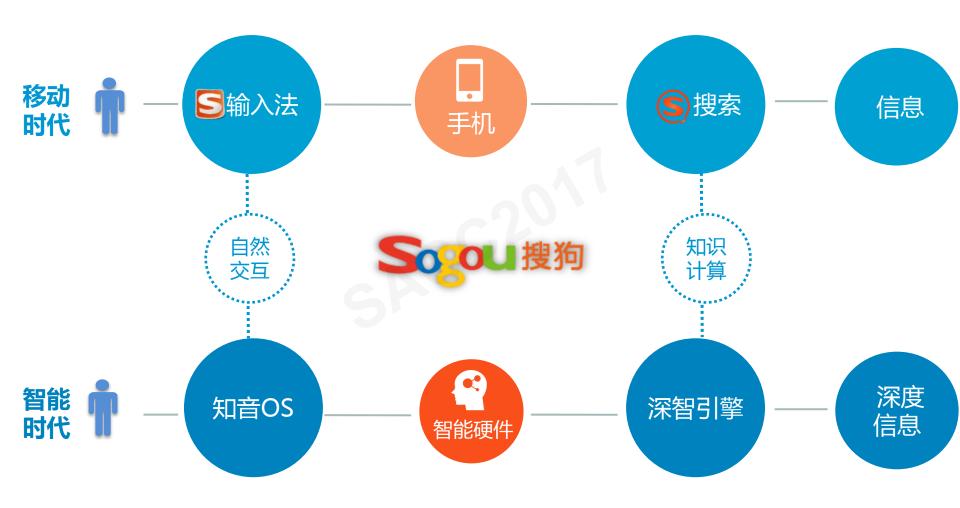


搜狗智能语音之路

搜狗语音交互技术中心 | 陈伟

从移动互联网时代迈向智能时代



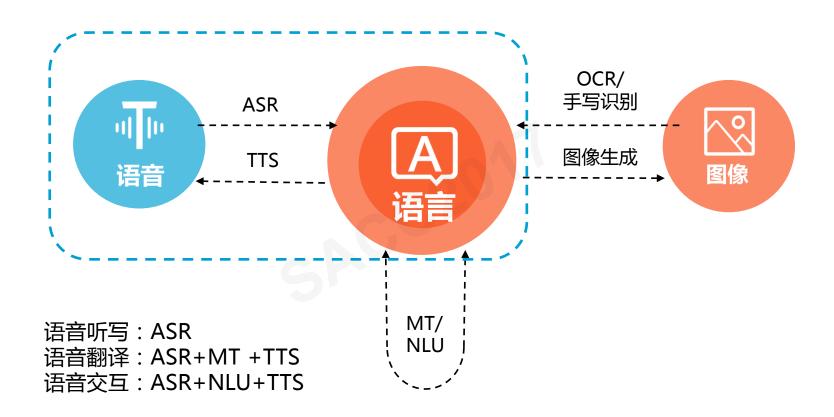






语言是人工智能的核心

语言是思想和知识的载体









语音听写技术已走向实用

语音识别可以更好提升输入/记录的效率





语音听写技术已走向实用

语音识别可以更好提升输入/记录的效率











语音听写技术落地法院庭审





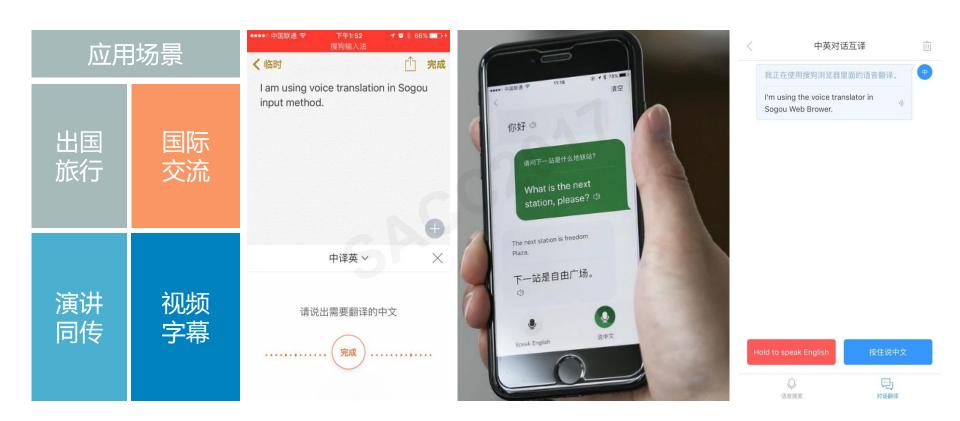






语音翻译技术逐渐可用

更好服务跨语言的交流

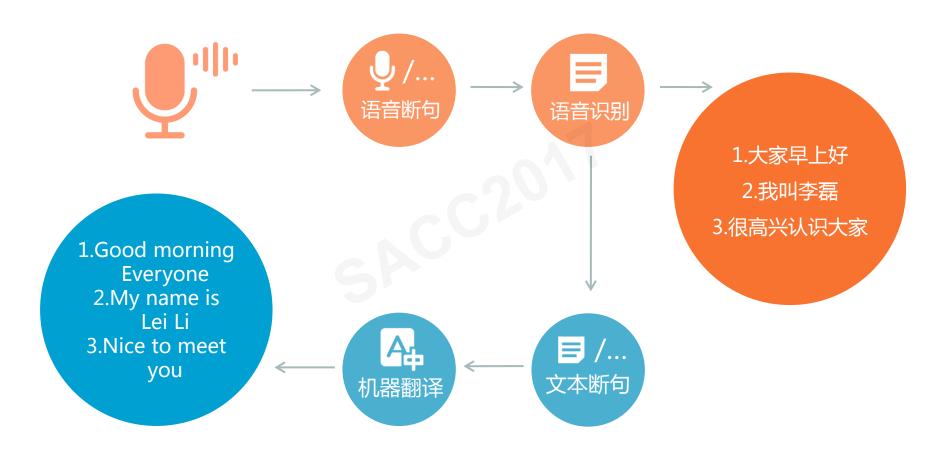








搜狗语音同传技术







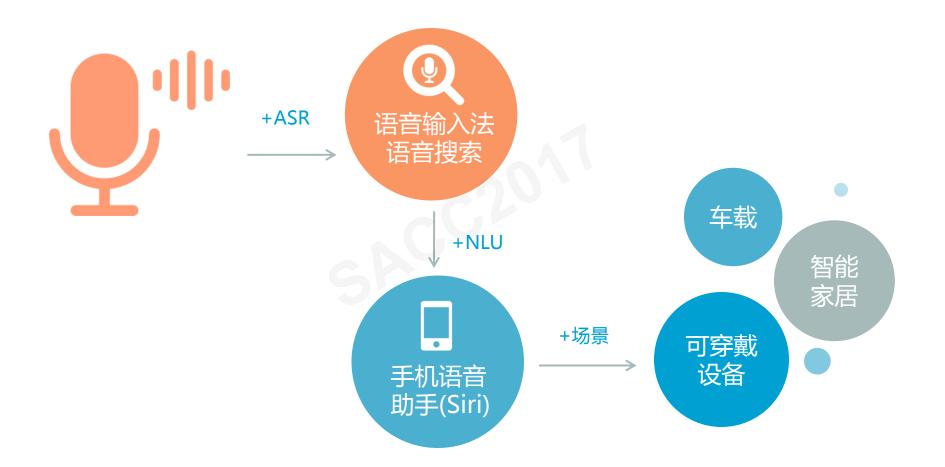
搜狗语音同传技术







语音交互产品的不断演进









针对刚需场景提供更自然的交互体验









刚需场景下的语音交互产品









车载







智能 家居











搜狗语音深度学习规模演进



超大规模的语音数据

复杂的深度学习算法

超强的运算平台

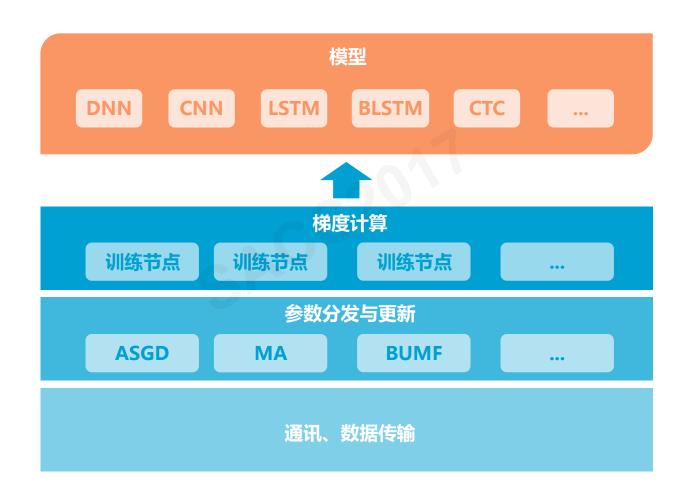








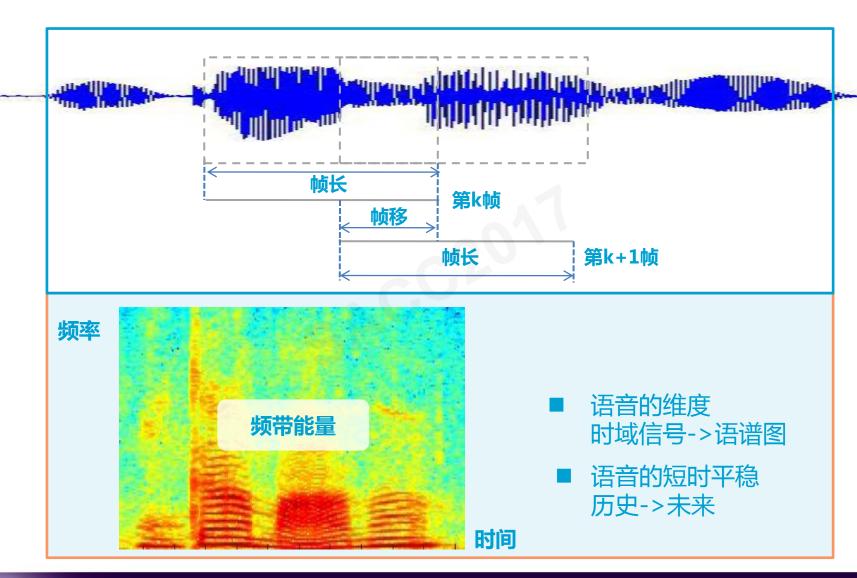
