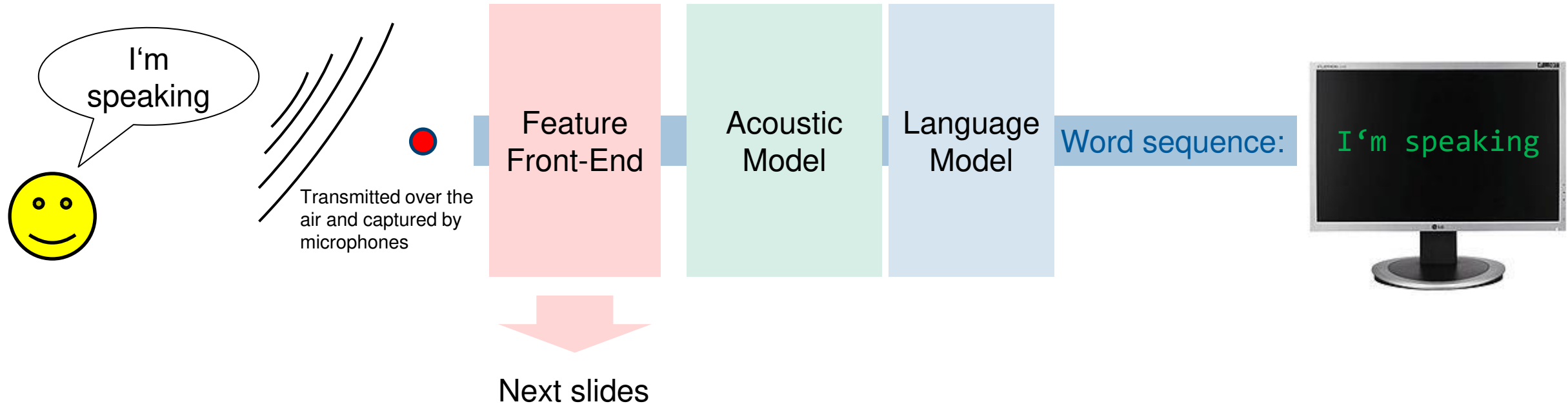


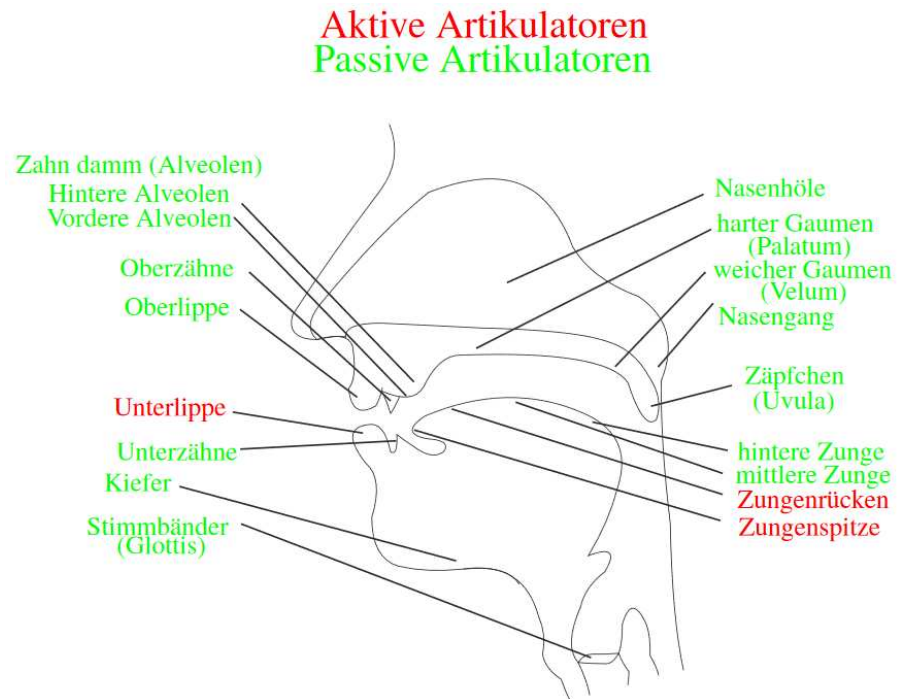
Automatic Speech Recognition, is the technology that allows human beings to use their voices to speak with a computer interface in a way that, in its most sophisticated variations, resembles normal human conversation.



Subfield of articulatory phonetics

Vowel tract & classification of sounds

The vocal tract can be well described as an all-pole filter, which can be useful, for example, for the analysis or synthesis of speech signals. The speech organs that play a special role in sound production or shaping are called articulators. A distinction is made between the more or less consciously influenced articulators and those that are only used, or between active and passive articulators. In order to describe the many, different sounds of the human language, one needs first a smallest unit, which can serve as basis for a description alphabet. In phonetics, this smallest unit is called a sound or a phon.

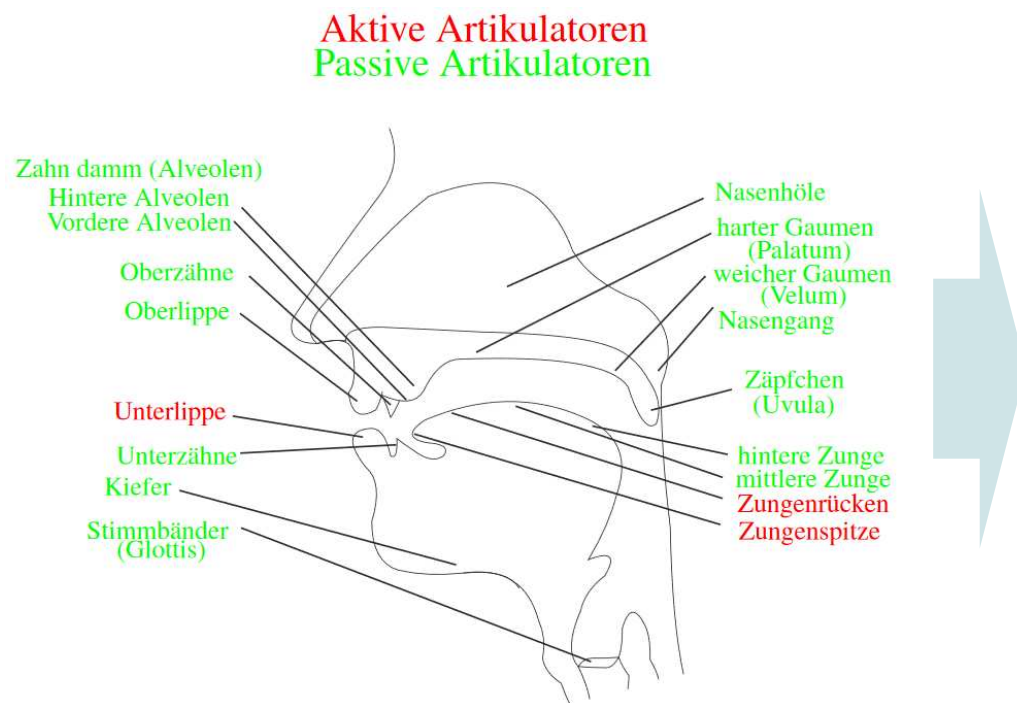


Subfield of articulatory phonetics

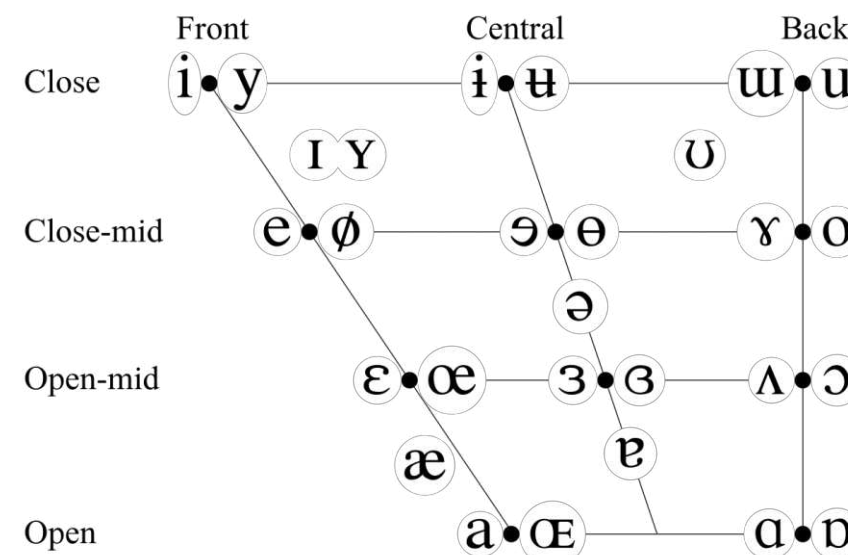
Vowel tract & classification of sounds



The vocal tract can be well described as an all-pole filter, which can be useful, for example, for the analysis or synthesis of speech signals. The speech organs that play a special role in sound production or shaping are called articulators. A distinction is made between the more or less consciously influenced articulators and those that are only used, or between active and passive articulators. In order to describe the many, different sounds of the human language, one needs first a smallest unit, which can serve as basis for a description alphabet. In phonetics, this smallest unit is called a sound or a phon.



VOWELS



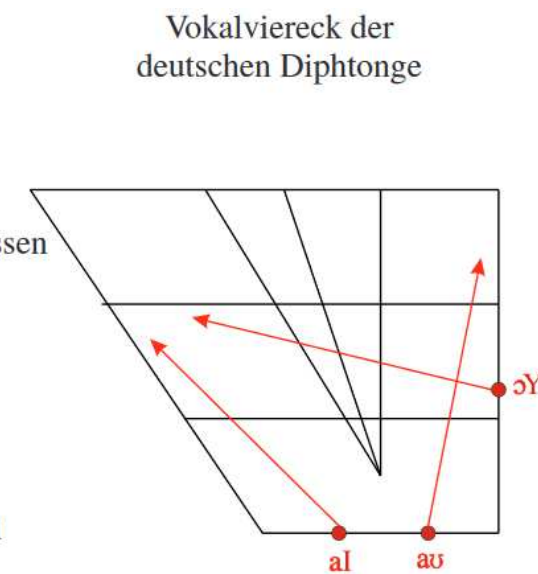
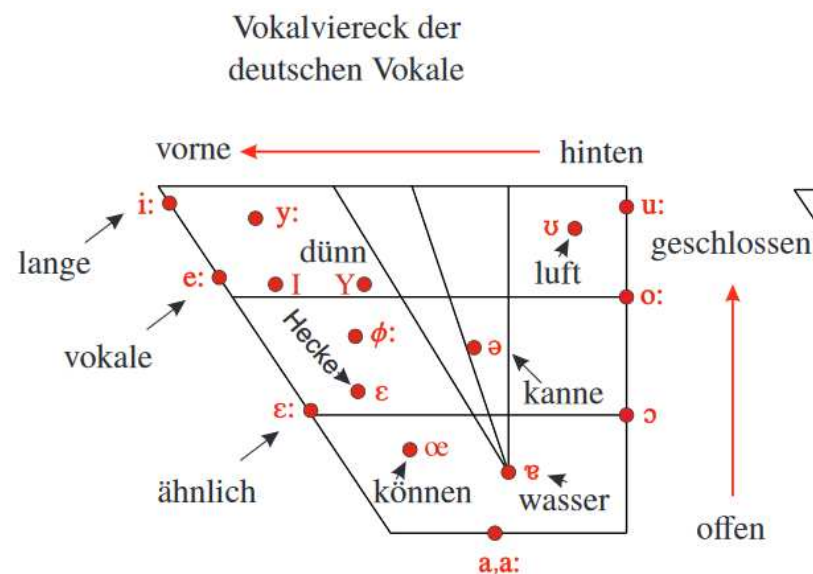
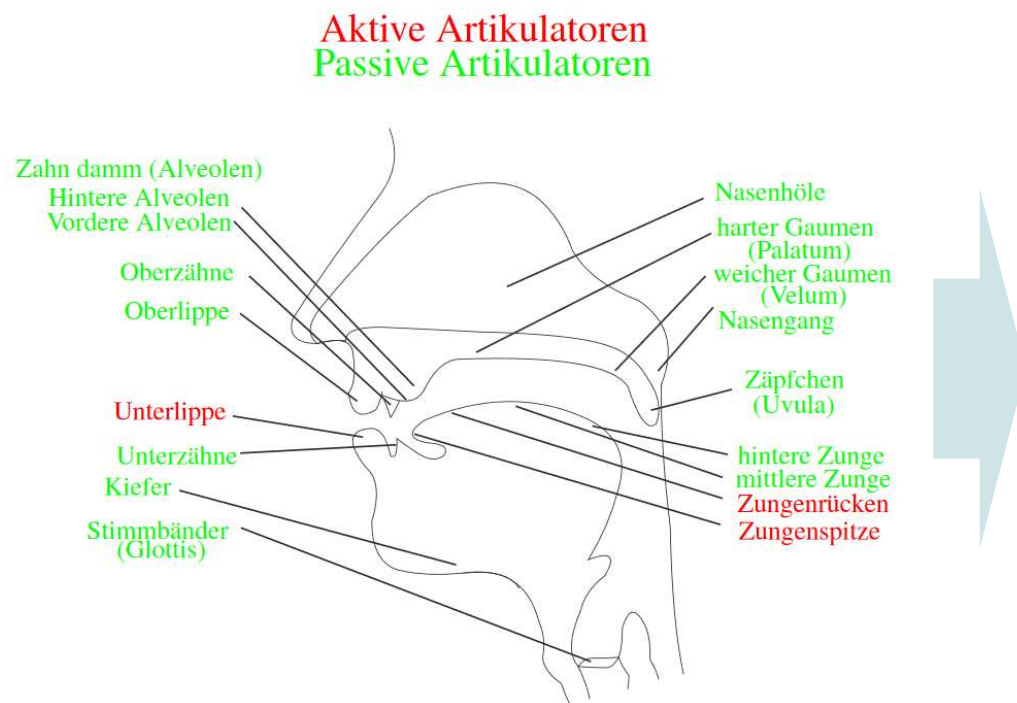
Where symbols appear in pairs, the one to the right represents a rounded vowel.

Subfield of articulatory phonetics

Vowel tract & classification of sounds



The vocal tract can be well described as an all-pole filter, which can be useful, for example, for the analysis or synthesis of speech signals. The speech organs that play a special role in sound production or shaping are called articulators. A distinction is made between the more or less consciously influenced articulators and those that are only used, or between active and passive articulators. In order to describe the many, different sounds of the human language, one needs first a smallest unit, which can serve as basis for a description alphabet. In phonetics, this smallest unit is called a sound or a phon.

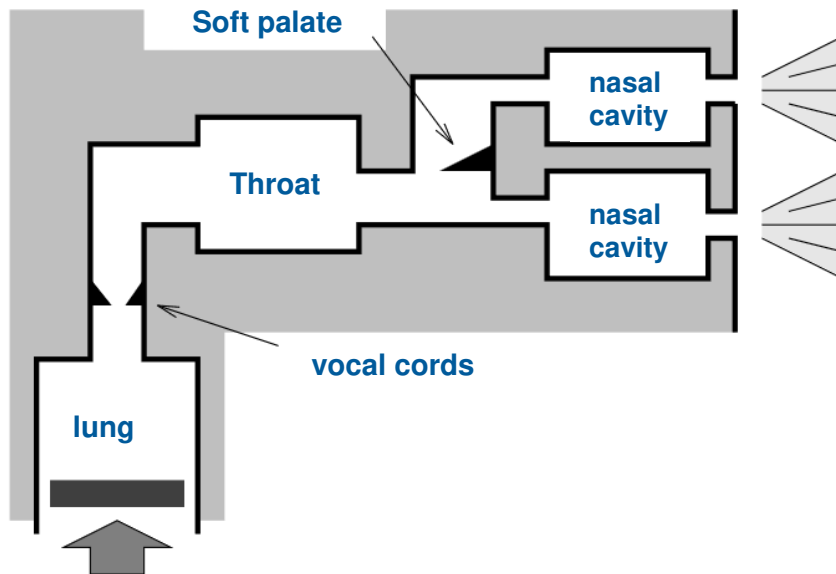


Subfield of articulatory phonetics

Physiologically motivated model of speech generation



To describe speech generation mathematically, the model in the lower left image is often used. Here, the lung serves as the source that provides the airflow for all further processes. The vocal cords determine whether the sound is to be voiced or unvoiced. In the case of unvoiced sounds, the vocal cords are so far apart that they are not influenced too much by the passing air stream; in the case of voiced sounds, they lie against each other and are moved apart at regular intervals by the air stream, thus causing them to vibrate. The frequency of this oscillation is also referred to as the fundamental frequency.

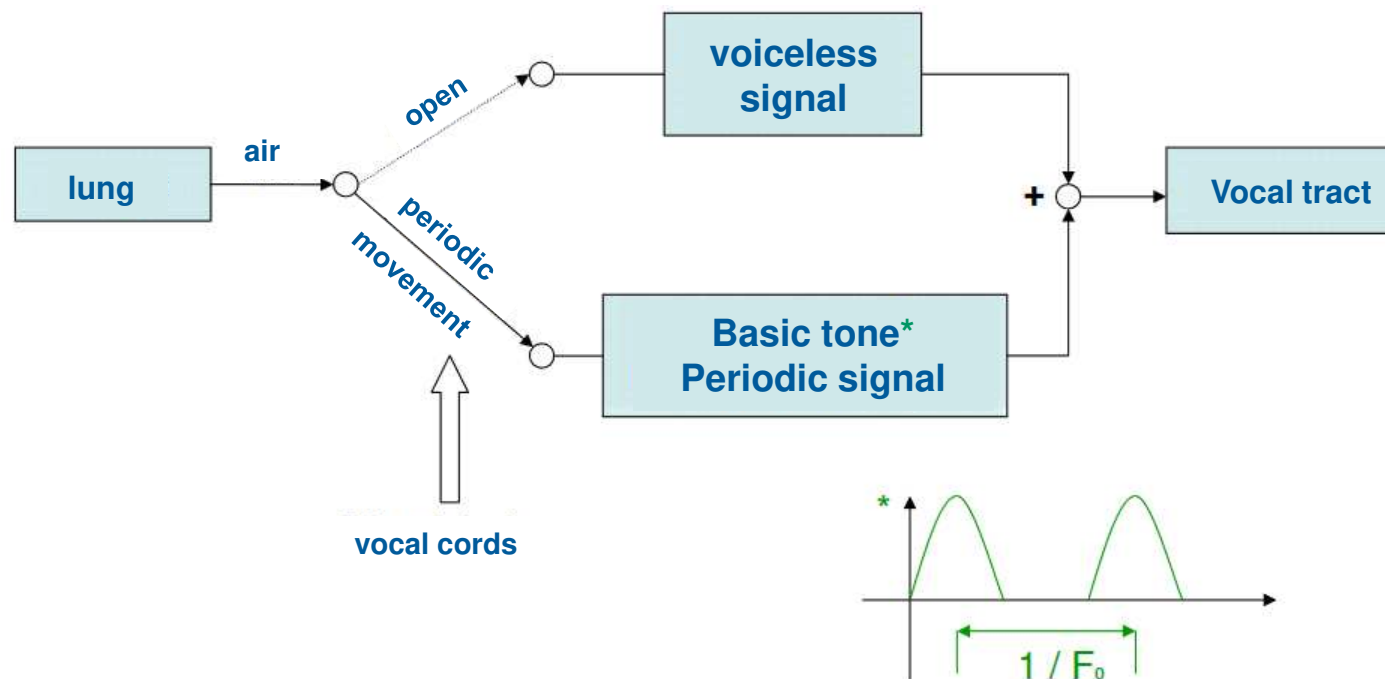
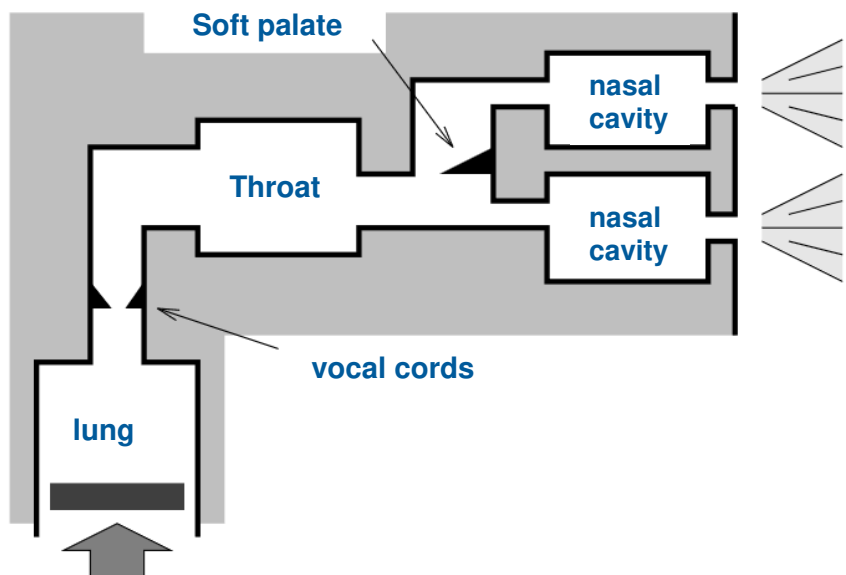


Subfield of articulatory phonetics

Physiologically motivated model of speech generation



To describe speech generation mathematically, the model in the lower left image is often used. Here, the lung serves as the source that provides the airflow for all further processes. The vocal cords determine whether the sound is to be voiced or unvoiced. In the case of unvoiced sounds, the vocal cords are so far apart that they are not influenced too much by the passing air stream; in the case of voiced sounds, they lie against each other and are moved apart at regular intervals by the air stream, thus causing them to vibrate. The frequency of this oscillation is also referred to as the fundamental frequency.

