_m = \{(x_i, y_i) | i = k_1, \dots, k_m\}$ , 计算偏导数
    - $\nabla \Theta = \frac{\partial \text{Obj}(\Theta, T_m)}{\partial \Theta} = \frac{\partial}{\partial \Theta} \sum_{i=k_1}^{k_m} \text{Obj}(\Theta, x_i, y_i)$
  3. 更新参数
    - $\Theta = \Theta - \eta \nabla \Theta$
  4. 适当的条件更新learning rate  $\eta$ , 返回2, 直到收敛

# 自然语言处理理论基础

## □ Recurrent Neural Network 循环神经网络

- $\frac{\partial J}{\partial W_t} = \frac{\partial J}{\partial h_{t+n}} \frac{\partial h_{t+n}}{\partial h_{t+n-1}} \cdots \frac{\partial h_{t+1}}{\partial h_t} \frac{\partial h_t}{\partial W_t}$
- $\frac{\partial h_{t+n}}{\partial h_{t+n-1}} \cdots \frac{\partial h_{t+1}}{\partial h_t} = \prod_{i=t}^{t+n-1} \frac{\partial \tanh(a_i)}{\partial a_i} W$

- 假设W是一个一维的矩阵：
  - $|W| < 1, \prod_{i=t}^{t+n-1} W \rightarrow 0$ 
    - Vanishing Gradients (梯度消失)
  - $|W| > 1, \prod_{i=t}^{t+n-1} W \rightarrow \infty$ 
    - Exploding Gradients (梯度爆炸)
- 如果W是高维的矩阵呢？
  - $|W| = \lambda_{max}(W)$ , 最大的特征值