搜狗语音深度学习平台







语音是时变+短时平稳的信号







语音识别整体框架

Hand-Crafted -> Trainable 端点检测 搜狗语音 解码器 特征提取 \mathcal{W} B A N K 声学模型 语言模型 发音词典 训练 语音数据 文本数据







语音识别-贝叶斯公式

$$\arg\max_{\overrightarrow{w}} p(\overrightarrow{w} \mid \overrightarrow{\mathbf{x}}) = \arg\max_{\overrightarrow{w}} \sum_{\overrightarrow{q}} p(\overrightarrow{w}, \overrightarrow{q} \mid \overrightarrow{\mathbf{x}})$$

$$= \arg\max_{\overrightarrow{w}} \sum_{\overrightarrow{q}} \frac{p(\overrightarrow{\mathbf{x}} \mid \overrightarrow{w}, \overrightarrow{q}) P(\overrightarrow{w}, \overrightarrow{q})}{P(\overrightarrow{\mathbf{x}})}$$

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$$= \arg\max_{\overrightarrow{w}} \sum_{\overrightarrow{q}} \frac{p(\overrightarrow{\mathbf{x}} \mid \overrightarrow{w}, \overrightarrow{q}) P(\overrightarrow{q} \mid \overrightarrow{w}) P(\overrightarrow{w})}{P(\overrightarrow{\mathbf{x}})}$$

$$= \arg\max_{\overrightarrow{w}} \sum_{\overrightarrow{q}} \frac{p(\overrightarrow{\mathbf{x}} \mid \overrightarrow{q}) P(\overrightarrow{q} \mid \overrightarrow{w}) P(\overrightarrow{w})}{P(\overrightarrow{\mathbf{x}})}$$

$$= \arg\max_{\overrightarrow{w}} \sum_{\overrightarrow{q}} \frac{p(\overrightarrow{\mathbf{x}} \mid \overrightarrow{q}) P(\overrightarrow{q} \mid \overrightarrow{w}) P(\overrightarrow{w})}{P(\overrightarrow{\mathbf{x}})}$$