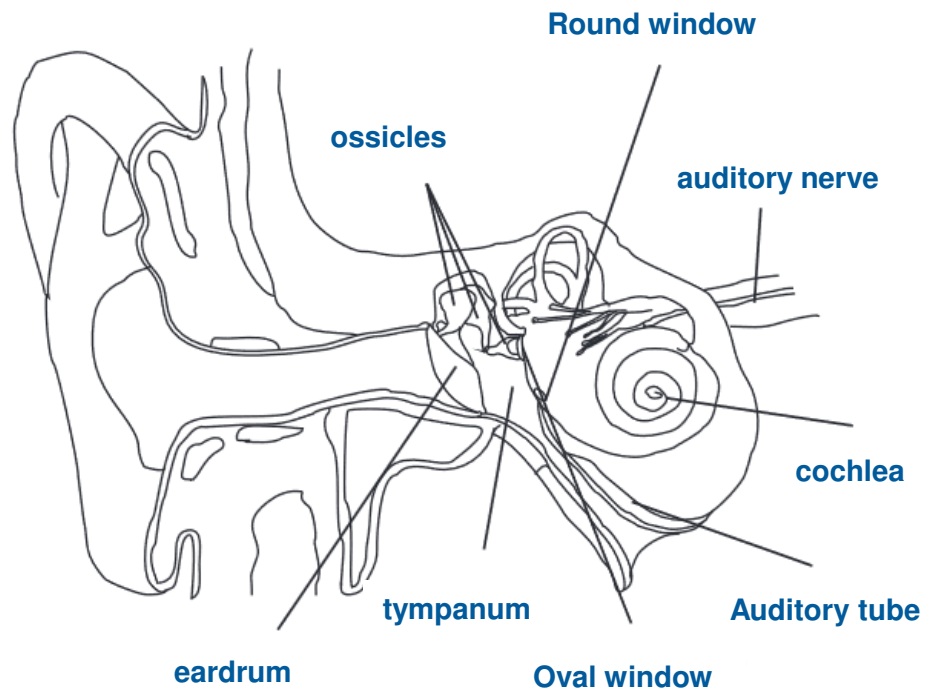


Auditory apparatus

Structure of the auditory system



The outer ear, for its part, consists of the auricle, whose directional characteristics, among other things, make it easier to focus on sounds from a particular direction of incidence, the auditory canal, which primarily keeps foreign bodies out, and is bounded by the eardrum, which is stimulated to vibrate by sound waves. In the middle ear, the three ossicles, malleus, incus, and stapes, effect impedance matching, which is necessary because the sound resistance of the fluid-filled inner ear is much greater than that of air, so that without appropriate mechanical transduction, sound would have no appreciable effect on the inner ear

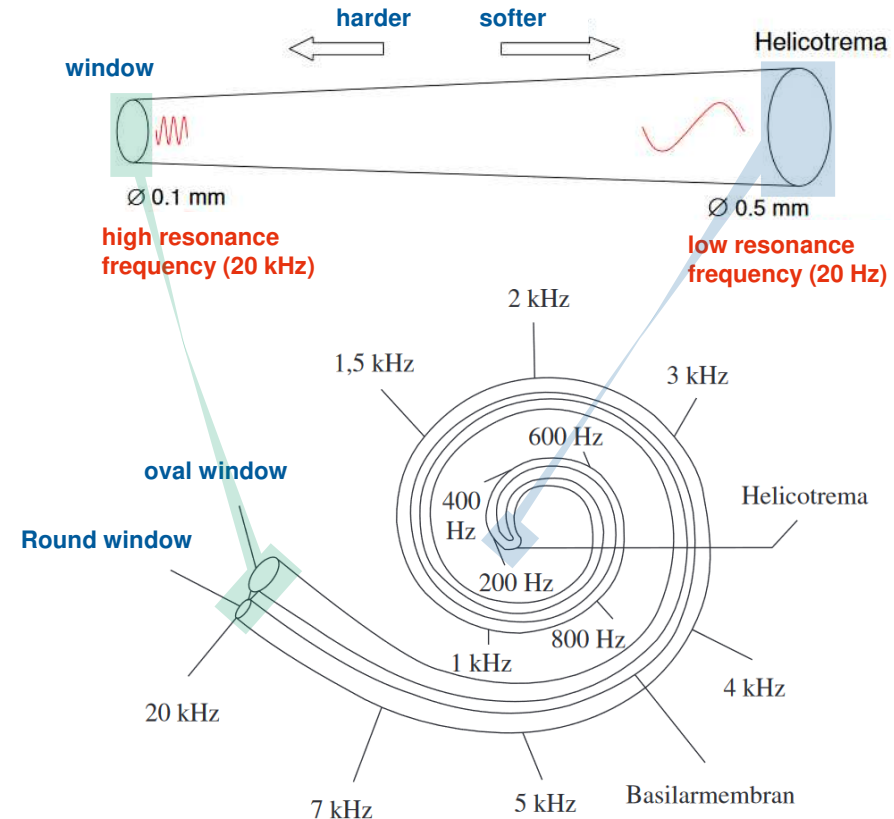
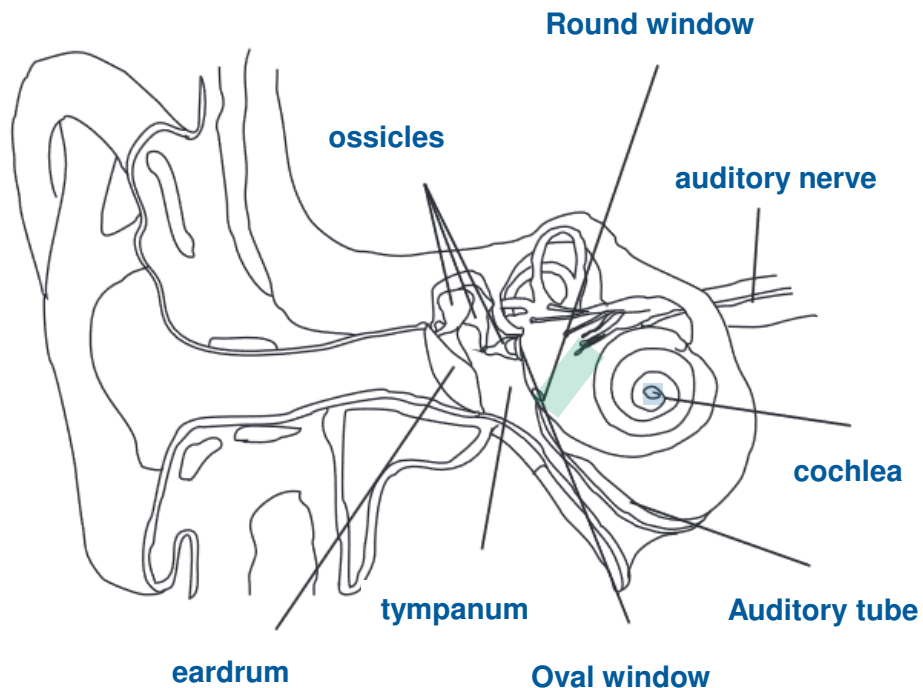


Auditory apparatus

Structure of the auditory system



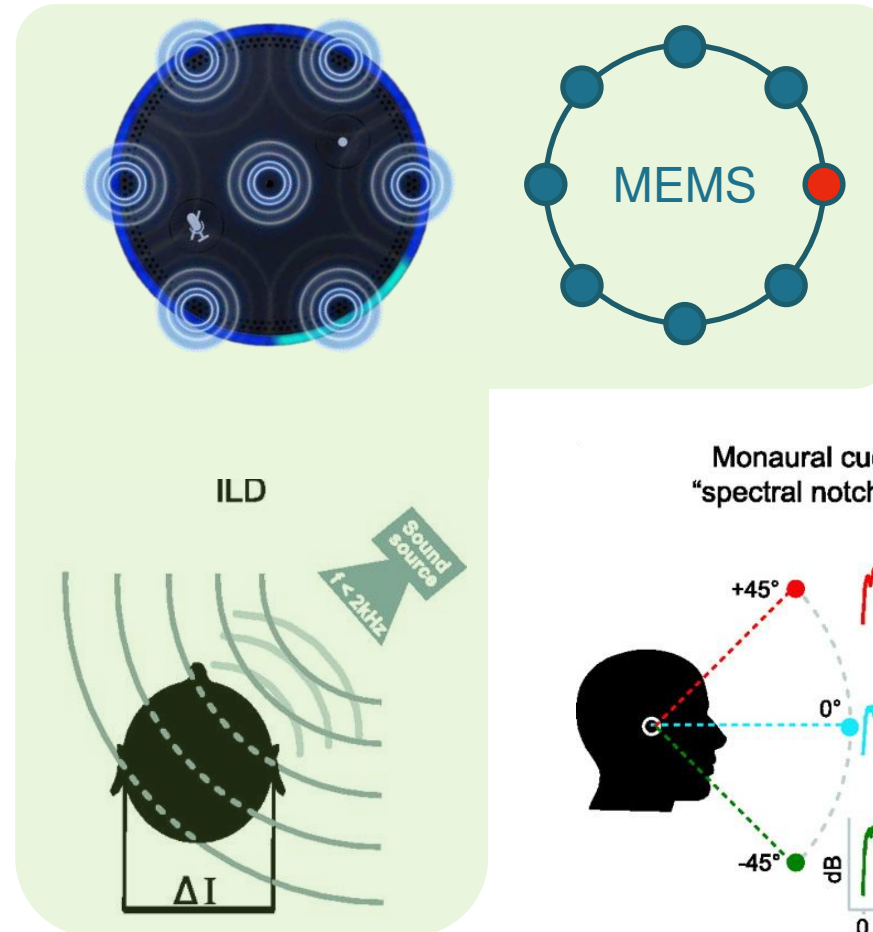
The outer ear, for its part, consists of the auricle, whose directional characteristics, among other things, make it easier to focus on sounds from a particular direction of incidence, the auditory canal, which primarily keeps foreign bodies out, and is bounded by the eardrum, which is stimulated to vibrate by sound waves. In the middle ear, the three ossicles, malleus, incus, and stapes, effect impedance matching, which is necessary because the sound resistance of the fluid-filled inner ear is much greater than that of air, so that without appropriate mechanical transduction, sound would have no appreciable effect on the inner ear



Feature Extraction for Speech Applications

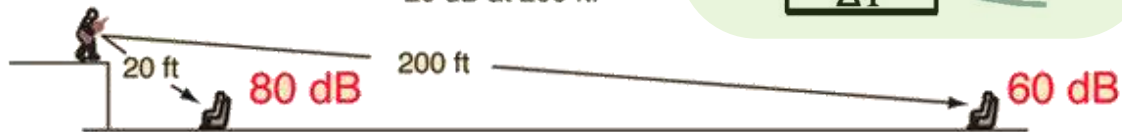


N-Channel Audio Capturing



The required point source acoustic power is only 0.05 watts.

By the inverse square law, the sound intensity will drop by 20 dB at 200 ft.



10 x distance implies $\frac{1}{100}$ x intensity or a 20 dB drop

Feature Extraction for Speech Applications



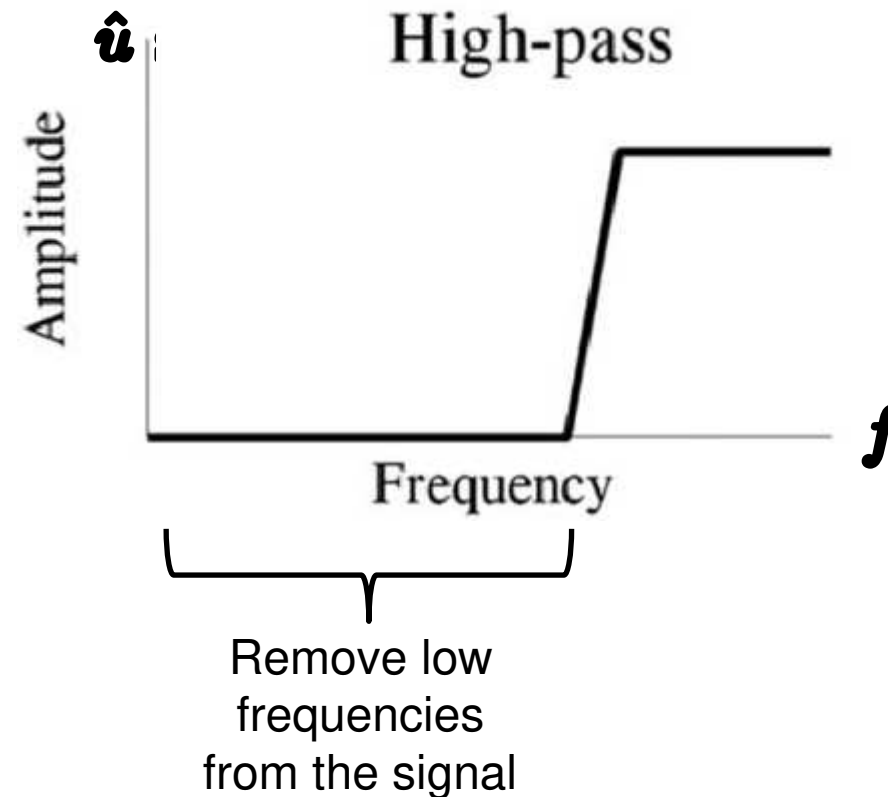
N-Channel Audio Capturing

High pass filter

$$u(t) = \hat{u} \sin(\omega t) = \hat{u} \sin(2\pi f t) = \hat{u} \sin \frac{2\pi t}{T} \quad \text{with } f = \frac{1}{T}$$

Example:

$$f = 50 \text{ Hz}$$



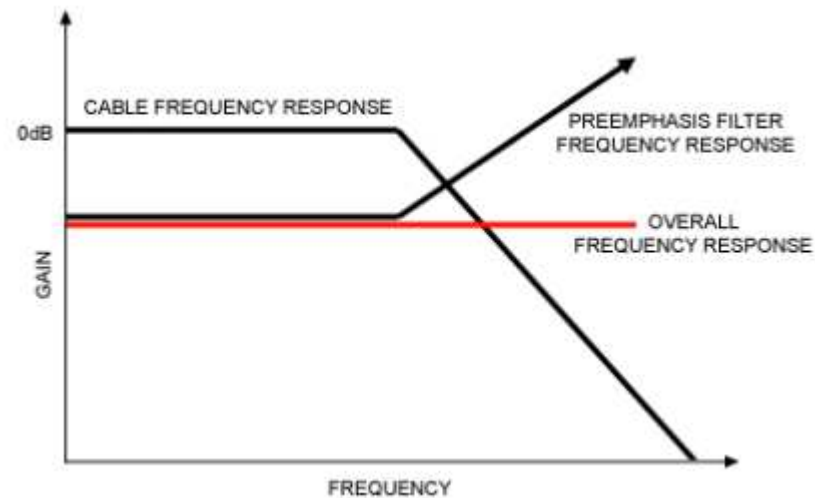
Feature Extraction for Speech Applications



