

# VQA

## 视觉问答机器人

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七月在线 加号

微博：@翻滚吧\_加号

# 目录

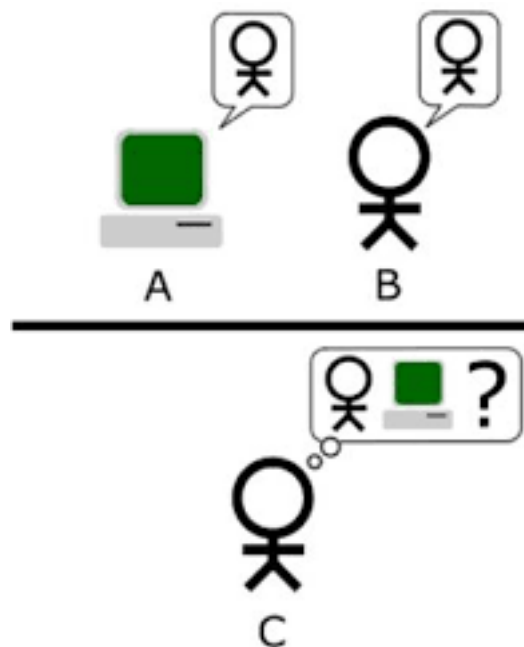
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- 决策树 Decision Tree
  - 介绍
  - 熵 Entropy
  - 信息增益 Information Gain
  - 常见算法
  - 过度拟合 Overfitting
    - 剪枝 Prune
    - 增益率 GainRatio
  - 更多的DT应用场景类别：
    - 连续函数，多分类，回归
- 决策树的究极进化 Ensemble
  - Bagging
  - Random Forest
  - Boosting
    - GBDT
    - XGBoost



# VQA的起源

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source: <https://zh.wikipedia.org/wiki/%E5%9B%BE%E7%81%B5%E6%B5%8B%E8%AF%95>

# VQA的起源

## Computer AI passes Turing test in 'world first'

🕒 9 June 2014 | Technology

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source: <http://www.bbc.co.uk/news/technology-27762088>

# VQA的起源

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## 新图灵测试的探讨

Malinowski, M. and Fritz, M., 2014. Towards a visual turing challenge. *arXiv preprint arXiv:1410.8027*.  
<https://arxiv.org/abs/1410.8027>

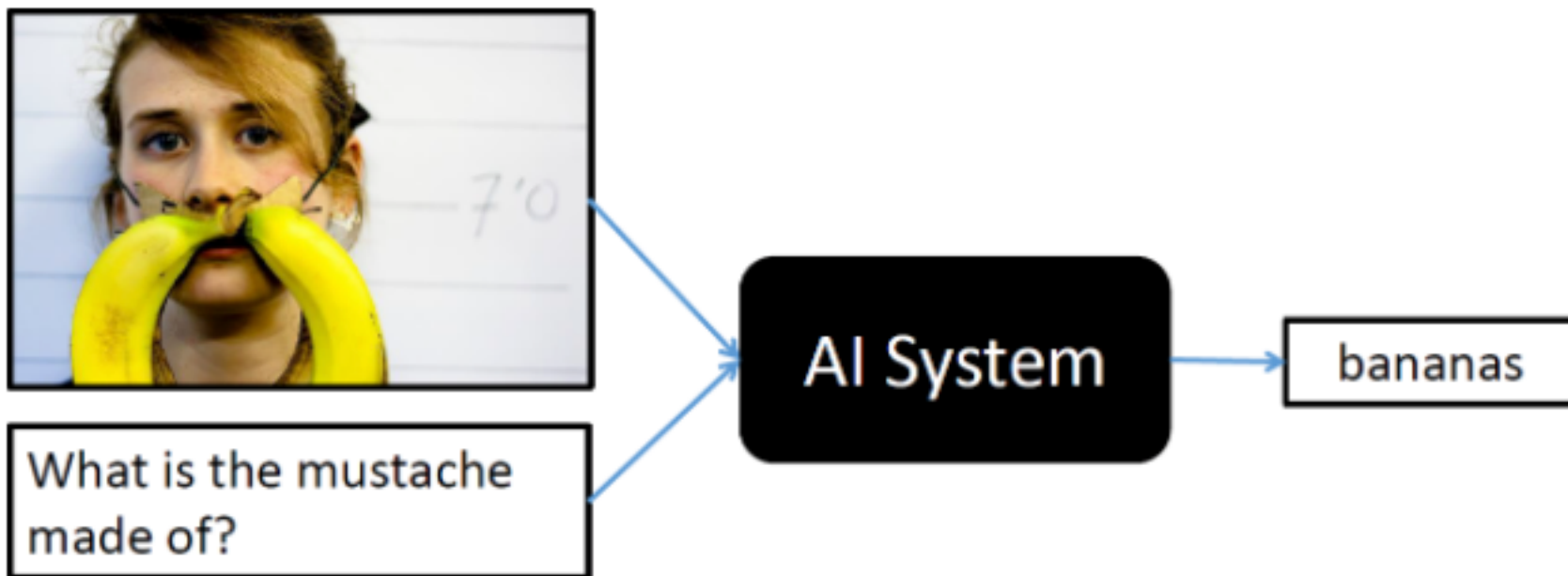
Geman, D., Geman, S., Hallonquist, N. and Younes, L., 2015. Visual turing test for computer vision systems. *Proceedings of the National Academy of Sciences*, 112(12), pp.3618-3623.  
<http://www.pnas.org/content/112/12/3618.abstract>