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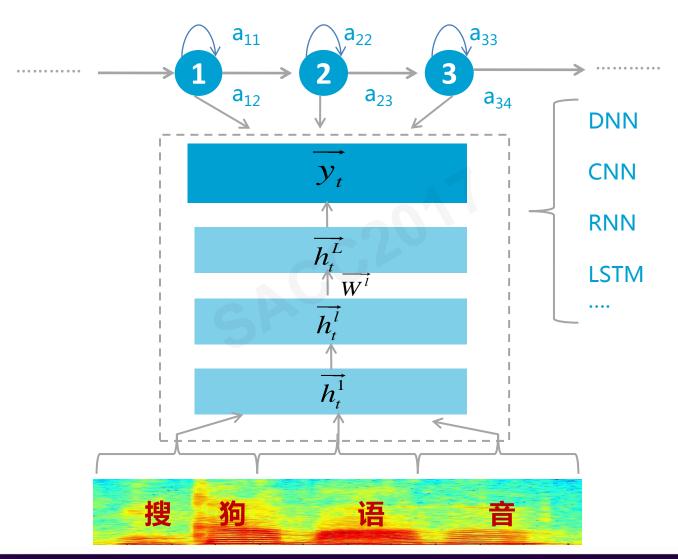
$$= \arg\max_{\overrightarrow{w}} \sum_{\overrightarrow{q}} \frac{p(\overrightarrow{\mathbf{x}} \mid \overrightarrow{q}) P(\overrightarrow{q} \mid \overrightarrow{w}) P(\overrightarrow{w})}{P(\overrightarrow{\mathbf{x}})}$$

$$= \arg\max_{\overrightarrow{q}} \sum_{\overrightarrow{q}} \frac{p(\overrightarrow{\mathbf{x}} \mid \overrightarrow{q}) P(\overrightarrow{q} \mid \overrightarrow{w}) P(\overrightarrow{w})}{P(\overrightarrow{\mathbf{x}})}$$