N-Channel Audio Capturing

High pass filter

Pre-emphasis



$$\tau = \frac{1}{2\pi \cdot f_c}$$

Feature Extraction for Speech Applications

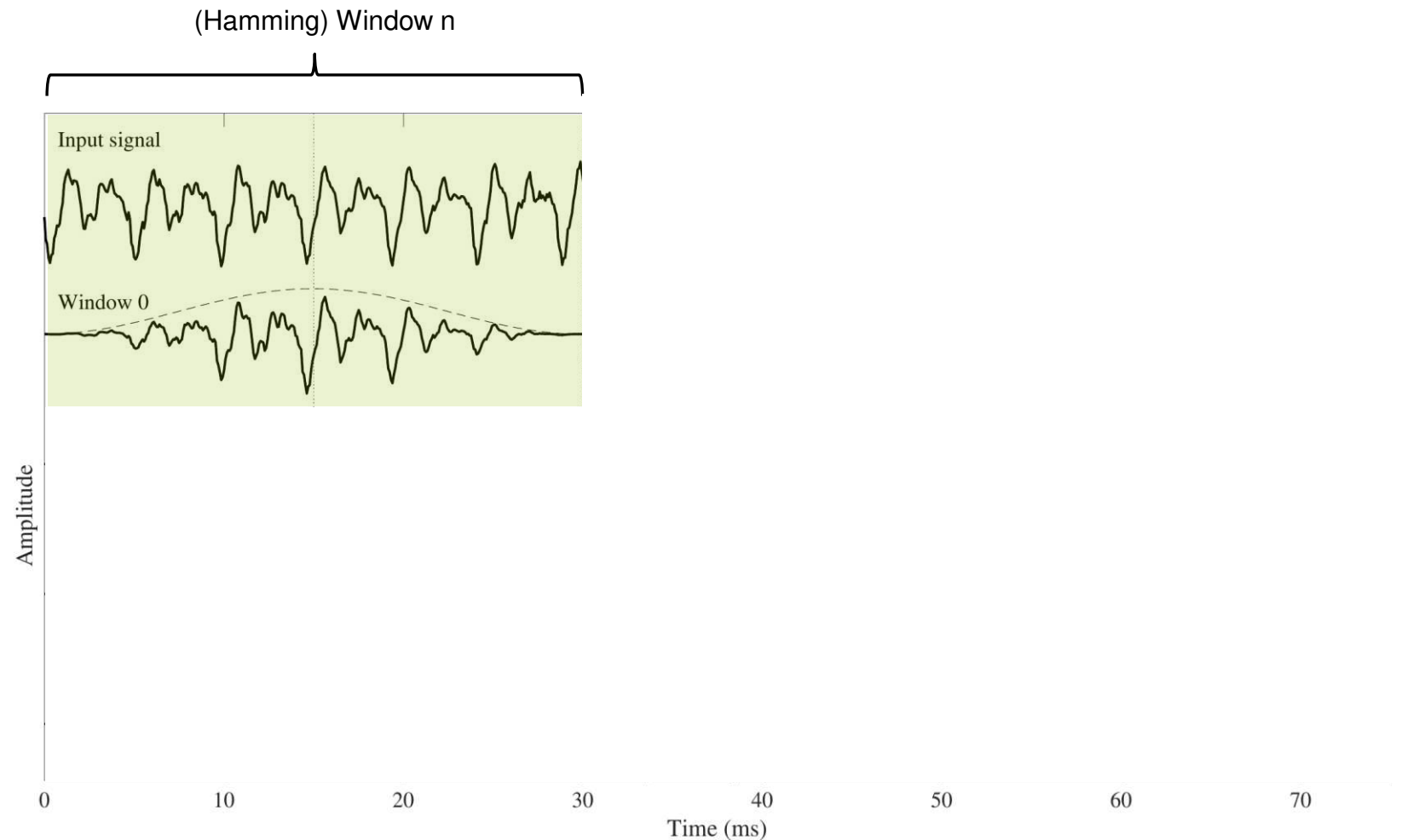


N-Channel Audio Capturing

High pass filter

Pre-emphasis

Windowing



Compute Feature Vector

Feature Extraction for Speech Applications

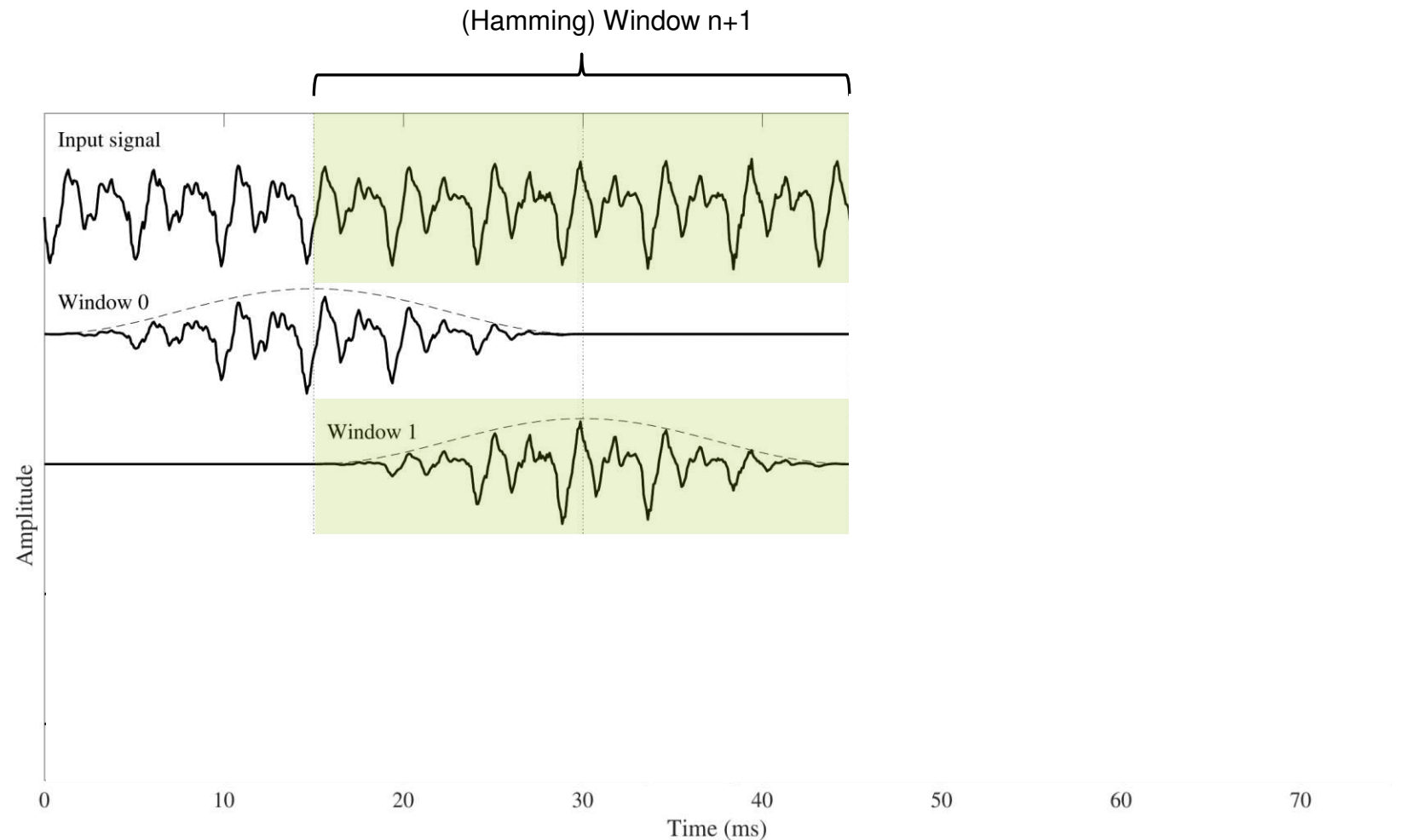


N-Channel Audio Capturing

High pass filter

Pre-emphasis

Windowing



Compute Feature Vector

Feature Extraction for Speech Applications

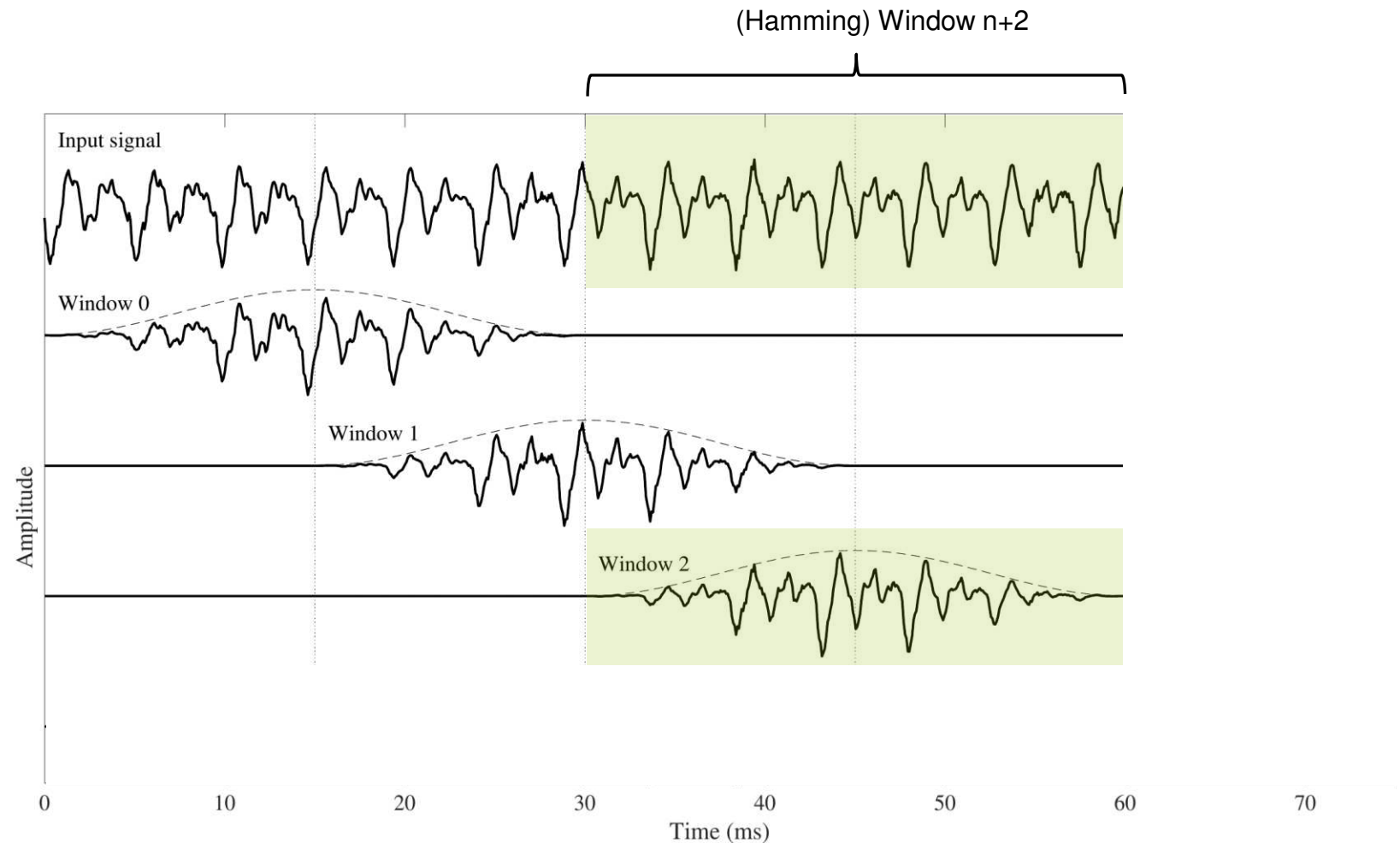


N-Channel Audio Capturing

High pass filter

Pre-emphasis

Windowing



Compute Feature Vector

Feature Extraction for Speech Applications

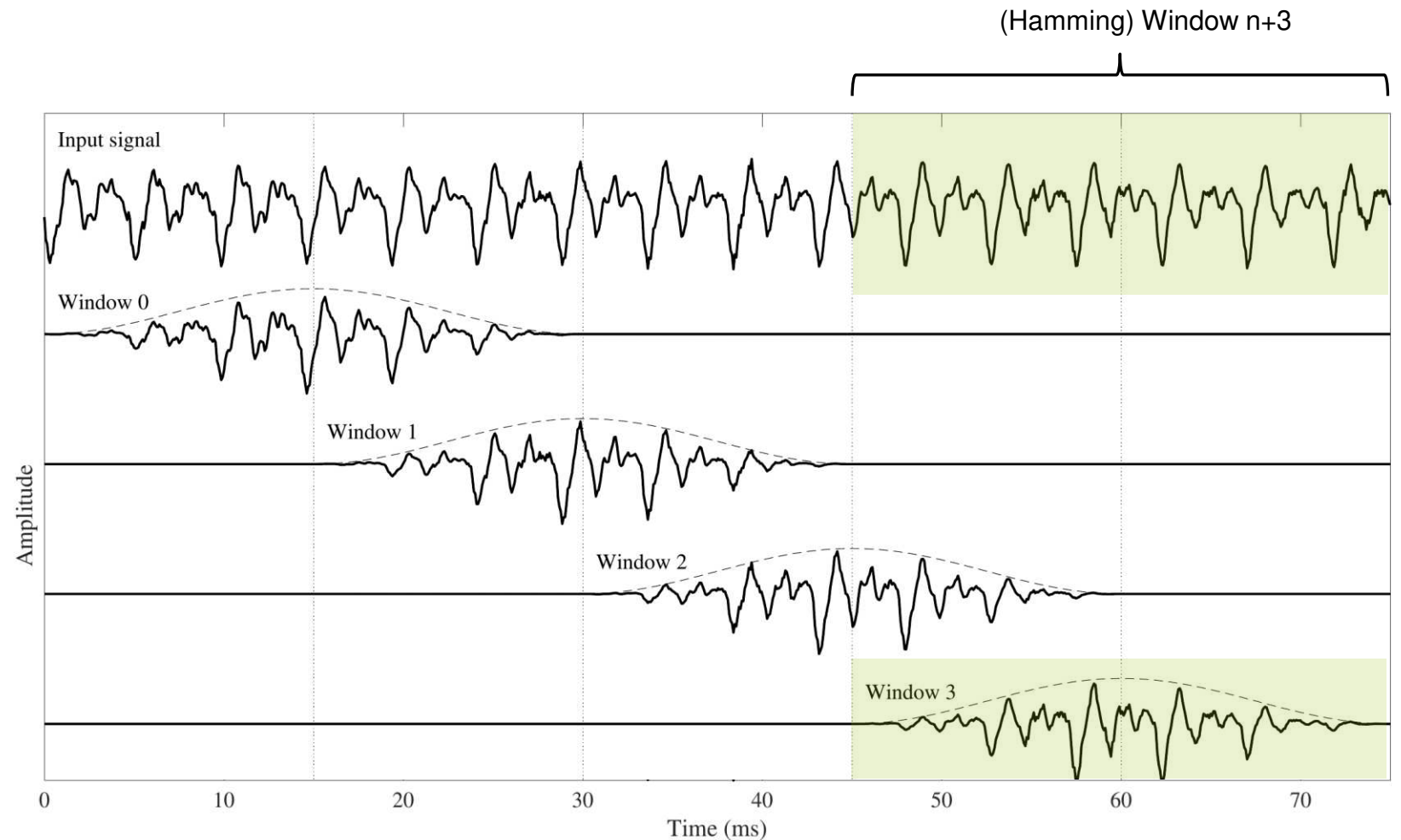


N-Channel Audio Capturing

High pass filter

Pre-emphasis

Windowing



Compute Feature Vector

Feature Extraction for Speech Applications



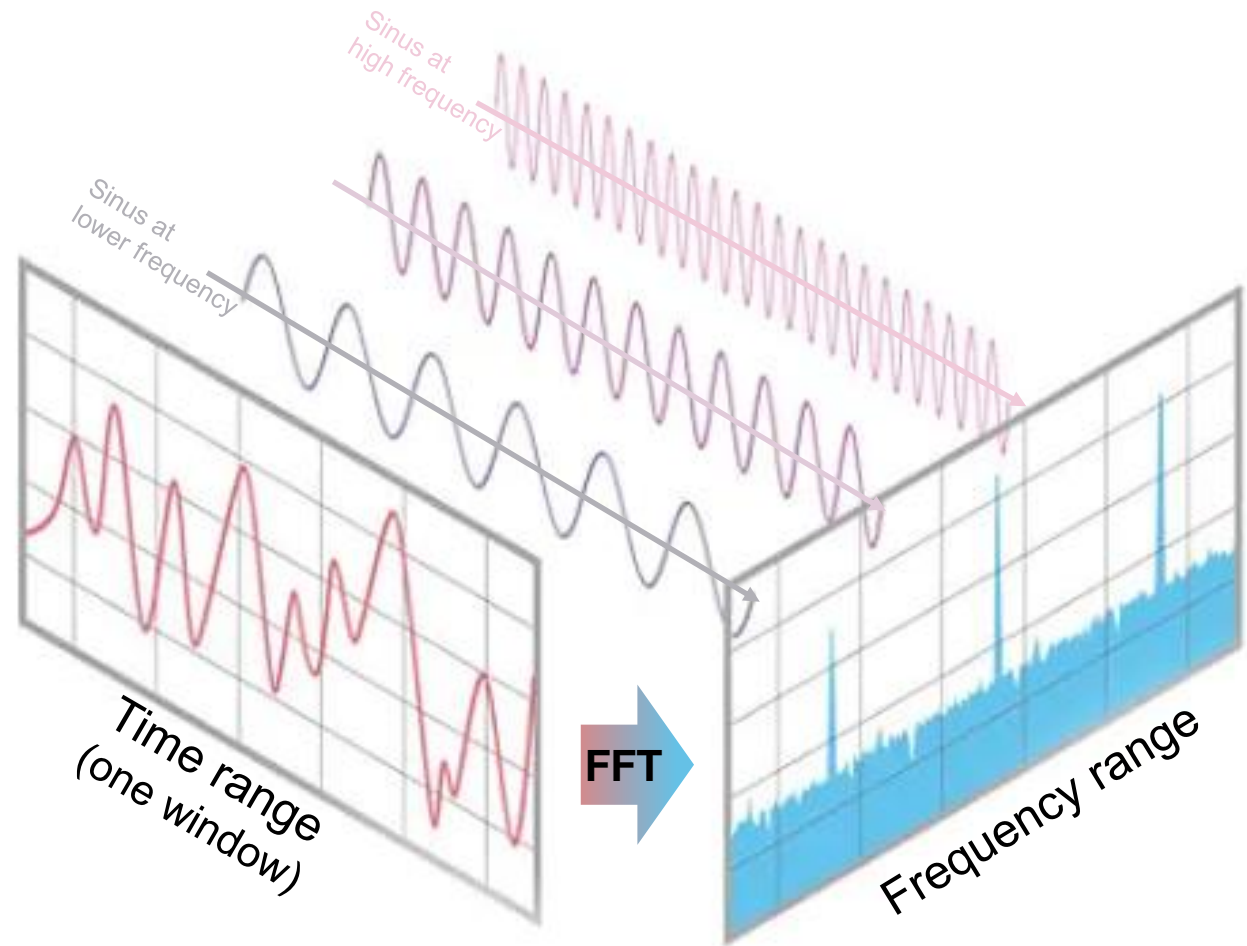
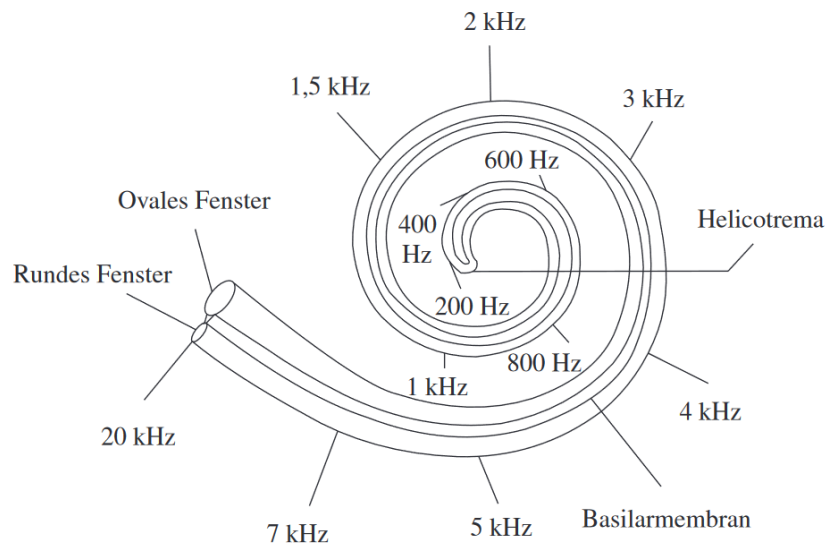
N-Channel Audio Capturing