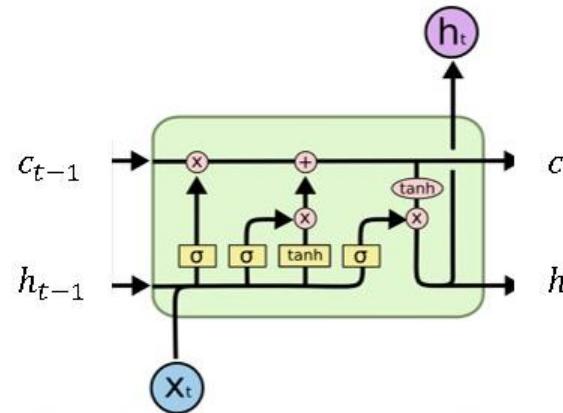
# 自然语言处理理论基础

## □ Long Short-Term Memory (LSTM)

### ■ 尝试解决的问题：Vanishing Gradients

□ Vanilla RNN:  $h_t = f(h_{t-1}, x_t)$

□ LSTM:  $h_t, c_t = f(h_{t-1}, c_{t-1}, x_t)$

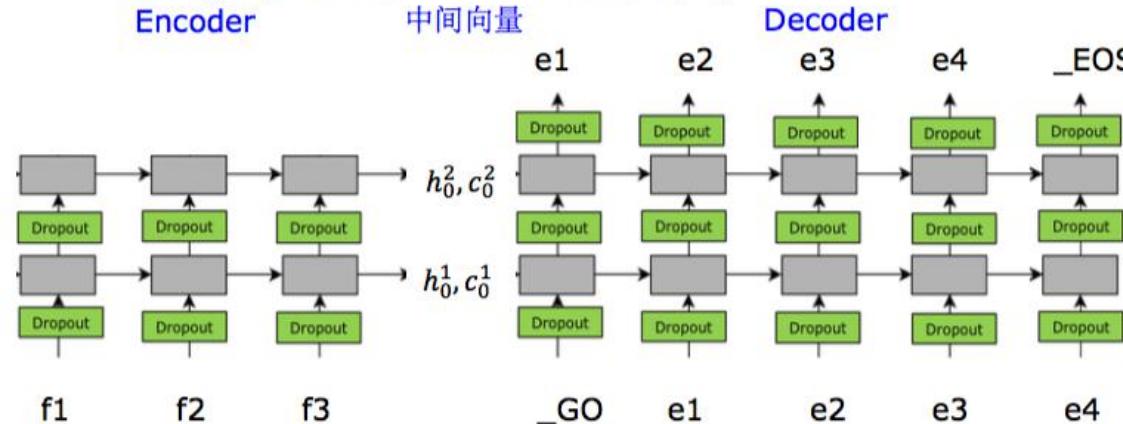


# 自然语言处理理论基础

## □ seq2seq 的结构

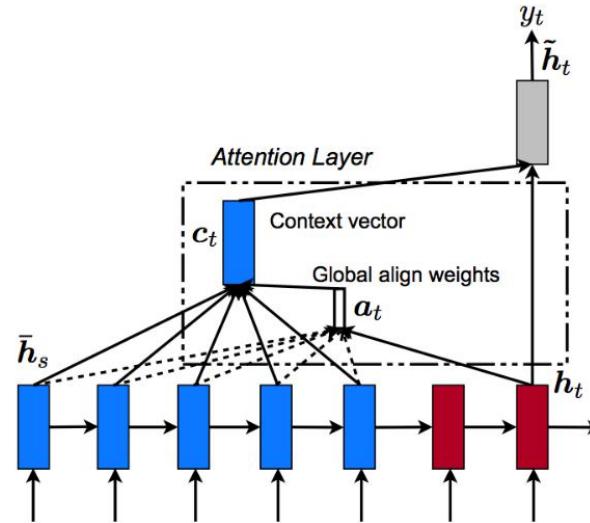
■ 核心:  $p(e_1^N | f_1^M) = \prod_{i=1}^N p(e_i | e_1^{i-1}, f_1^M)$

■  $h_0, c_0$  由另外一个LSTM来计算

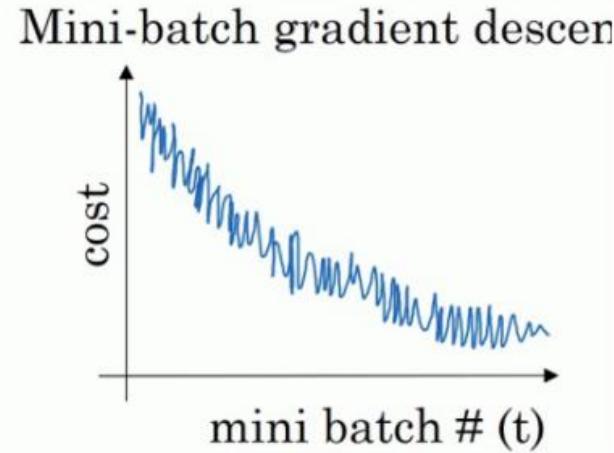
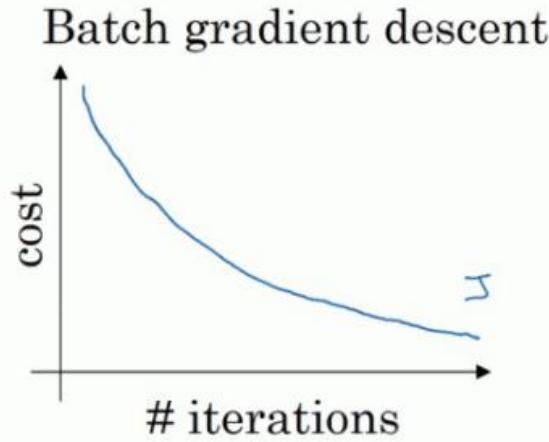