基于深度神经网络的声学建模

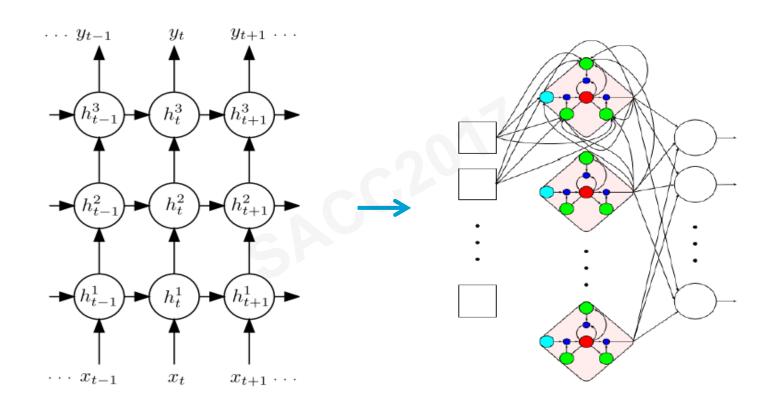






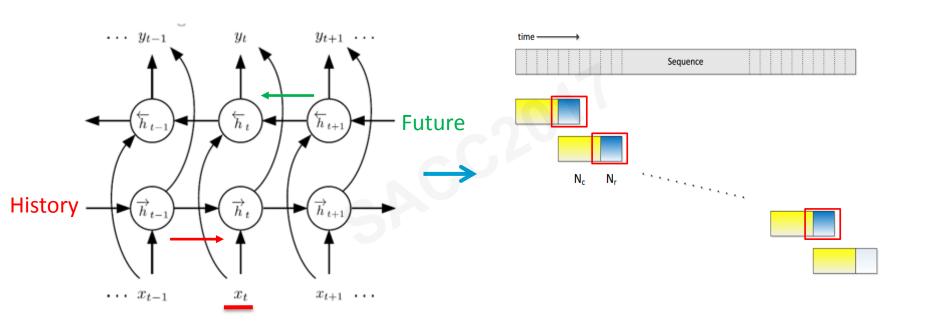


LSTM-RNN是声学模型建模的主要结构



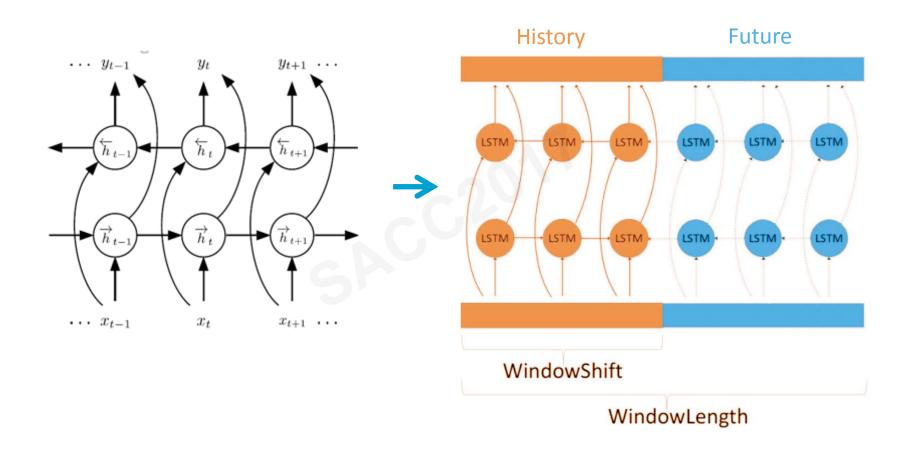


LSTM-RNN+双向特性





LSTM-RNN+双向特性

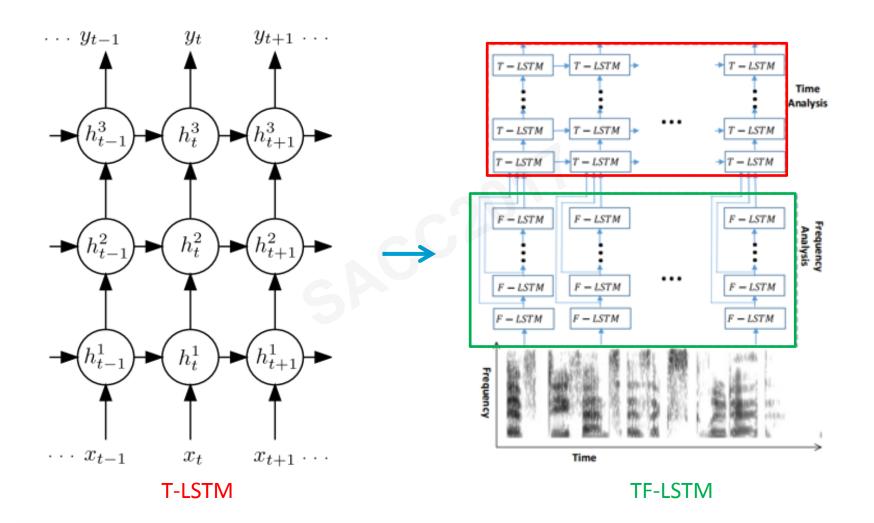






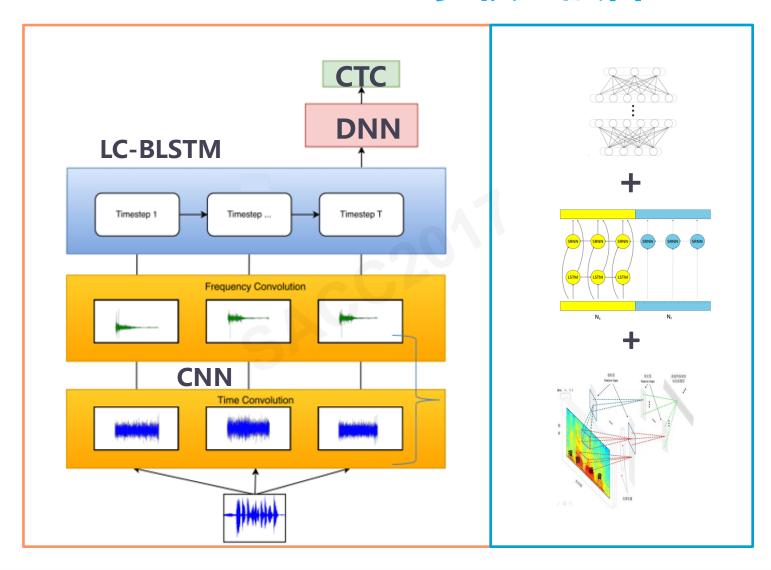


LSTM-RNN+频域扩展





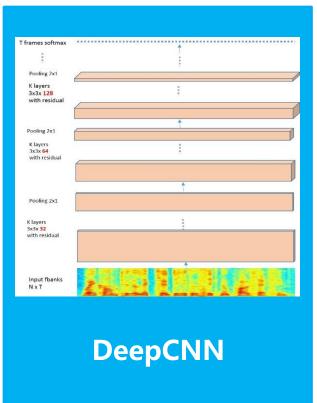
LSTM-RNN+多模型融合