High pass filter

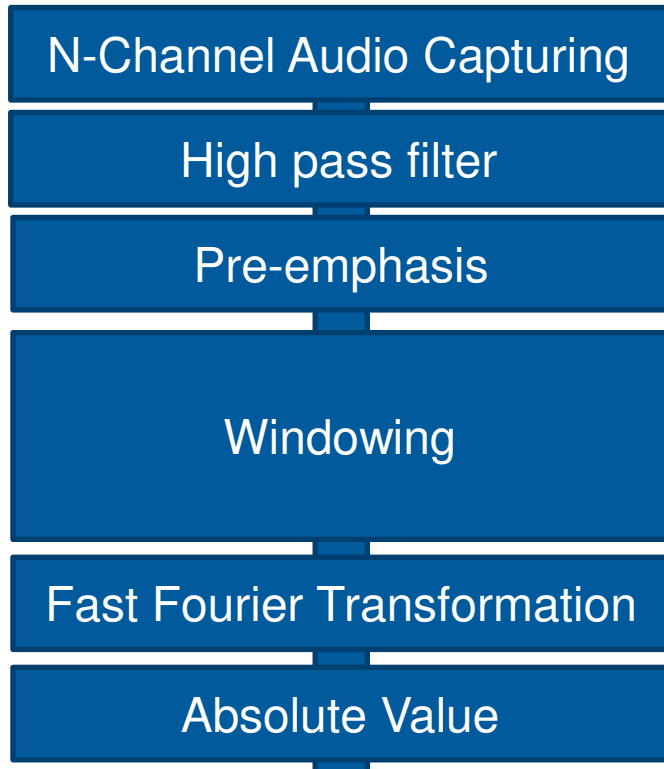
Pre-emphasis

Windowing

Fast Fourier Transformation



Feature Extraction for Speech Applications



$$|x| = \begin{cases} x & \text{if } x \geq 0 \\ -x & \text{if } x < 0 \end{cases}$$

Feature Extraction for Speech Applications



N-Channel Audio Capturing

High pass filter

Pre-emphasis

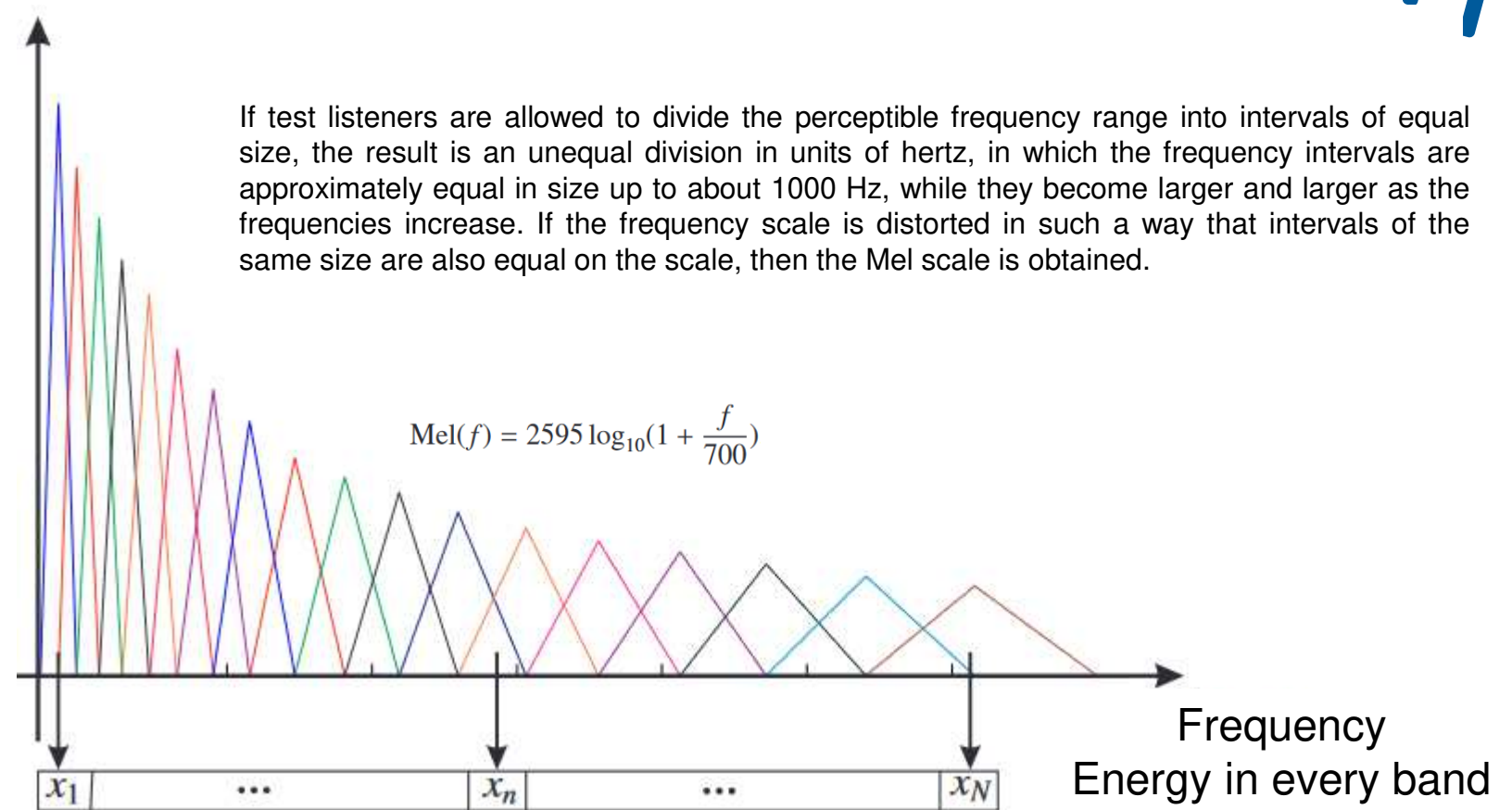
Windowing

Fast Fourier Transformation

Absolute Value

Mel-Scale Filter bank

If test listeners are allowed to divide the perceptible frequency range into intervals of equal size, the result is an unequal division in units of hertz, in which the frequency intervals are approximately equal in size up to about 1000 Hz, while they become larger and larger as the frequencies increase. If the frequency scale is distorted in such a way that intervals of the same size are also equal on the scale, then the Mel scale is obtained.



Stevens, Volkmann, and Newman in 1937

Feature Extraction for Speech Applications



N-Channel Audio Capturing

High pass filter

Pre-emphasis

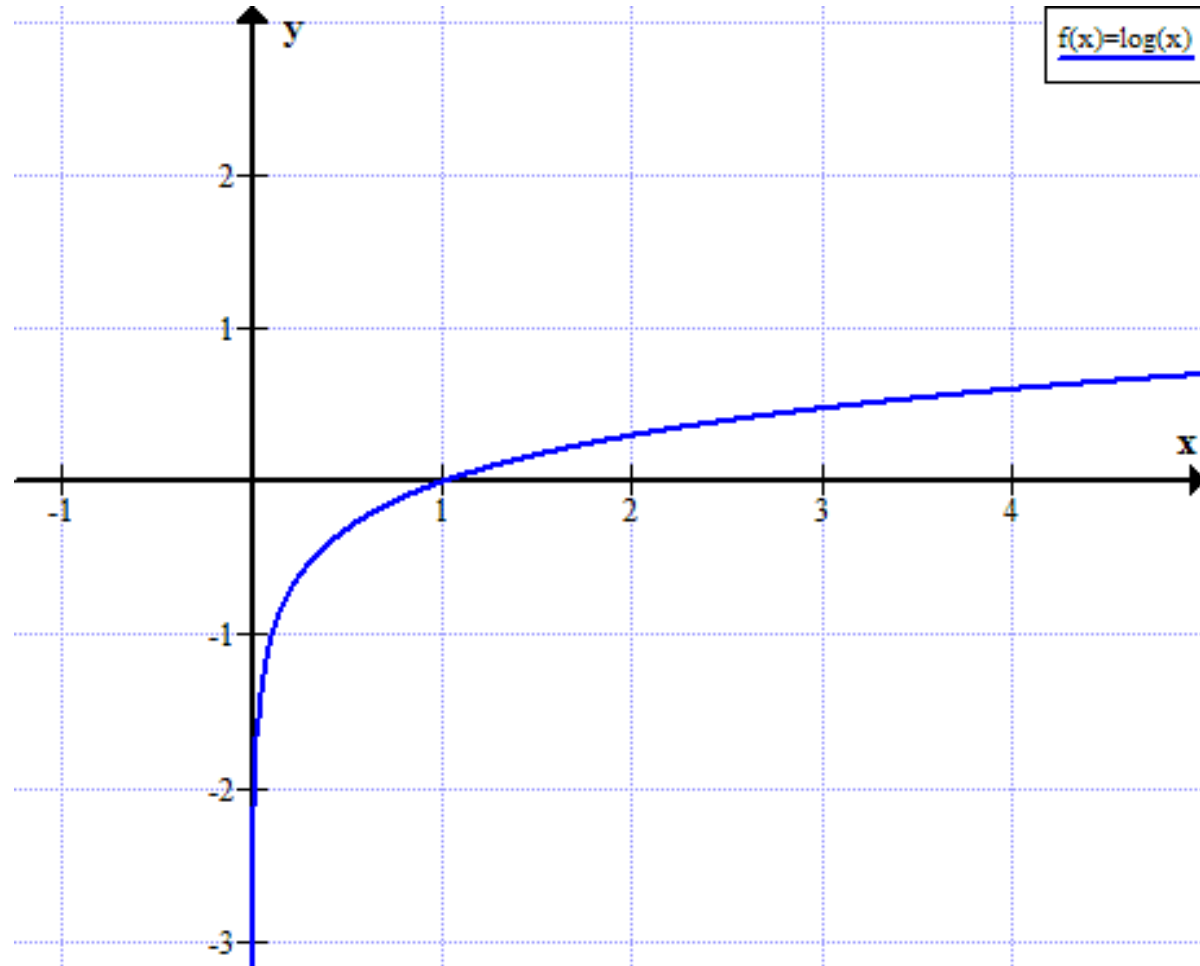
Windowing

Fast Fourier Transformation

Absolute Value

Mel-Scale Filter bank

Log-scale



Feature Extraction for Speech Applications



N-Channel Audio Capturing

High pass filter

Pre-emphasis

Windowing

Fast Fourier Transformation

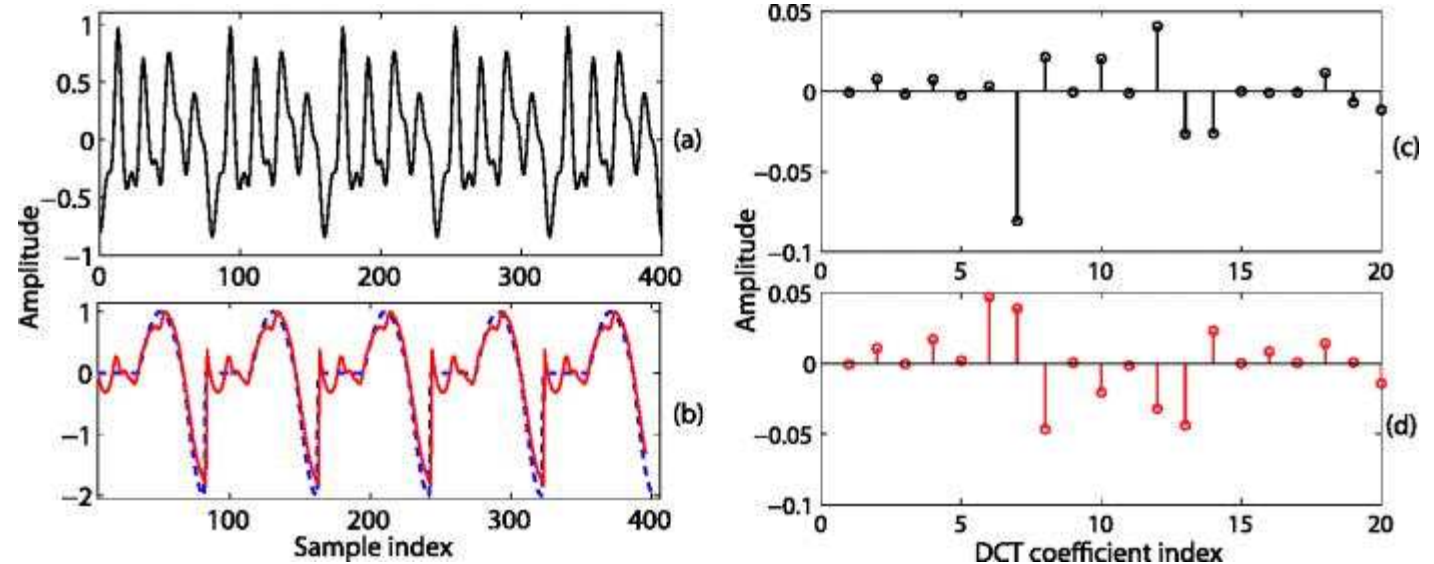
Absolute Value

Mel-Scale Filter bank

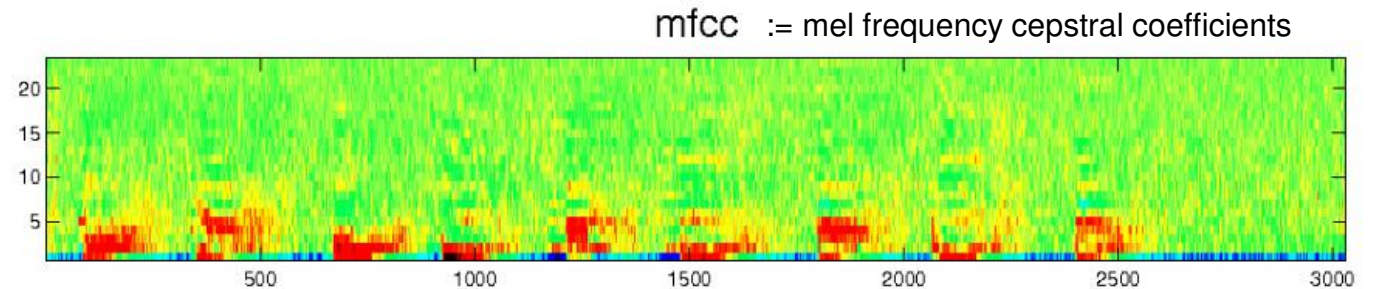
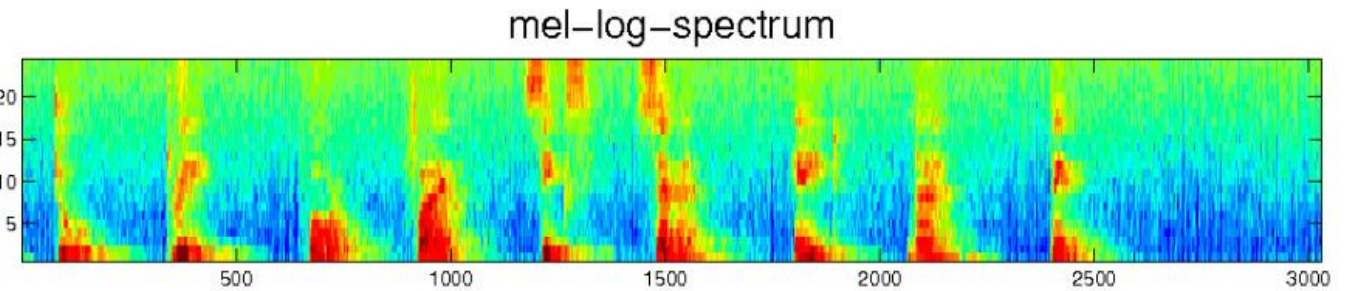
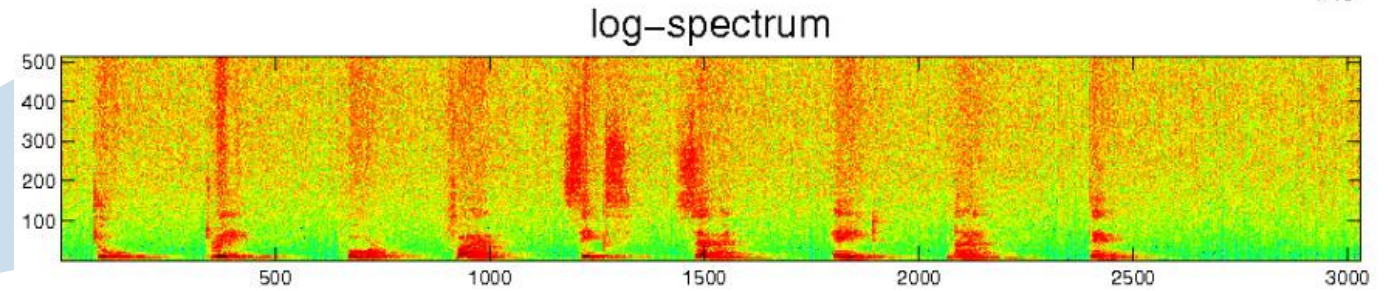
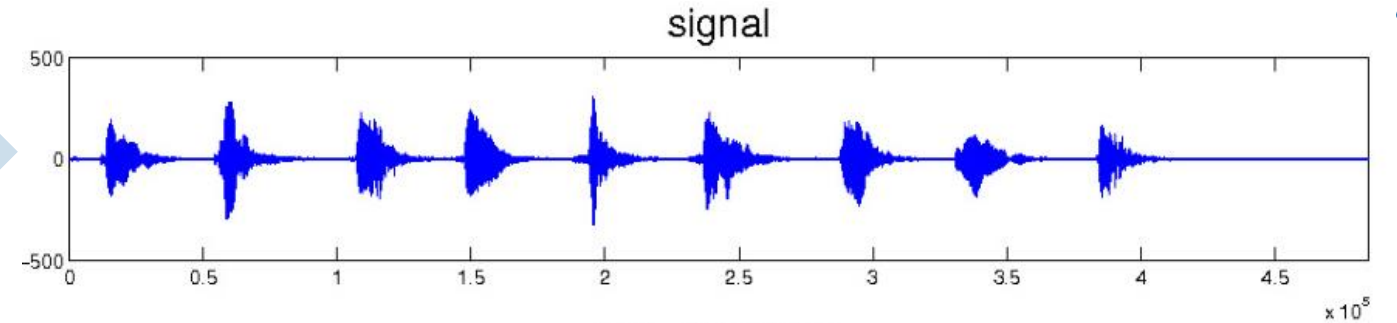
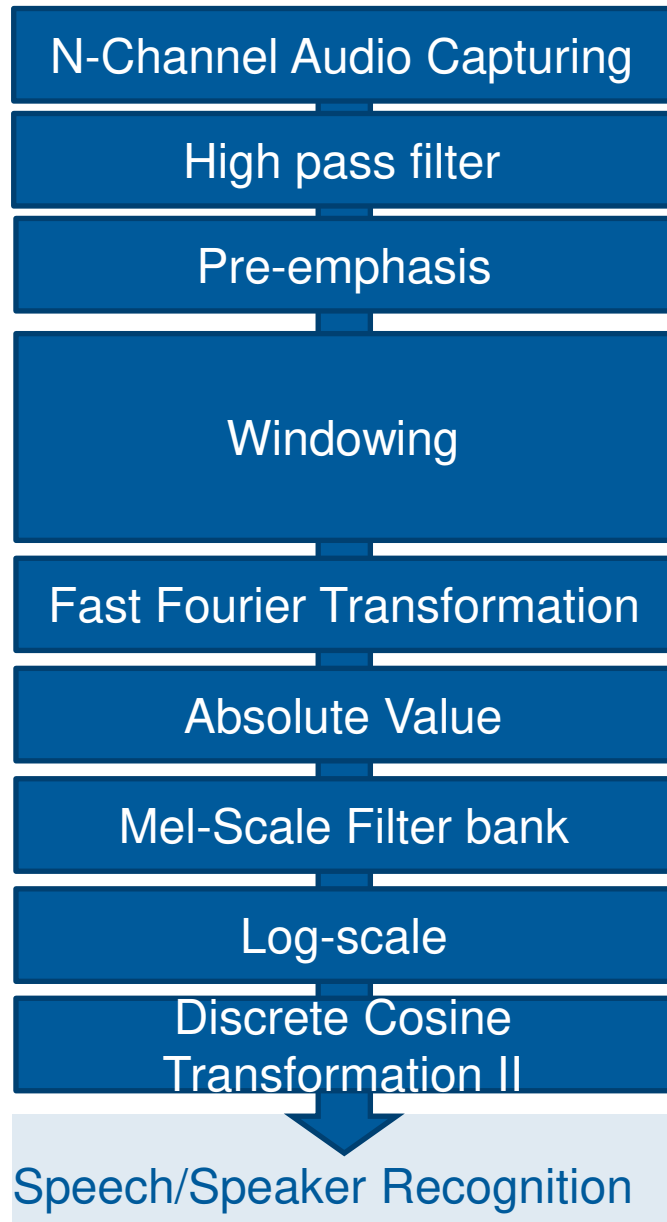
Log-scale

Discrete Cosine Transformation II

Decorrelation is a general term for any process that is used to reduce autocorrelation within a signal, or cross-correlation within a set of signals, while preserving other aspects of the signal.