# 自然语言处理理论基础

## □ Attention



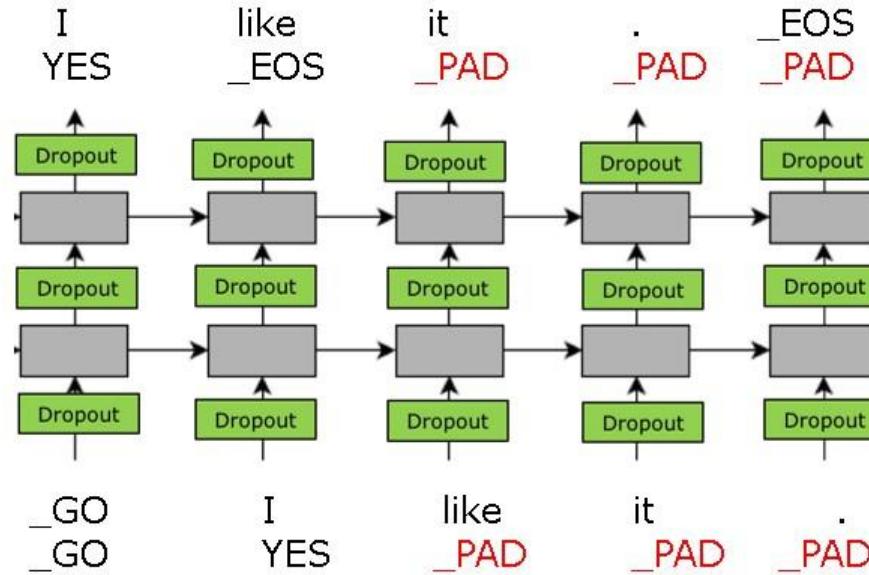
# 自然语言处理理论基础



# 自然语言处理理论基础

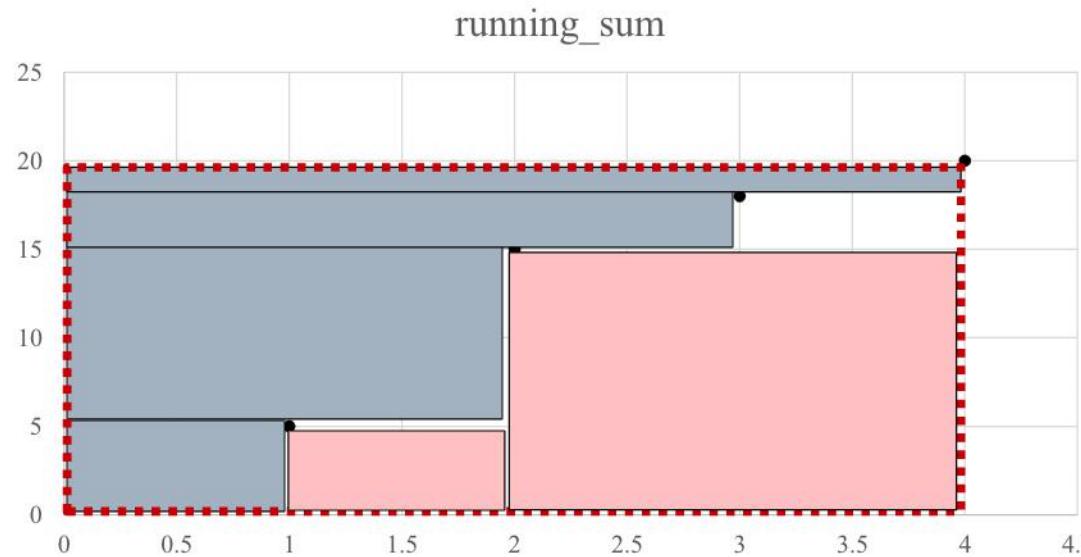
## □ mini-batch在RNN上问题

■ 句子的长度不一样：增加padding



# 自然语言处理理论基础

□ buckets = [2,4,1]; 最好的buckets = [1,2,4]