# VQA的起源



<http://www.visualqa.org/>



# VQA的起源

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主要涉及：

NLP

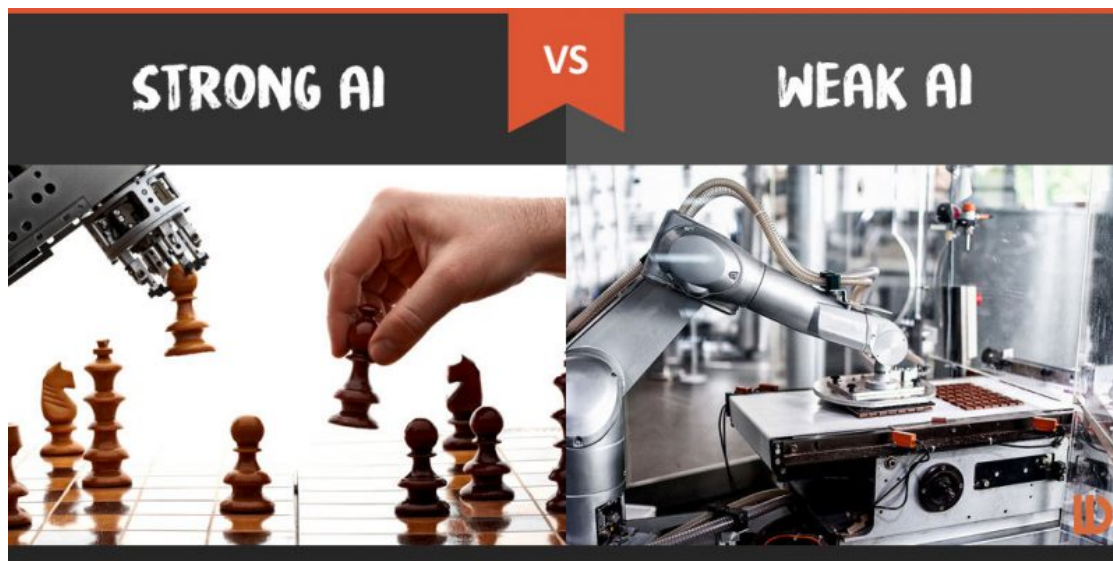
Computer Vision

“common-sense” reasoning

# VQA模型

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第一步 生成答案

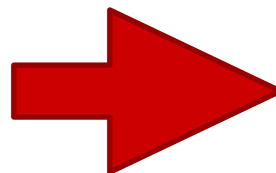




# VQA模型

## 第一步 生成答案

Antol, S., Agrawal, A., Lu, J., Mitchell, M., Batra, D., Lawrence Zitnick, C. and Parikh, D., 2015. Vqa: Visual question answering. In *Proceedings of the IEEE International Conference on Computer Vision* (pp. 2425-2433). <https://arxiv.org/pdf/1505.00468v6.pdf>

[illegible]