Feature Extraction for Speech Applications

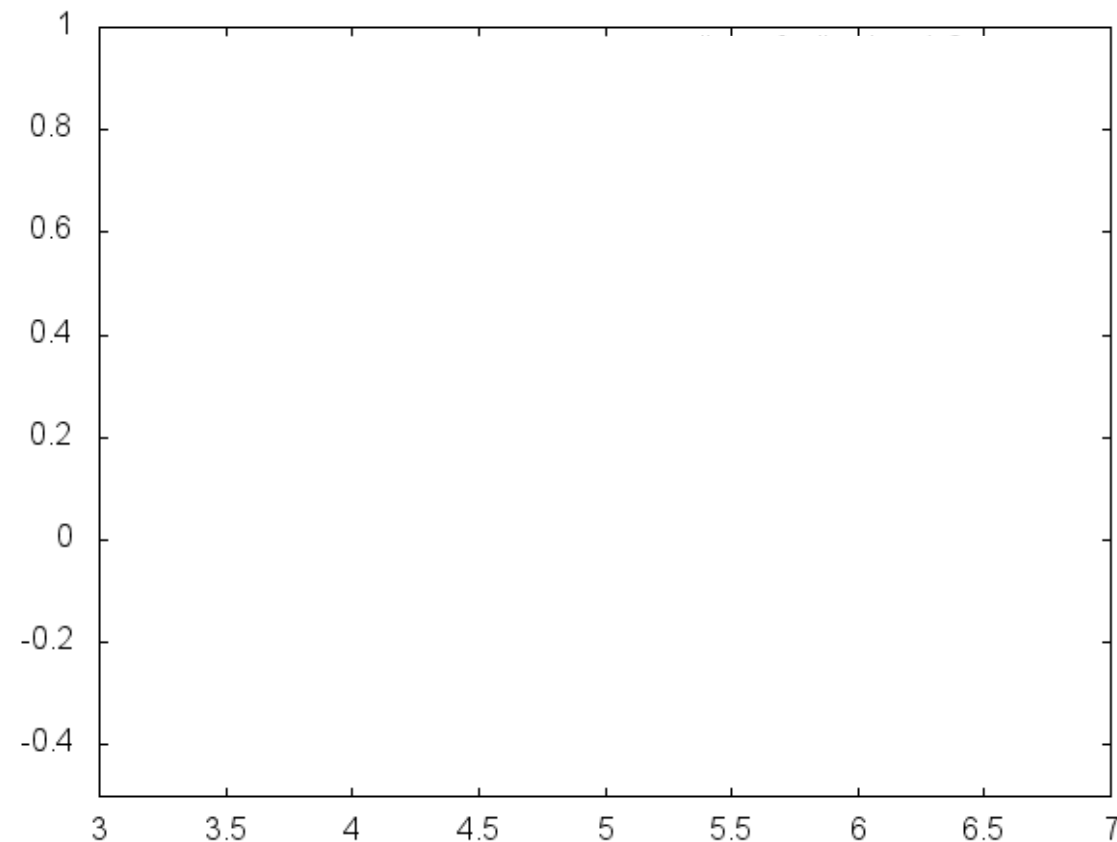
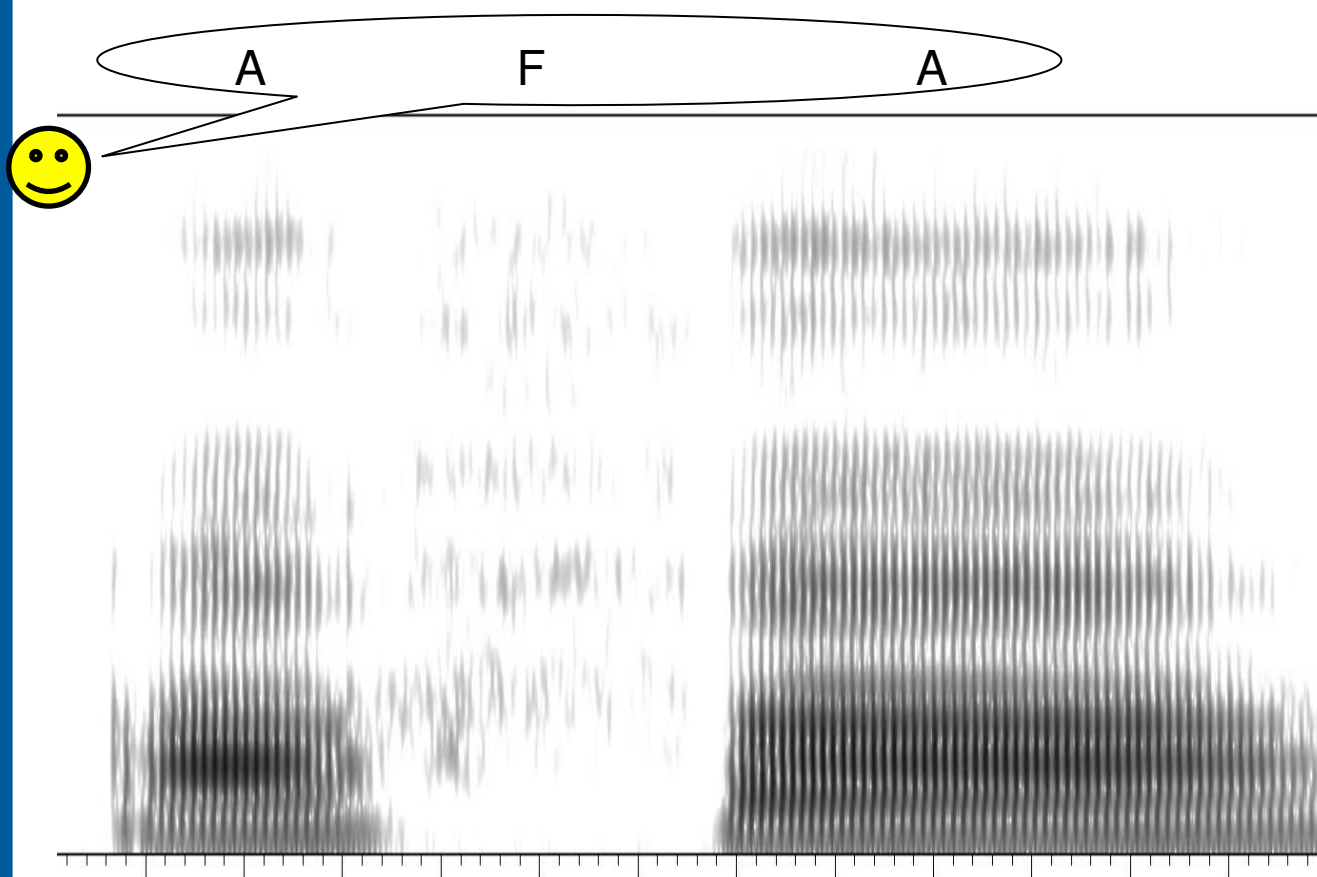


Classification with Speech Features

Phone Classification



MFCC features are used for Automatic Speech Recognition

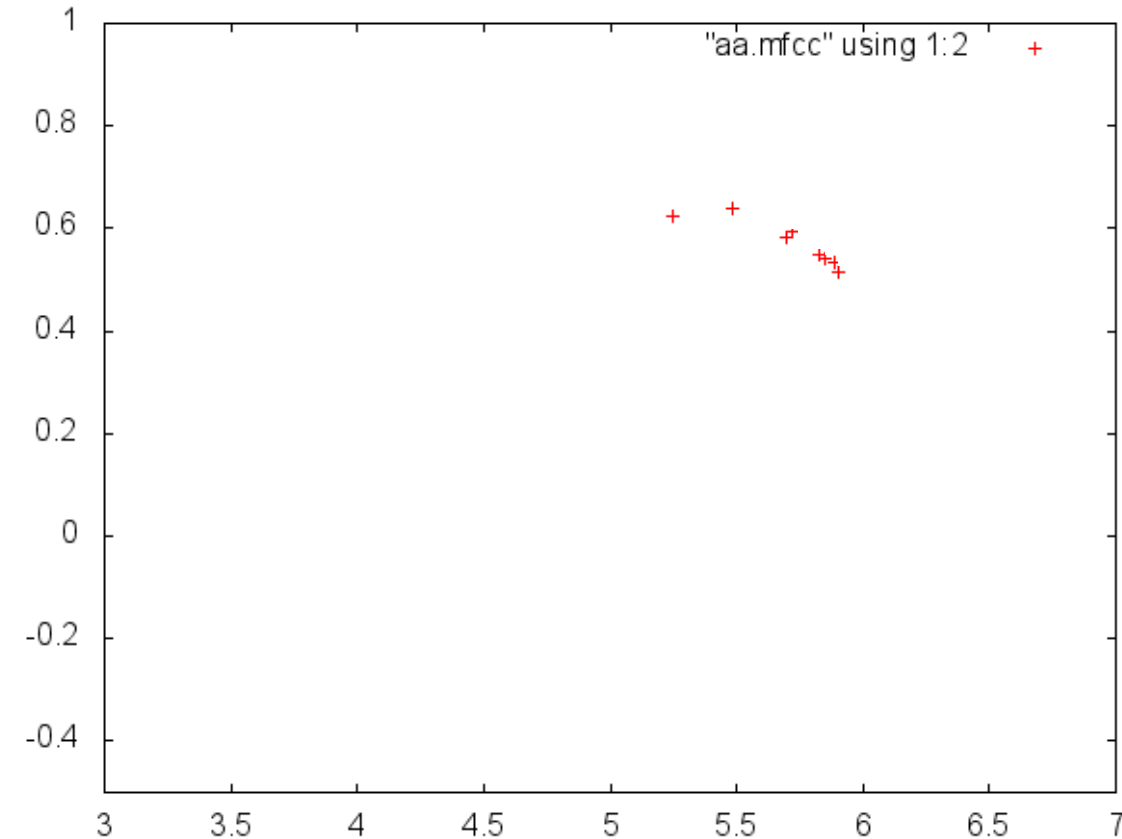
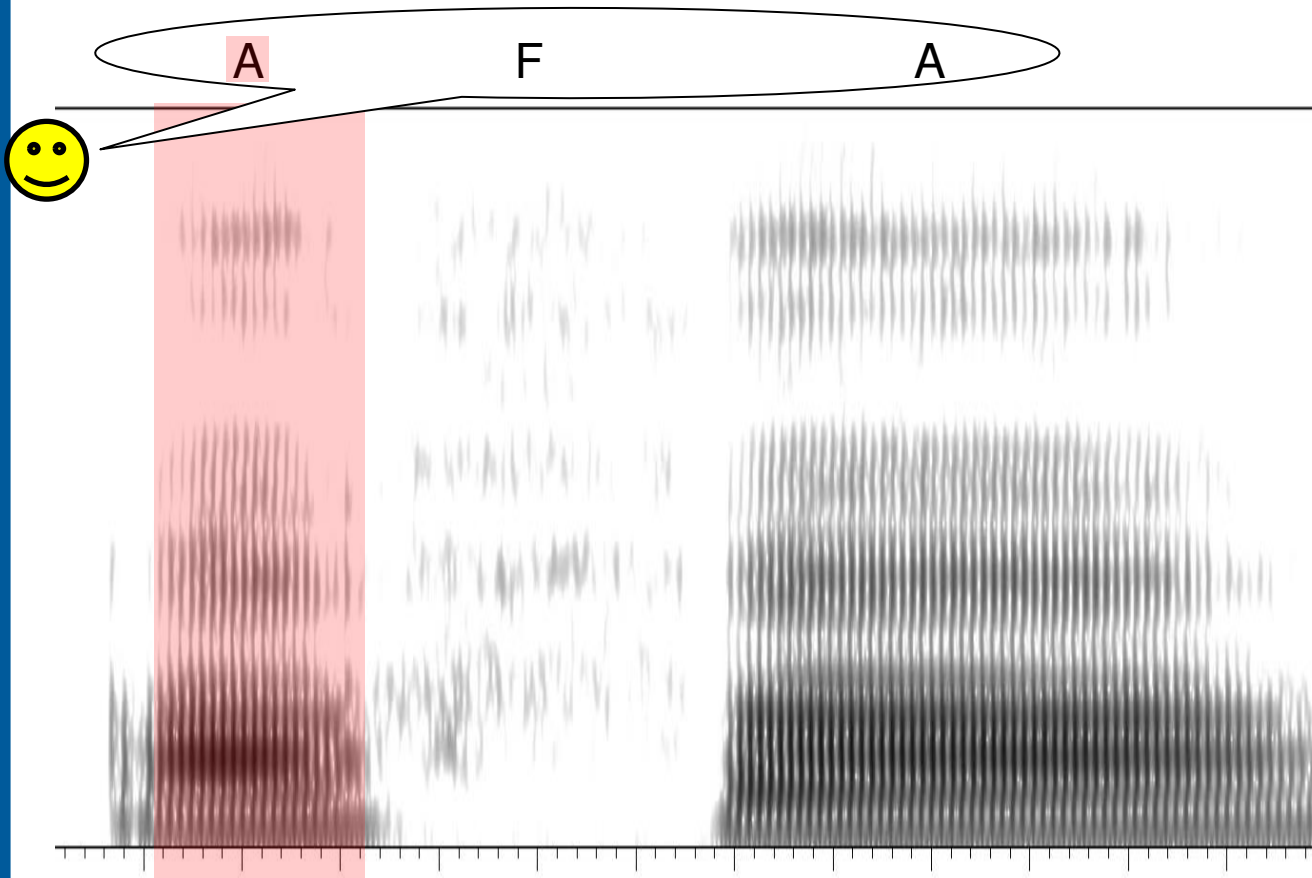


Classification with Speech Features

Phone Classification



MFCC features are use for Automatic Speech Recognition

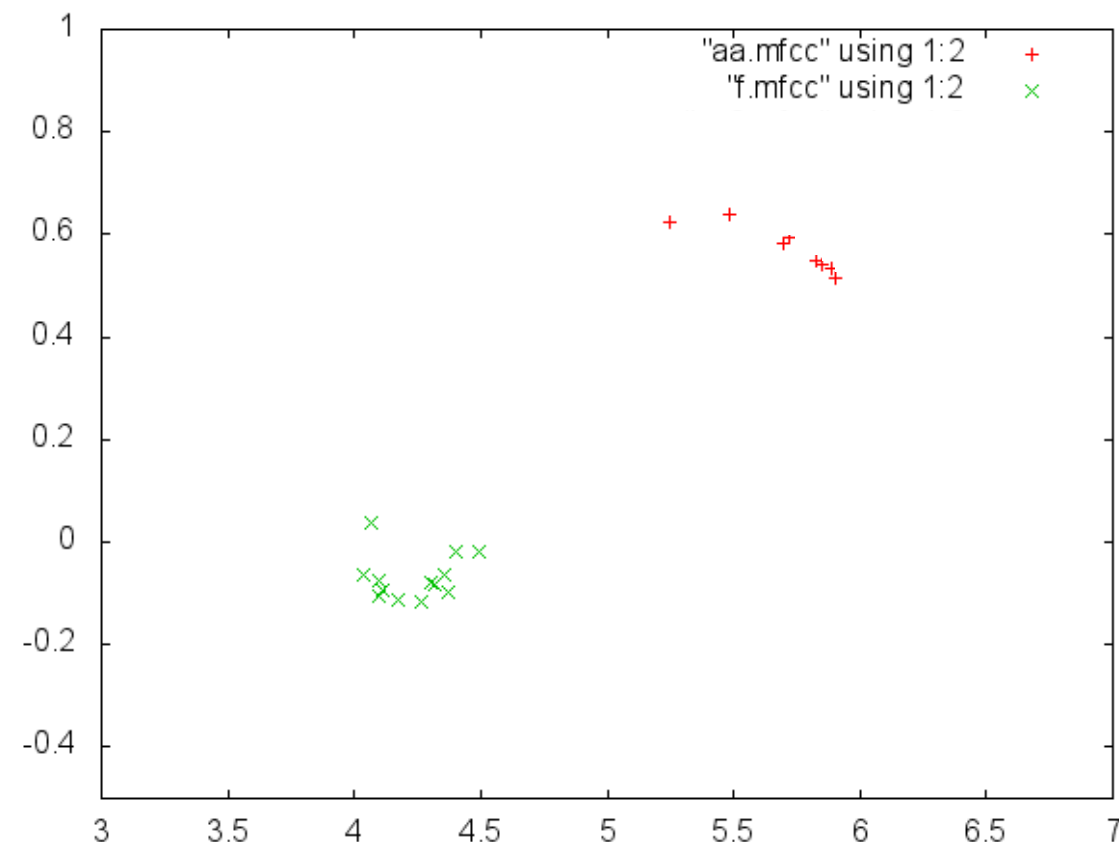
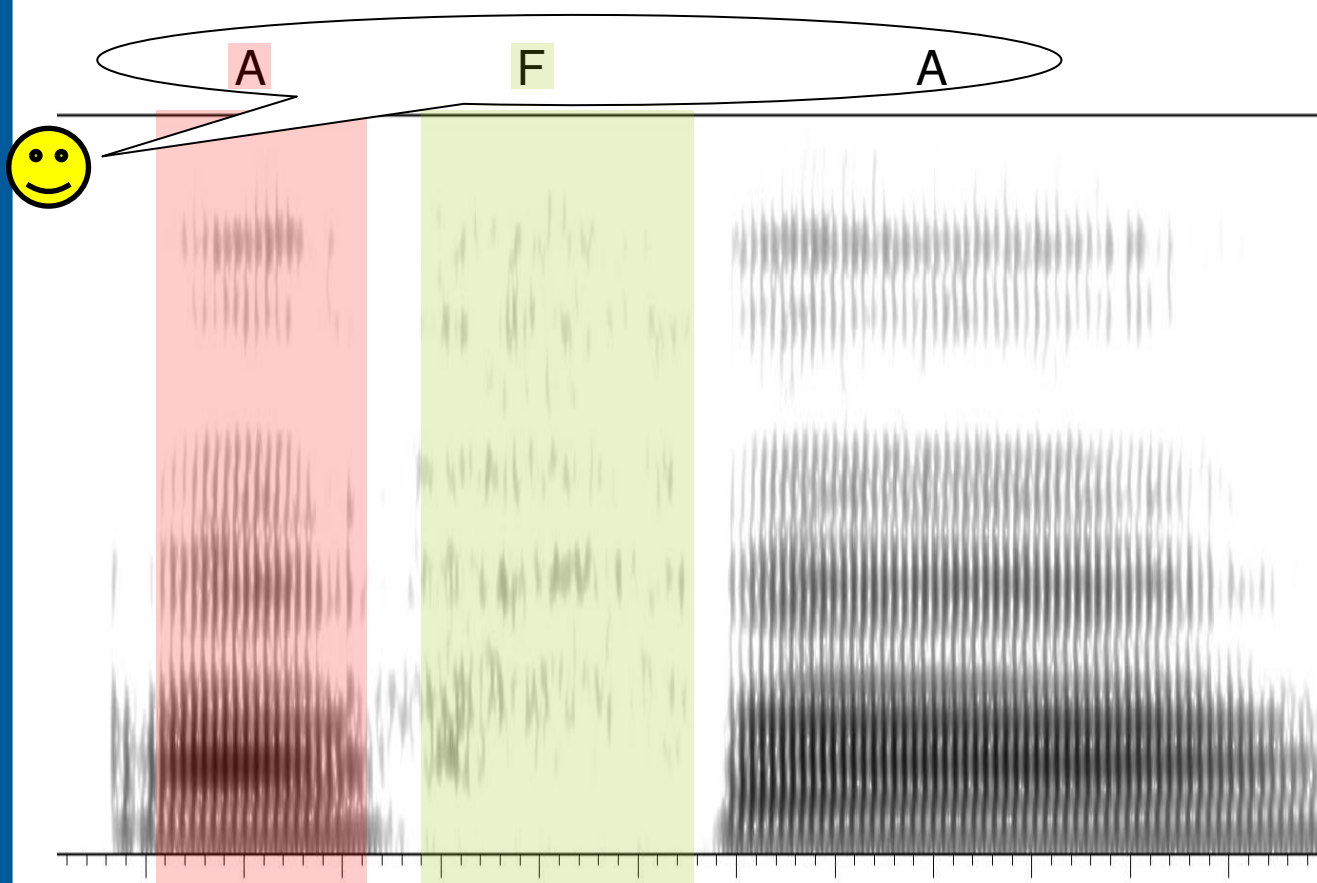