# 用到的高级API详解

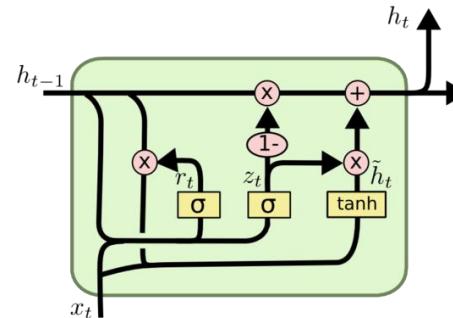
`tf.contrib.rnn.GRUCell()`

Args:

- `num_units`
- `activation`
- `reuse`
- `kernel_initializer`
- `bias_initializer`
- `name`
- `dtype`

`tf.contrib.rnn.BasicLSTMCell()`

- `num_units`
- `forget_bias`
- `state_is_tuple`
- `activation`
- `reuse`
- `name`
- `dtype`

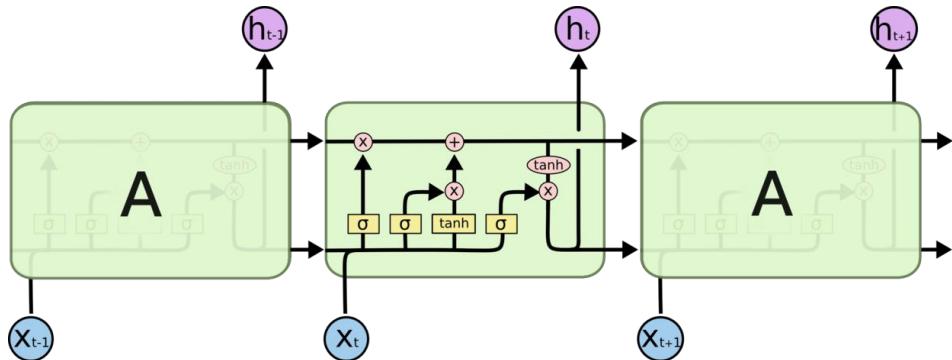


$$z_t = \sigma(W_z \cdot [h_{t-1}, x_t])$$

$$r_t = \sigma(W_r \cdot [h_{t-1}, x_t])$$

$$\tilde{h}_t = \tanh(W \cdot [r_t * h_{t-1}, x_t])$$

$$h_t = (1 - z_t) * h_{t-1} + z_t * \tilde{h}_t$$



知识链接: [https://www.cnblogs.com/wangduo/p/6773601.html?utm\\_source=itdadao&utm\\_medium=referral](https://www.cnblogs.com/wangduo/p/6773601.html?utm_source=itdadao&utm_medium=referral)

# 用到的高级API详解

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## tf.contrib.rnn.MultiRNNCell()

**cells:** list of RNNCells that will be composed in this order.

**state\_is\_tuple:** If True, accepted and returned states are n-tuples, where  $n = \text{len}(\text{cells})$ .  
If False, the states are all concatenated along the column axis. This latter behavior will soon be deprecated

# 用到的高级API详解

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## tf.contrib.legacy\_seq2seq.embedding\_attention\_seq2seq

Args:

- **encoder\_inputs**: A list of 1D int32 Tensors of shape [batch\_size].
- **decoder\_inputs**: A list of 1D int32 Tensors of shape [batch\_size].
- **cell**: tf.nn.rnn\_cell.RNNCell defining the cell function and size.
- **num\_encoder\_symbols**: Integer; number of symbols on the encoder side.
- **num\_decoder\_symbols**: Integer; number of symbols on the decoder side.
- **embedding\_size**: Integer, the length of the embedding vector for each symbol.
- **num\_heads**: Number of attention heads that read from attention\_states.
- **output\_projection**: None or a pair (W, B) of output projection weights and biases; W has shape [output\_size x num\_decoder\_symbols] and B has shape [num\_decoder\_symbols]; if provided and feed\_previous=True, each fed previous output will first be multiplied by W and added B.
- **feed\_previous**: Boolean or scalar Boolean Tensor; if True, only the first of decoder\_inputs will be used (the "GO" symbol), and all other decoder inputs will be taken from previous outputs (as in embedding\_rnn\_decoder). If False, decoder\_inputs are used as given (the standard decoder case).
- **dtype**: The dtype of the initial RNN state (default: tf.float32).
- **scope**: VariableScope for the created subgraph; defaults to "embedding\_attention\_seq2seq".
- **initial\_state\_attention**: If False (default), initial attentions are zero. If True, initialize the attentions from the initial state and attention states.