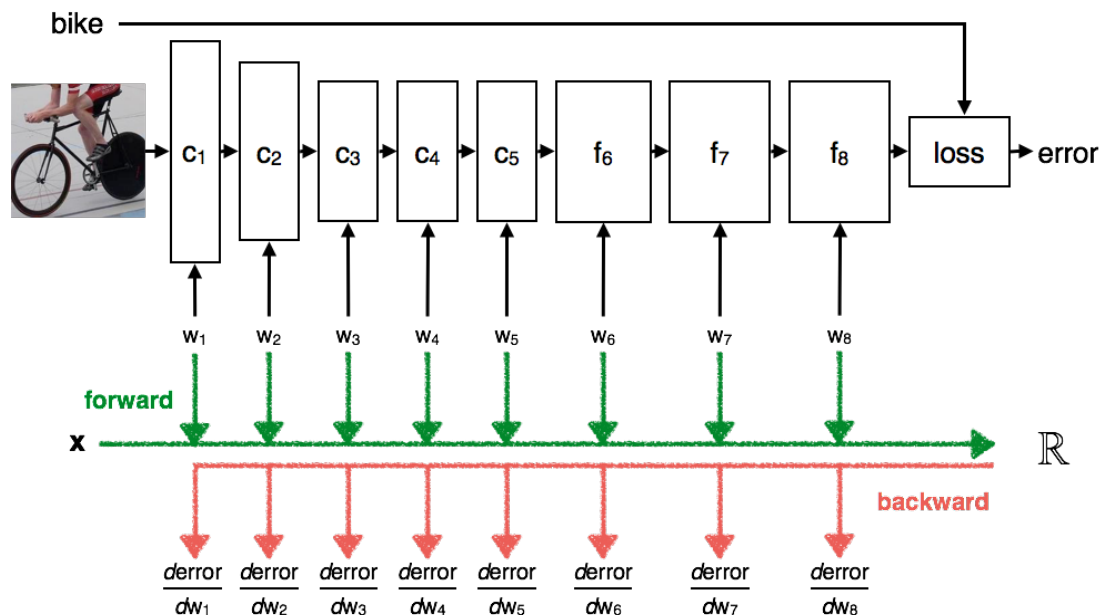
# VQA模型

第二步 处理输入源数据：图片

CNN:

VGG-16(Visual Geometry Group, Uni of Oxford)

<http://www.robots.ox.ac.uk/~vgg/>



# VQA模型

## VGG-16的标准构造 (keras)

```
def VGG_16(weights_path=None):
    model = Sequential()
    model.add(ZeroPadding2D((1,1),input_shape=(3,224,224)))
    model.add(Convolution2D(64, 3, 3, activation='relu'))
    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(64, 3, 3, activation='relu'))
    model.add(MaxPooling2D((2,2), strides=(2,2)))

    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(128, 3, 3, activation='relu'))
    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(128, 3, 3, activation='relu'))
    model.add(MaxPooling2D((2,2), strides=(2,2)))

    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(256, 3, 3, activation='relu'))
    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(256, 3, 3, activation='relu'))
    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(256, 3, 3, activation='relu'))
    model.add(MaxPooling2D((2,2), strides=(2,2)))

    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(512, 3, 3, activation='relu'))
    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(512, 3, 3, activation='relu'))
    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(512, 3, 3, activation='relu'))
    model.add(MaxPooling2D((2,2), strides=(2,2)))

    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(512, 3, 3, activation='relu'))
    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(512, 3, 3, activation='relu'))
    model.add(ZeroPadding2D((1,1)))
    model.add(Convolution2D(512, 3, 3, activation='relu'))
    model.add(MaxPooling2D((2,2), strides=(2,2)))

    model.add(Flatten())
    model.add(Dense(4096, activation='relu'))
    model.add(Dropout(0.5))
    model.add(Dense(4096, activation='relu'))
    model.add(Dropout(0.5))
    model.add(Dense(1000, activation='softmax'))

    if weights_path:
        model.load_weights(weights_path)

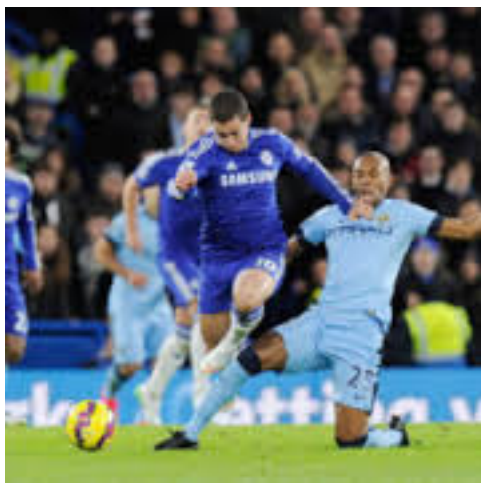
    return model
```