Classification with Speech Features

Phone Classification



MFCC features are use for Automatic Speech Recognition

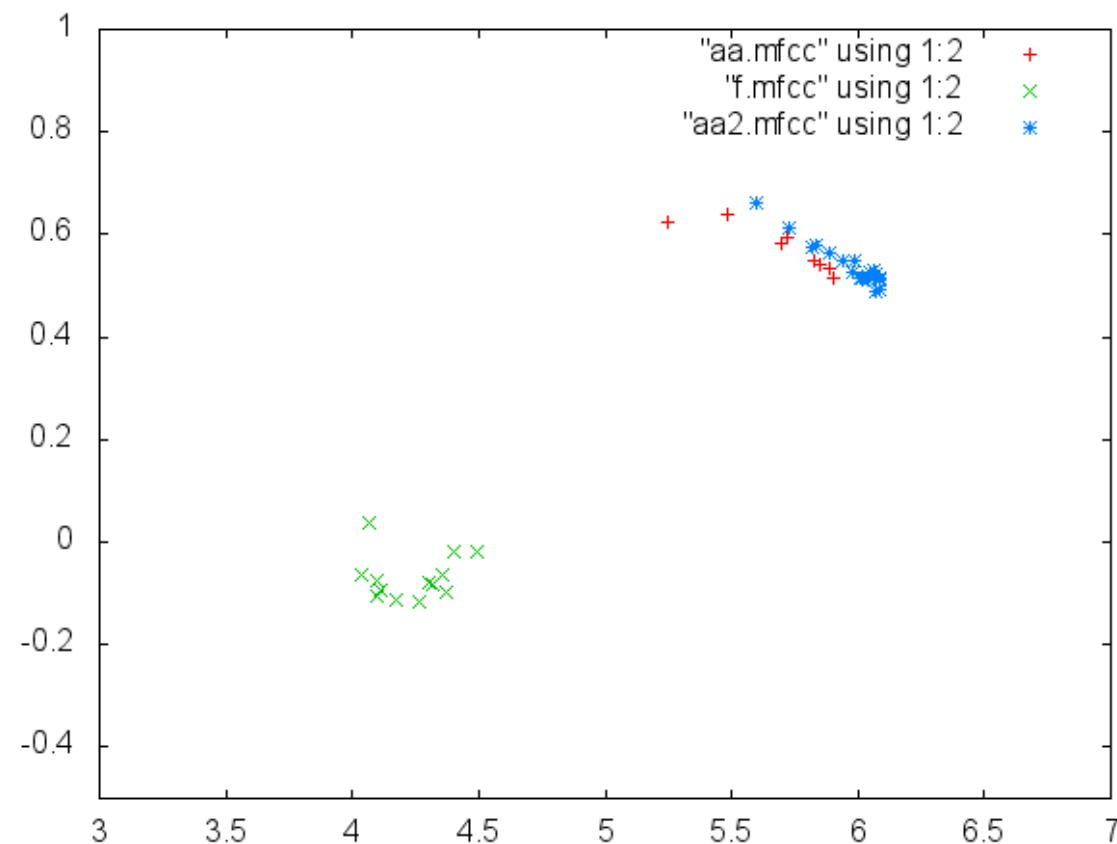
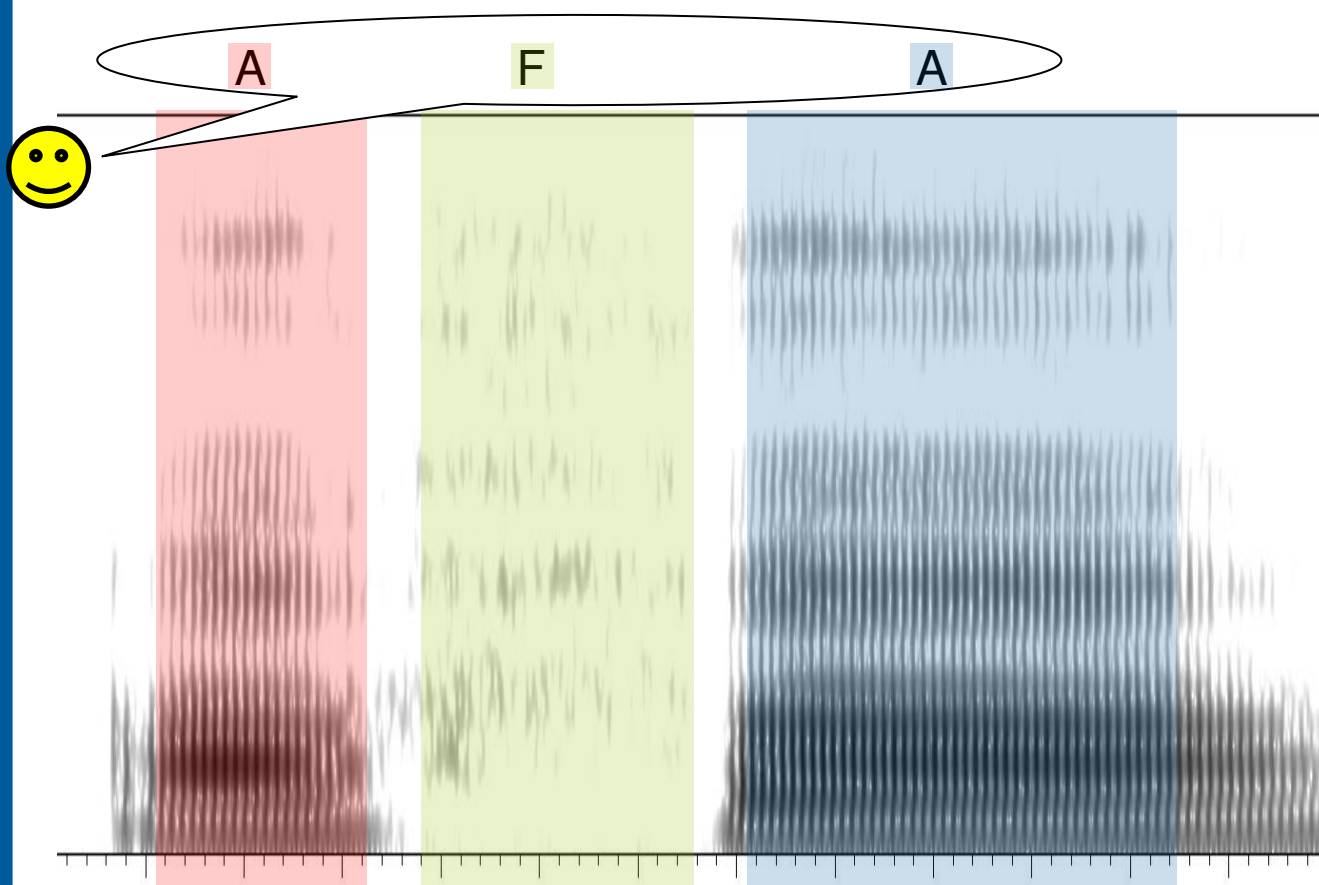


Classification with Speech Features

Phone Classification



MFCC features are use for Automatic Speech Recognition

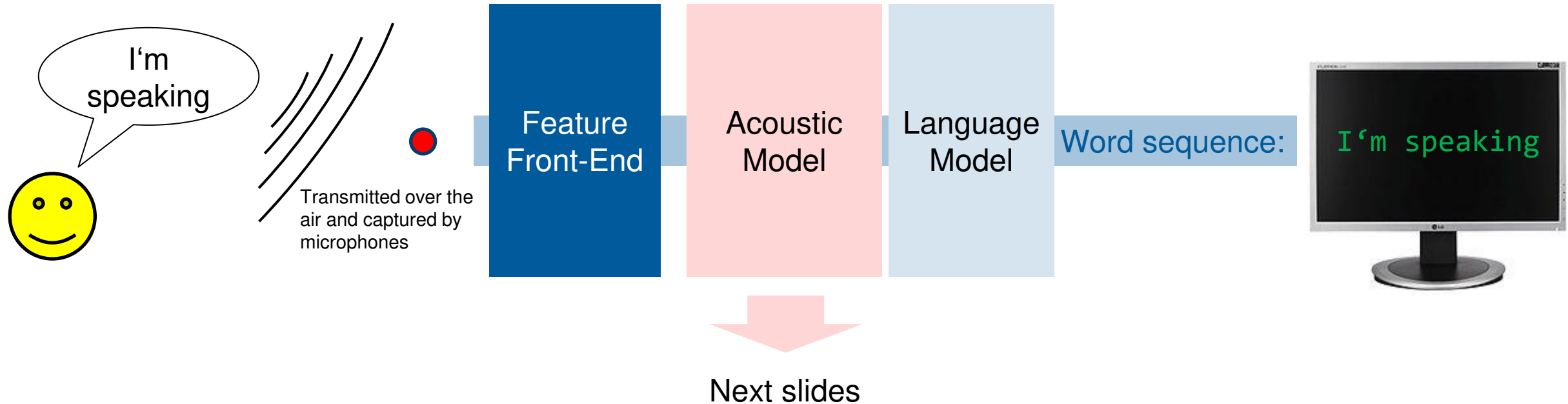


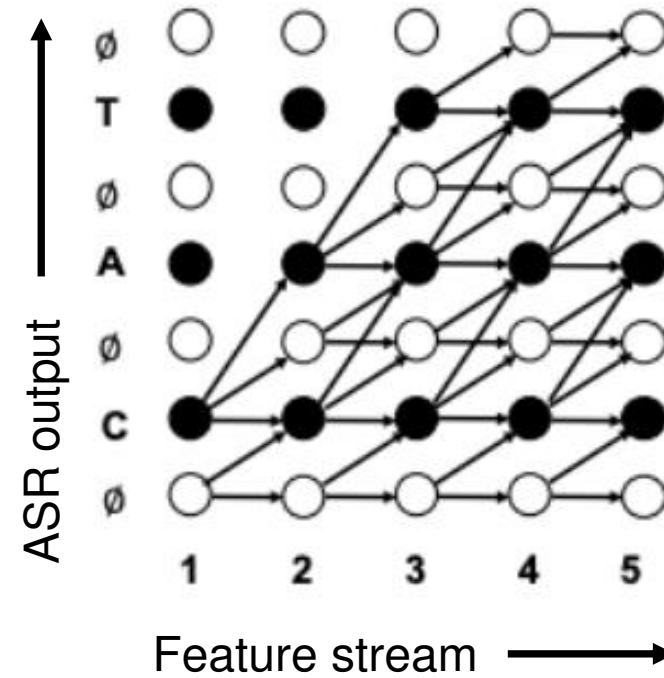
Automatic Speech Recognition

Introduction



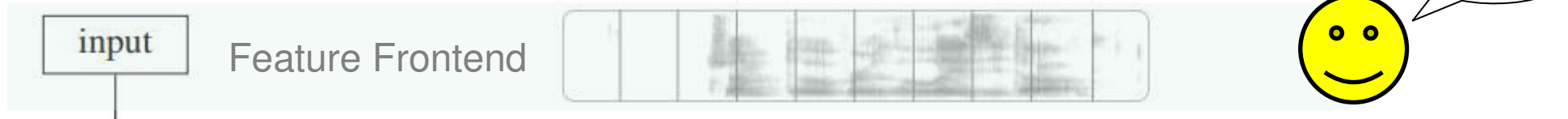
Automatic Speech Recognition, is the technology that allows human beings to use their voices to speak with a computer interface in a way that, in its most sophisticated variations, resembles normal human conversation.





Automatic Speech Recognition

Acoustic Modelling with Connectionist Temporal Classification (CTC) Training.



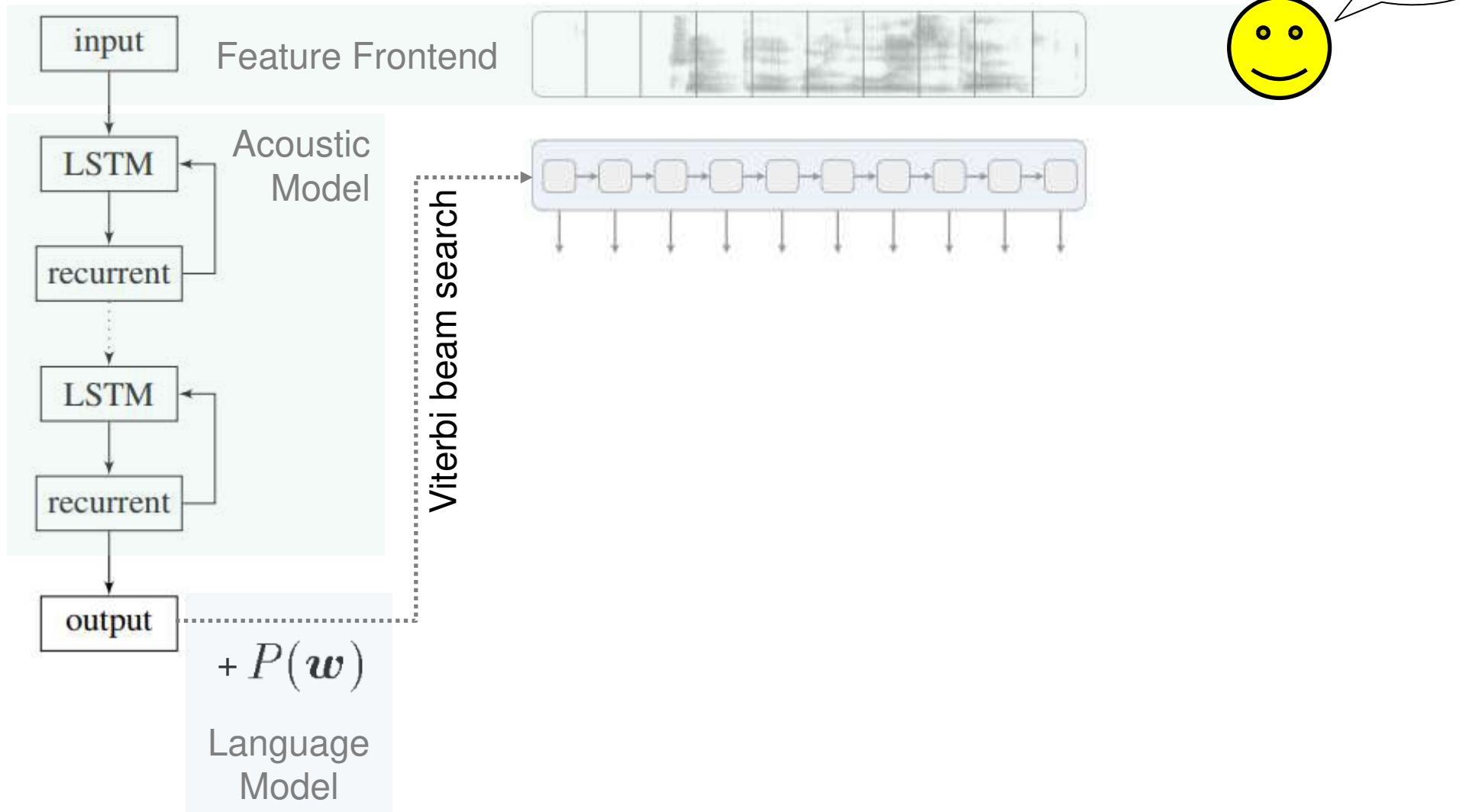
Automatic Speech Recognition

Acoustic Modelling with Connectionist Temporal Classification (CTC) Training.