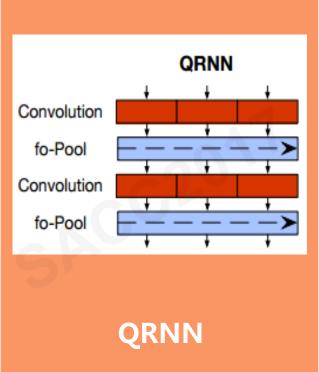


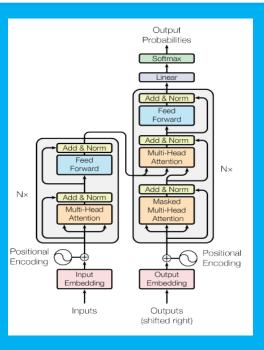




其他网络结构







Transformer-seq2seq









复杂场景下的语音识别问题仍未解决















风鸣



















如此搞笑的效果只是口音产生吗?



□ 语音识别: □音、噪声、远场

□ 语义理解: 多轮对话、纠错容错

□ 语音交互: 全双工持续交互

单点能力 -> 系统整体能力

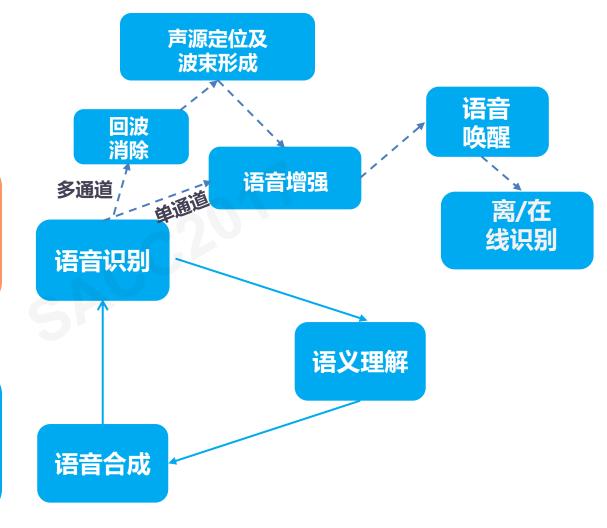








语音交互完整链路

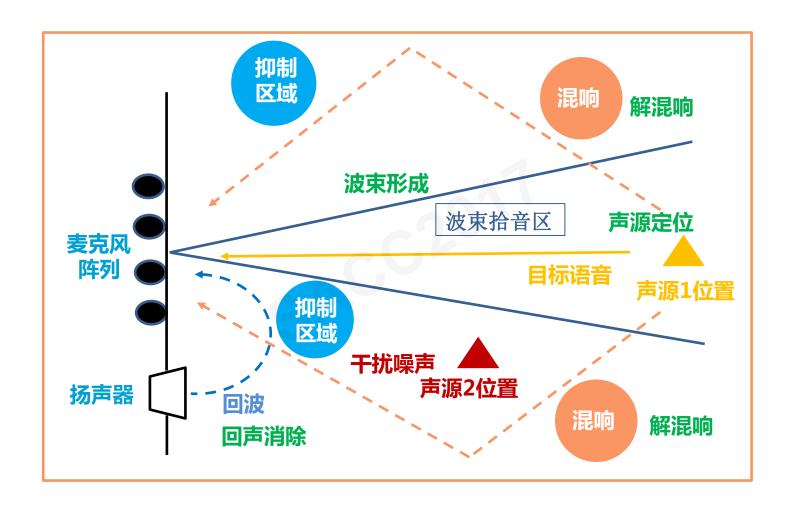








麦克风阵列技术







基于深度学习的单通道语音增强