# VQA模型

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第二步 处理输入源数据：图片



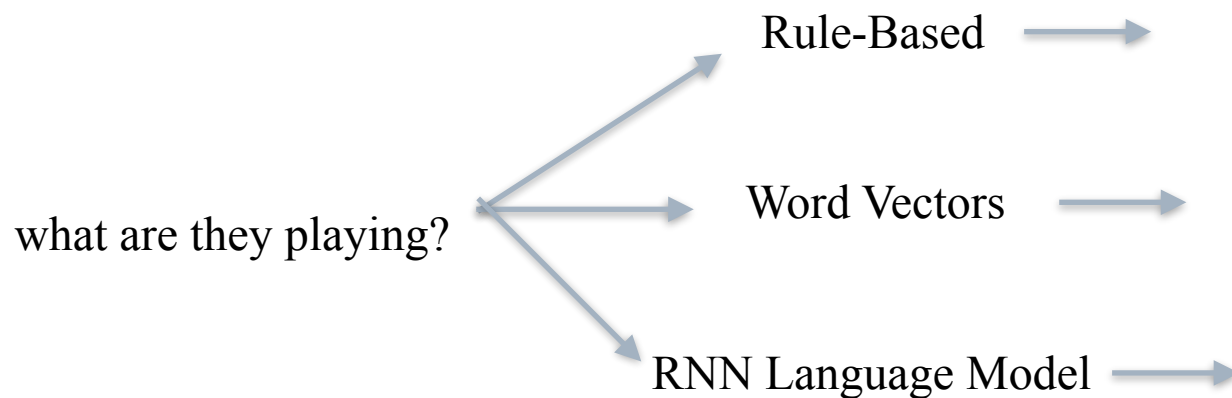
→ CNN →



# VQA模型

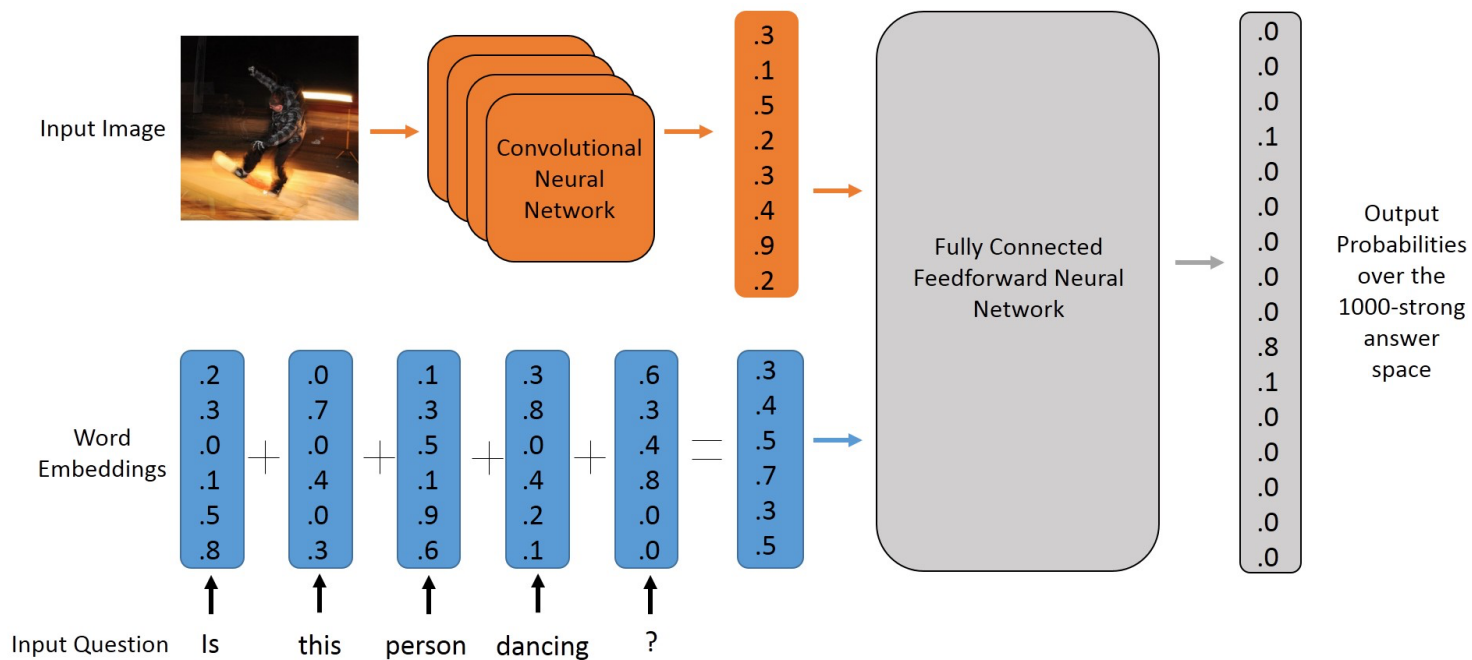
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第二步 处理输入源数据：文字