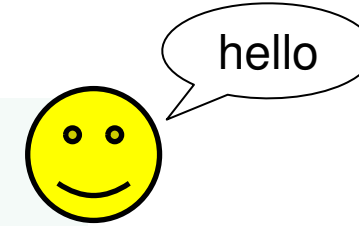
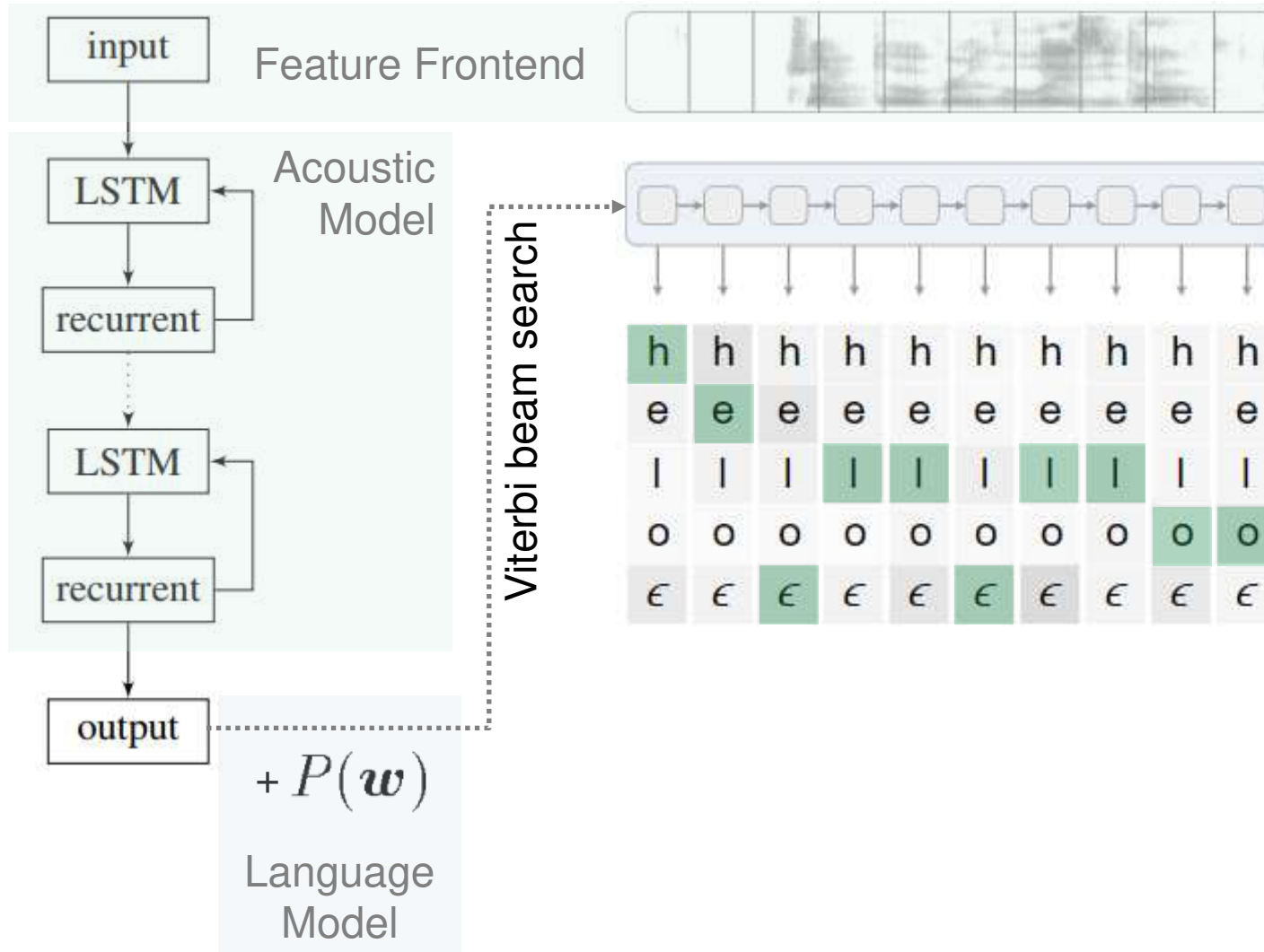
Automatic Speech Recognition

Acoustic Modelling with Connectionist Temporal Classification (CTC) Training.



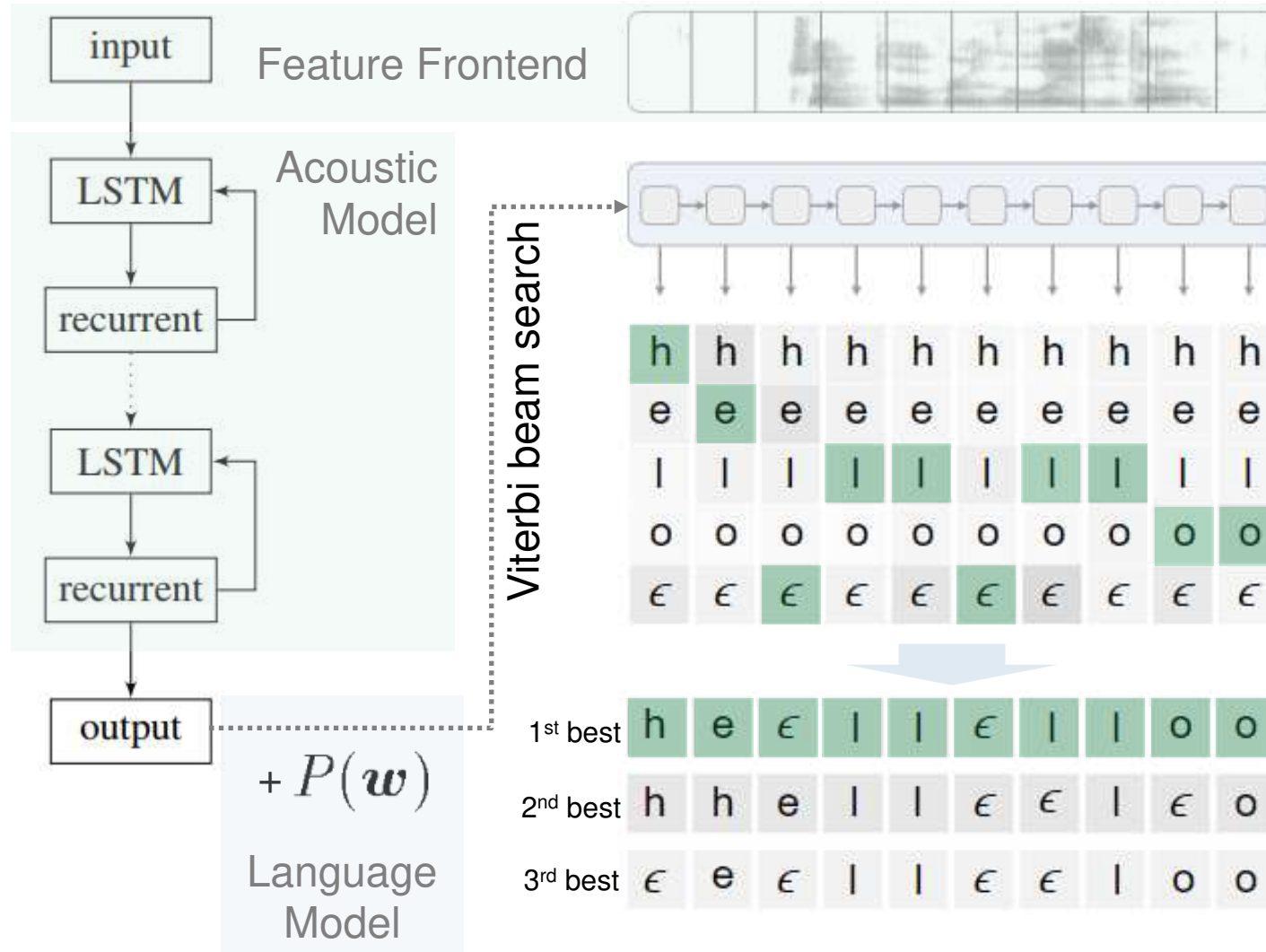
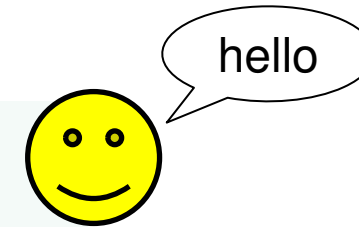
Automatic Speech Recognition

Acoustic Modelling with Connectionist Temporal Classification (CTC) Training.



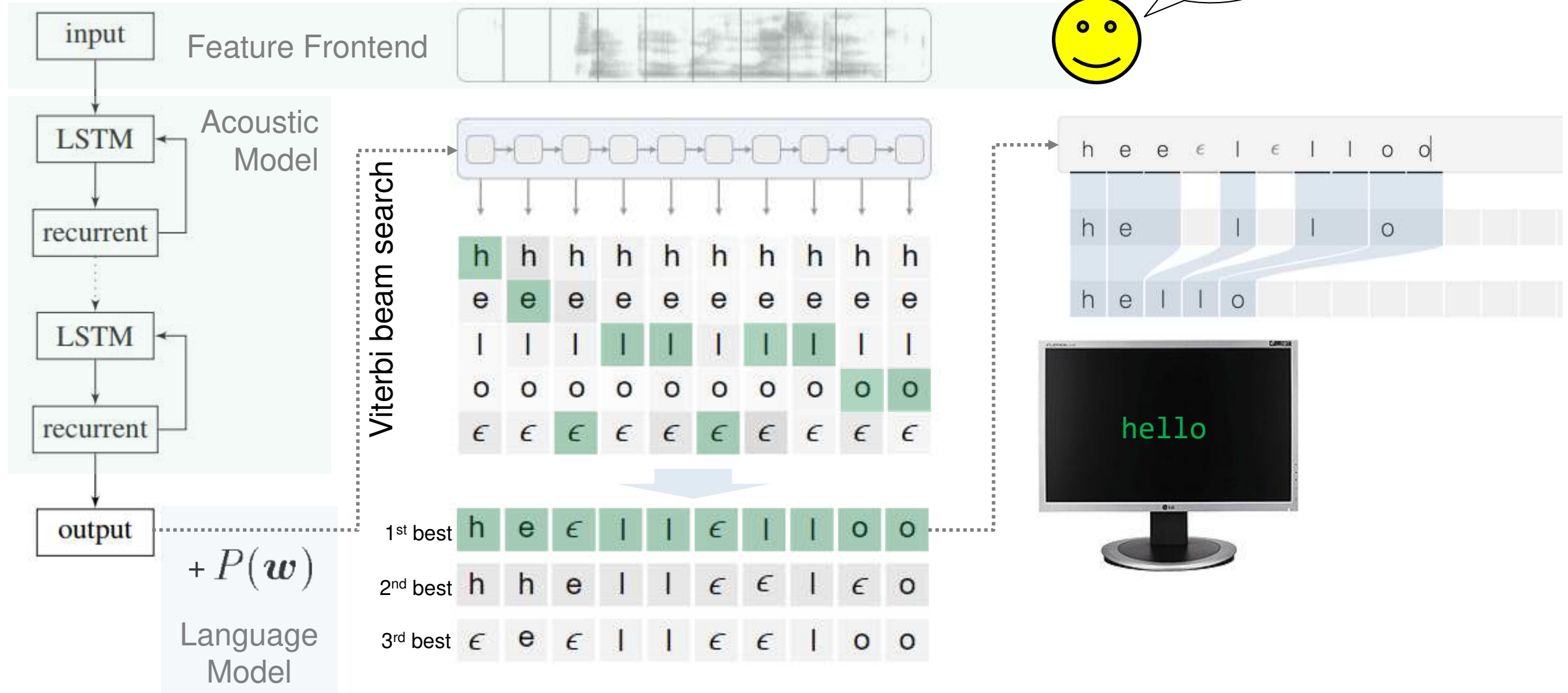
Automatic Speech Recognition

Acoustic Modelling with Connectionist Temporal Classification (CTC) Training.



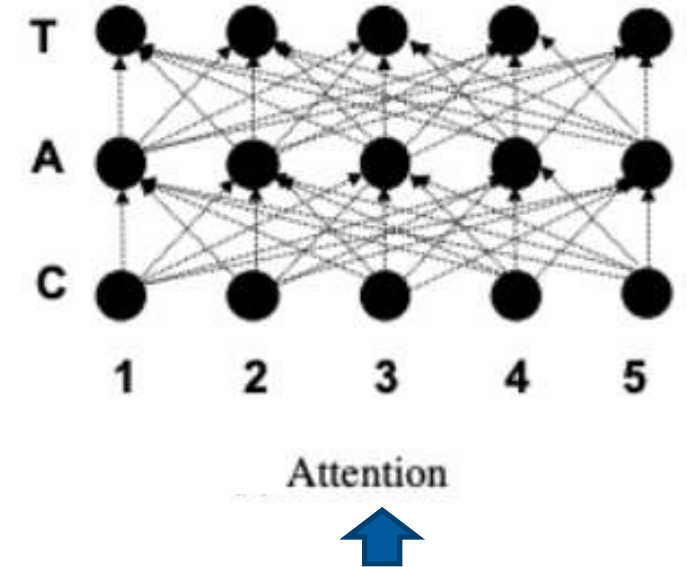
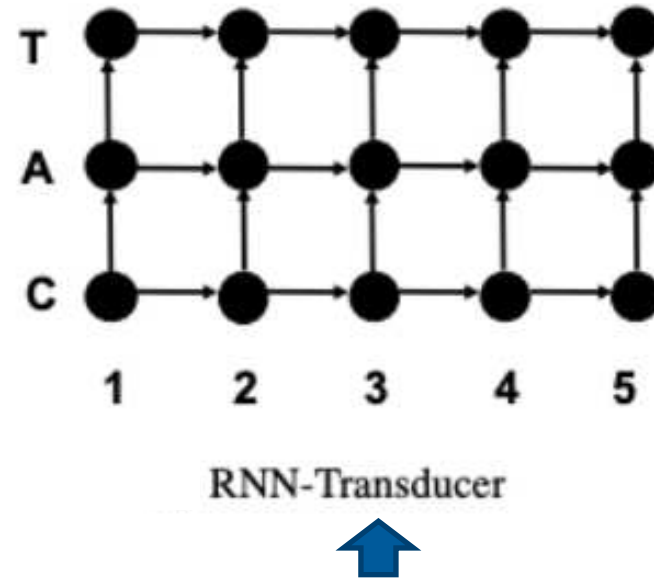
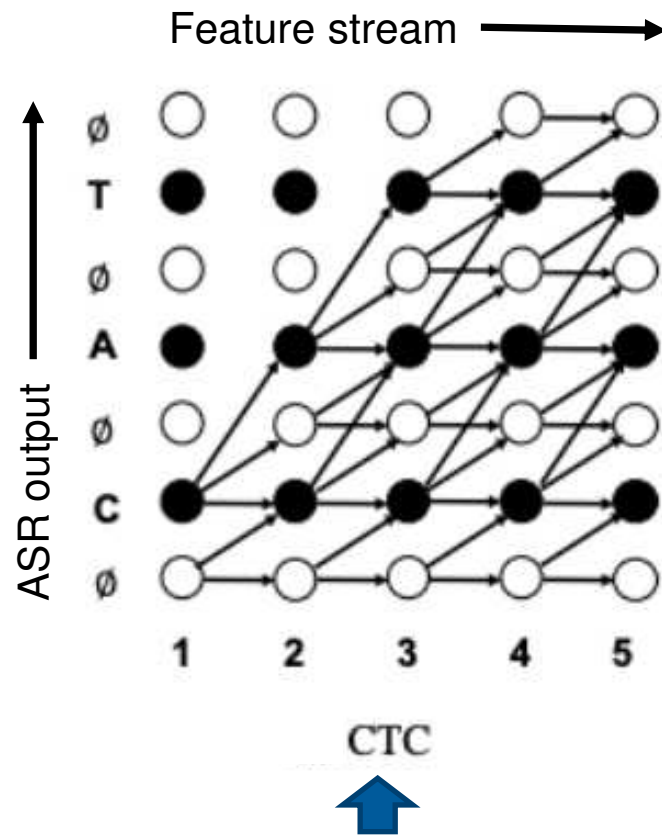
Automatic Speech Recognition

Acoustic Modelling with Connectionist Temporal Classification (CTC) Training.



How is it all connected?

CTC vs. RNN-T vs. Attention

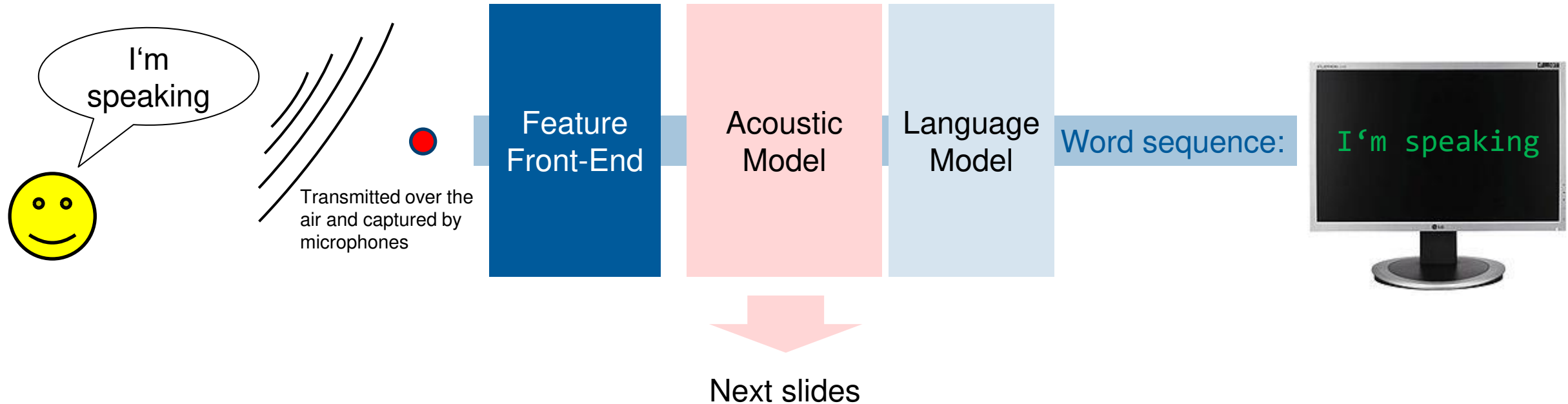


Automatic Speech Recognition

Introduction

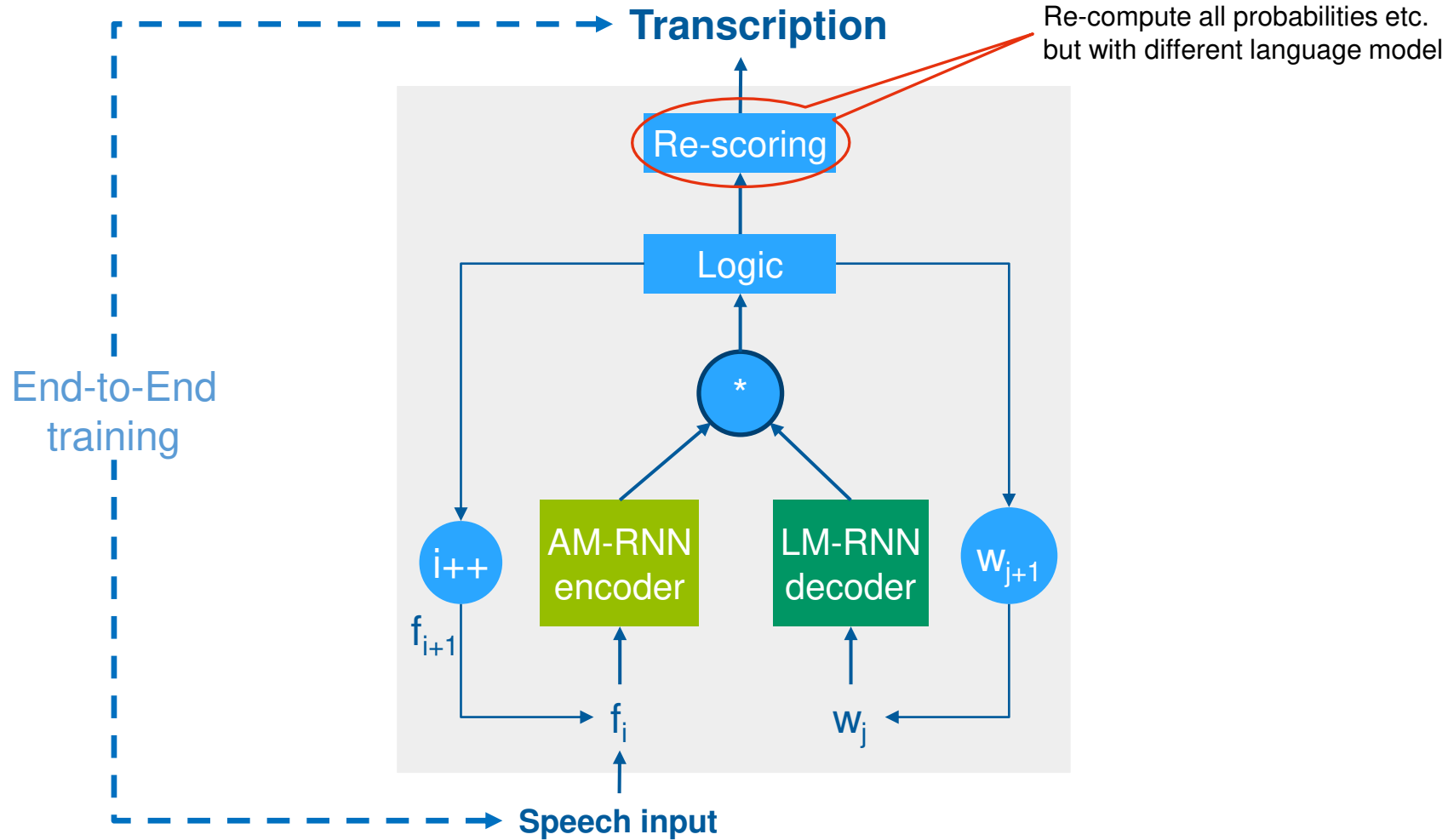


Automatic Speech Recognition, is the technology that allows human beings to use their voices to speak with a computer interface in a way that, in its most sophisticated variations, resembles normal human conversation.