# 用到的高级API详解

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## tf.contrib.legacy\_seq2seq.model\_with\_buckets

Args:

- **encoder\_inputs**: A list of Tensors to feed the encoder; first seq2seq input.
- **decoder\_inputs**: A list of Tensors to feed the decoder; second seq2seq input.
- **targets**: A list of 1D batch-sized int32 Tensors (desired output sequence).
- **weights**: List of 1D batch-sized float-Tensors to weight the targets.
- **buckets**: A list of pairs of (input size, output size) for each bucket.
- **seq2seq**: A sequence-to-sequence model function; it takes 2 input that agree with encoder\_inputs and decoder\_inputs, and returns a pair consisting of outputs and states (as, e.g., basic\_rnn\_seq2seq).
- **softmax\_loss\_function**: Function (labels, logits) -> loss-batch to be used instead of the standard softmax (the default if this is None). **Note that to avoid confusion, it is required for the function to accept named arguments.**
- **per\_example\_loss**: Boolean. If set, the returned loss will be a batch-sized tensor of losses for each sequence in the batch. If unset, it will be a scalar with the averaged loss from all examples.
- **name**: Optional name for this operation, defaults to "model\_with\_buckets".

# 用到的高级API详解

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## tf.train.GradientDescentOptimizer(self.learning\_rate)

Args:

**learning\_rate**: A Tensor or a floating point value. The learning rate to use.

**use\_locking**: If True use locks for update operations.

**name**: Optional name prefix for the operations created when applying gradients. Defaults to "GradientDesce

## apply\_gradients()

**grads\_and\_vars**: List of (gradient, variable) pairs as returned by compute\_gradients().

**global\_step**: Optional Variable to increment by one after the variables have been updated.

**name**: Optional name for the returned operation. Default to the name passed to the Optimizer constructor.

## minimize()

Args:

**loss**: A Tensor containing the value to minimize.

**global\_step**: Optional Variable to increment by one after the variables have been updated.

**var\_list**: Optional list or tuple of Variable objects to update to minimize loss. Defaults to the list of variables collected in the graph under the key GraphKeys.TRAINABLE\_VARIABLES.

**gate\_gradients**: How to gate the computation of gradients. Can be GATE\_NONE, GATE\_OP, or GATE\_GRAPH.

**aggregation\_method**: Specifies the method used to combine gradient terms. Valid values are defined in the class AggregationMethod.

**colocate\_gradients\_with\_ops**: If True, try colocating gradients with the corresponding op.

**name**: Optional name for the returned operation.

**grad\_loss**: Optional. A Tensor holding the gradient computed for loss.

# 用到的高级API详解

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## tf.clip\_by\_global\_norm()

Args:

- **t\_list**: A tuple or list of mixed Tensors, IndexedSlices, or None.
- **clip\_norm**: A 0-D (scalar) Tensor  $> 0$ . The clipping ratio.
- **use\_norm**: A 0-D (scalar) Tensor of type float (optional). The global norm to use. If not provided, global\_norm() is used to compute the norm.
- **name**: A name for the operation (optional)

# 项目和代码设计思路

Web应用

执行器

数据预处理器

Seq2Seq\_Model

数据基本处理器

Talk is cheap, Show me the code

# 问答互动

在所报课的课程页面，

- 1、点击“全部问题”显示本课程所有学员提问的问题。
- 2、点击“提问”即可向该课程的老师和助教提问问题。



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# 小象学院——互联网新技术在线教育领航者