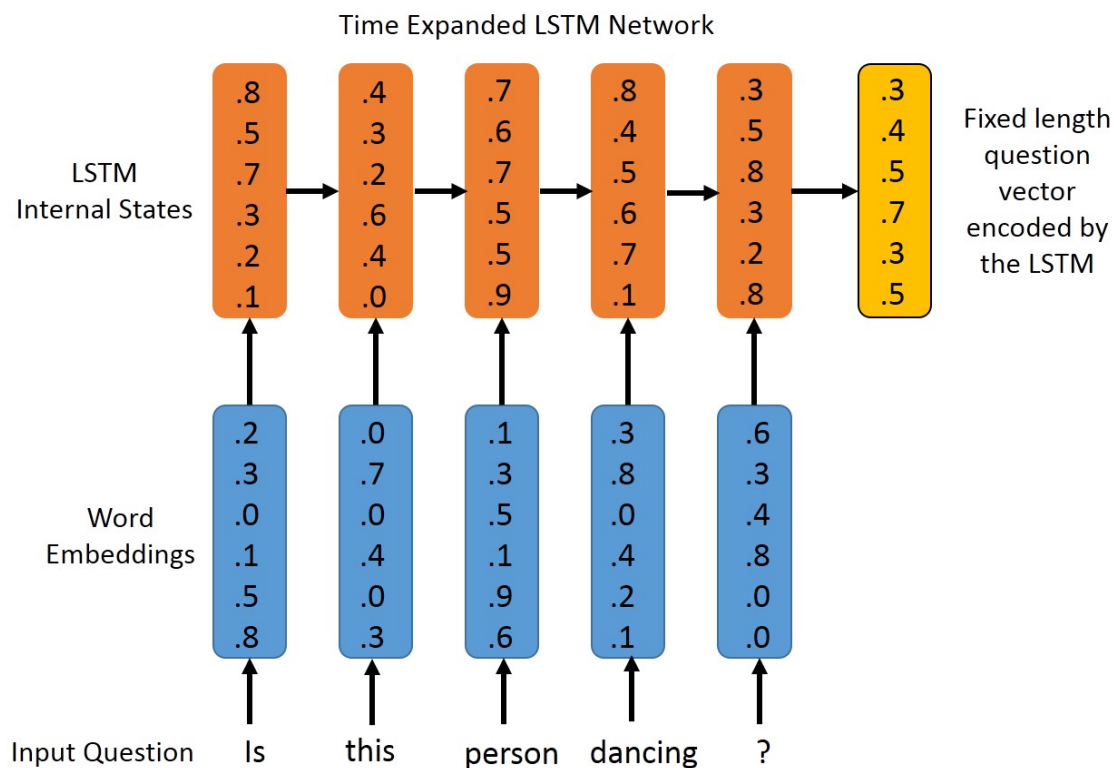
# VQA模型

### 第三部 选取VQA模型-MLP



# VQA模型

## 第三部 选取VQA模型-LSTM



# VQA模型

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# VQA实战

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**【代码见相应iPythonNotebook】**



# VQA一些开源代码参考

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<https://github.com/VT-vision-lab>

<https://github.com/abhshkdz/neural-vqa>

[https://github.com/iamaaditya/VQA\\_Demo](https://github.com/iamaaditya/VQA_Demo)

# VQA进阶发展

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## 带上注意力的模型

Lu, J., Yang, J., Batra, D. and Parikh, D., 2016. Hierarchical question-image co-attention for visual question answering. In *Advances In Neural Information Processing Systems* (pp. 289-297).  
<https://arxiv.org/abs/1606.00061>

## 『真正』的VQA（参考）

Sutskever, I., Vinyals, O. and Le, Q.V., 2014. Sequence to sequence learning with neural networks. In *Advances in neural information processing systems* (pp. 3104-3112).  
<https://arxiv.org/abs/1409.3215>



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感谢大家！

恳请大家批评指正！