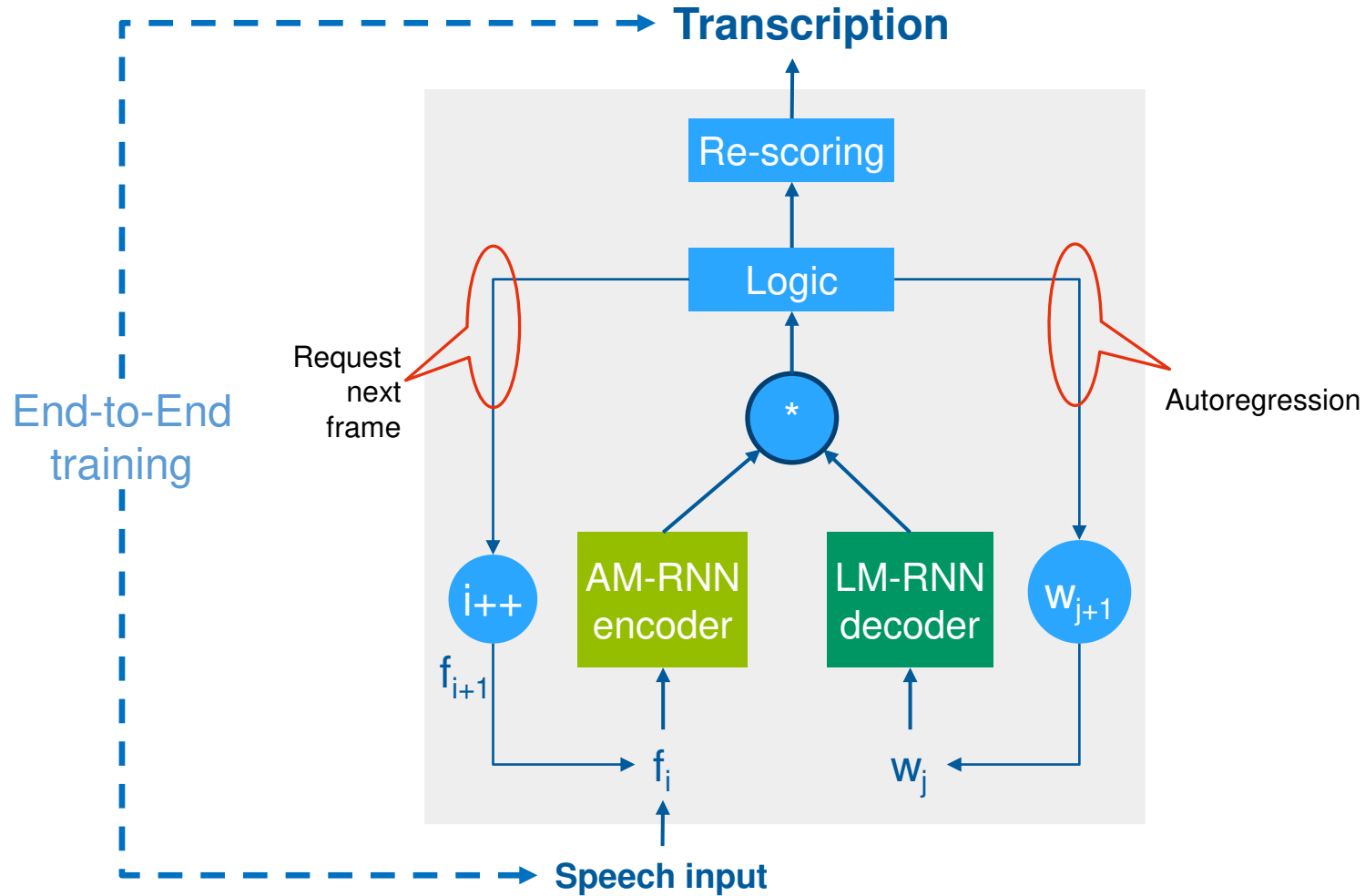
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



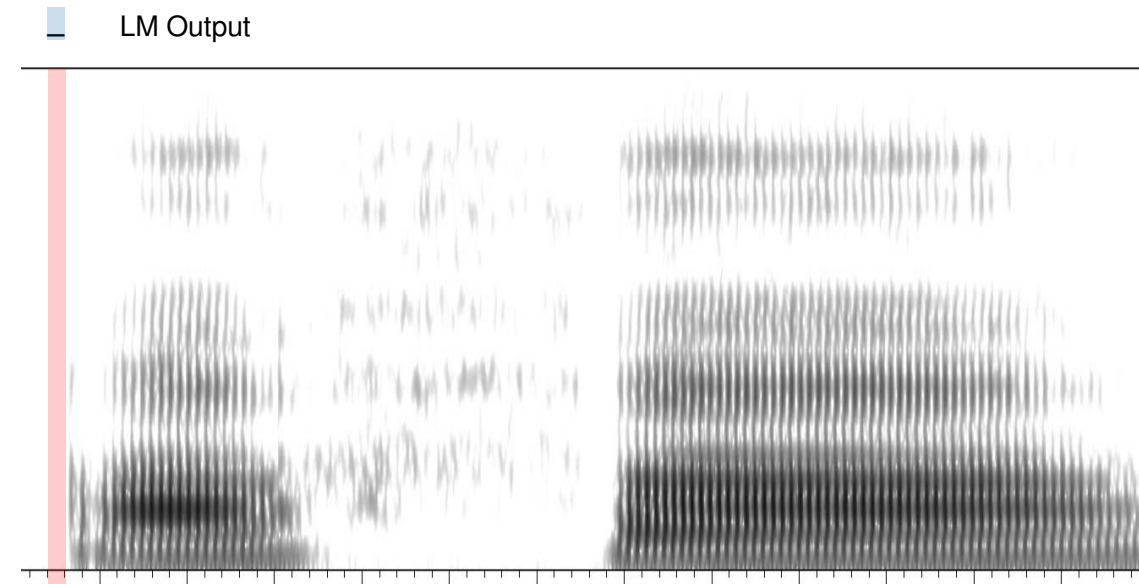
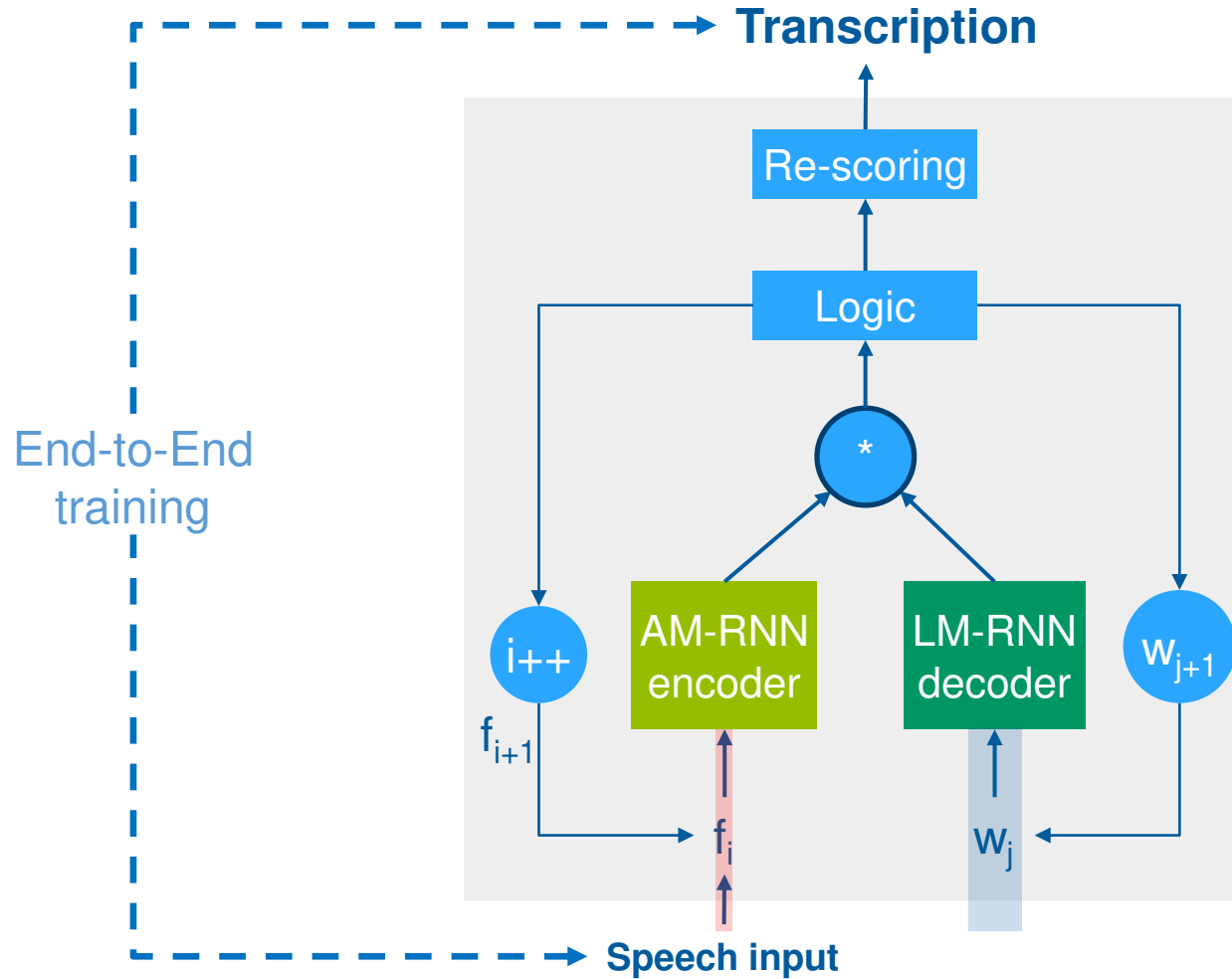
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



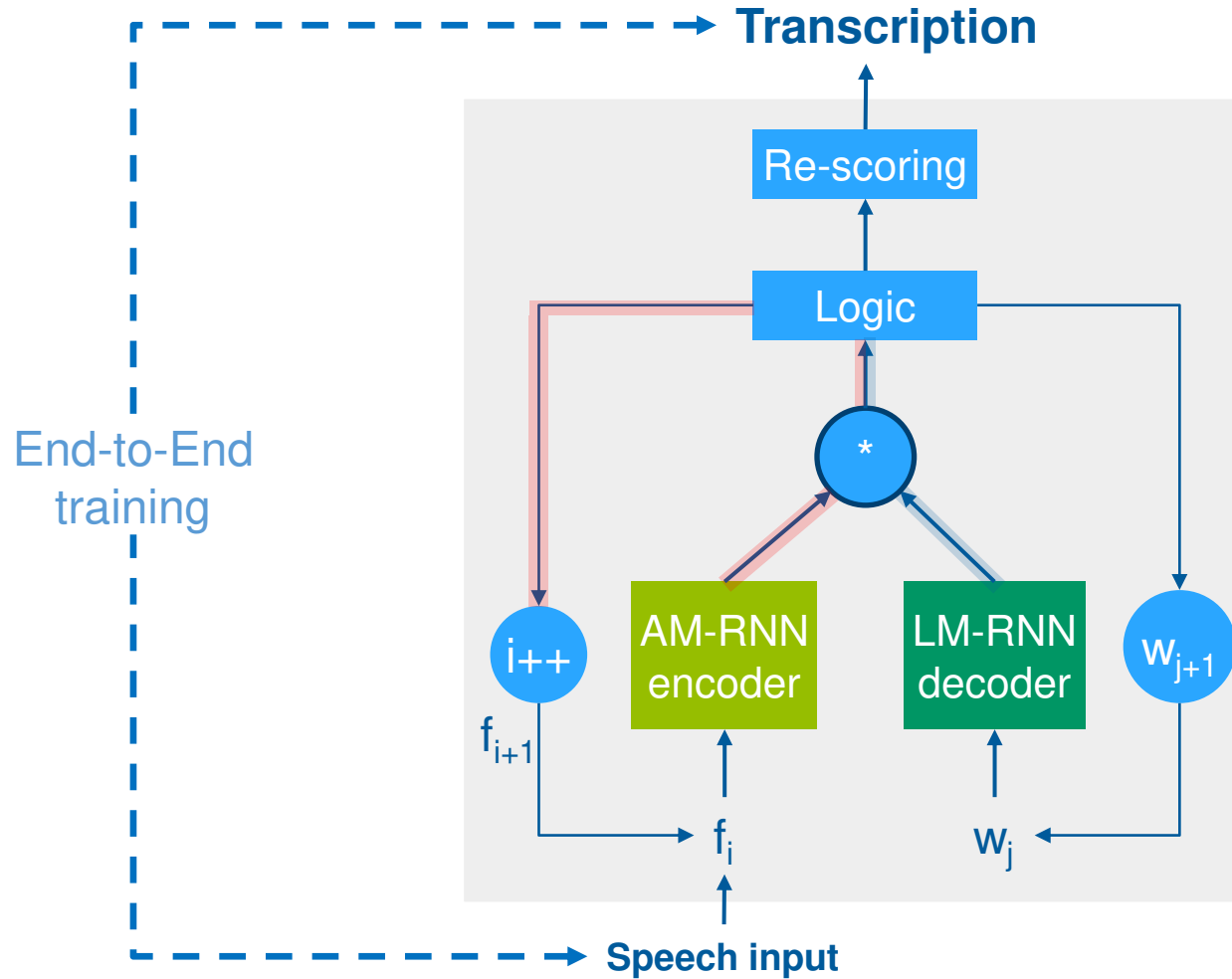
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



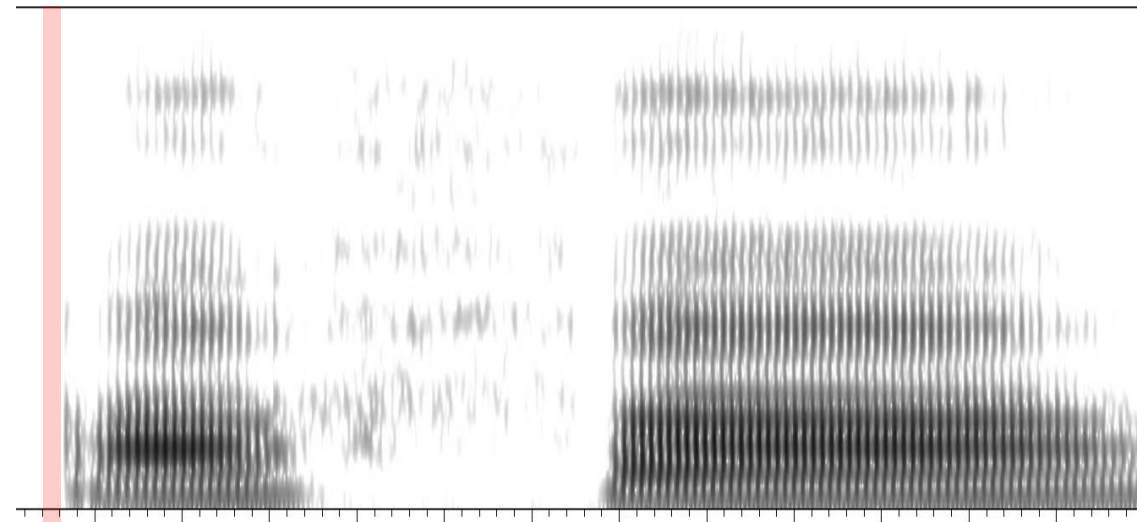
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



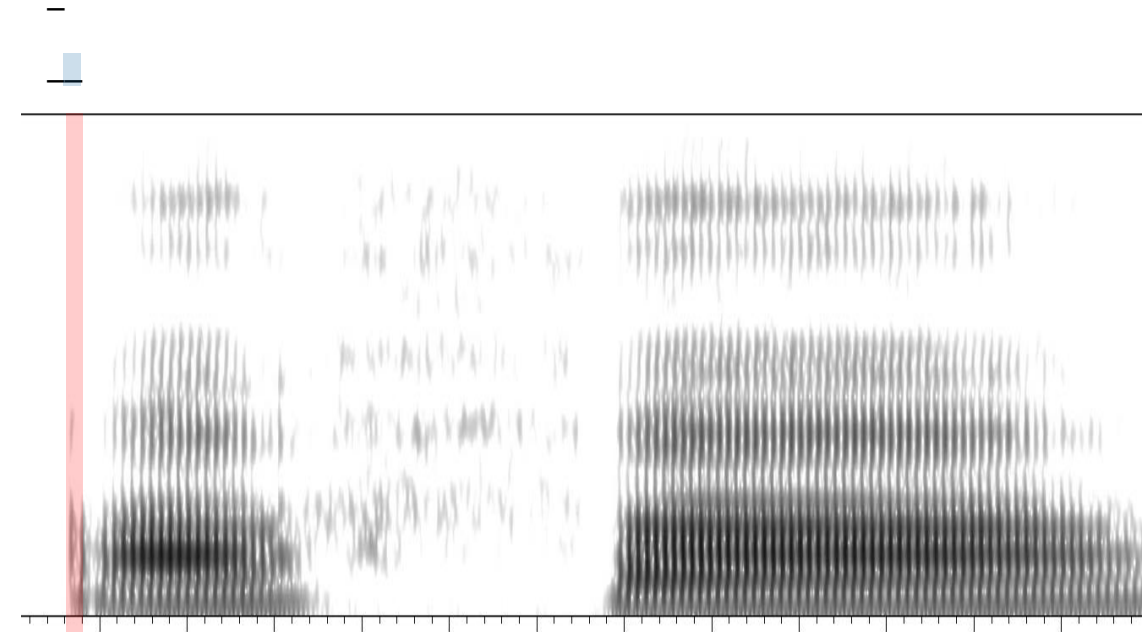
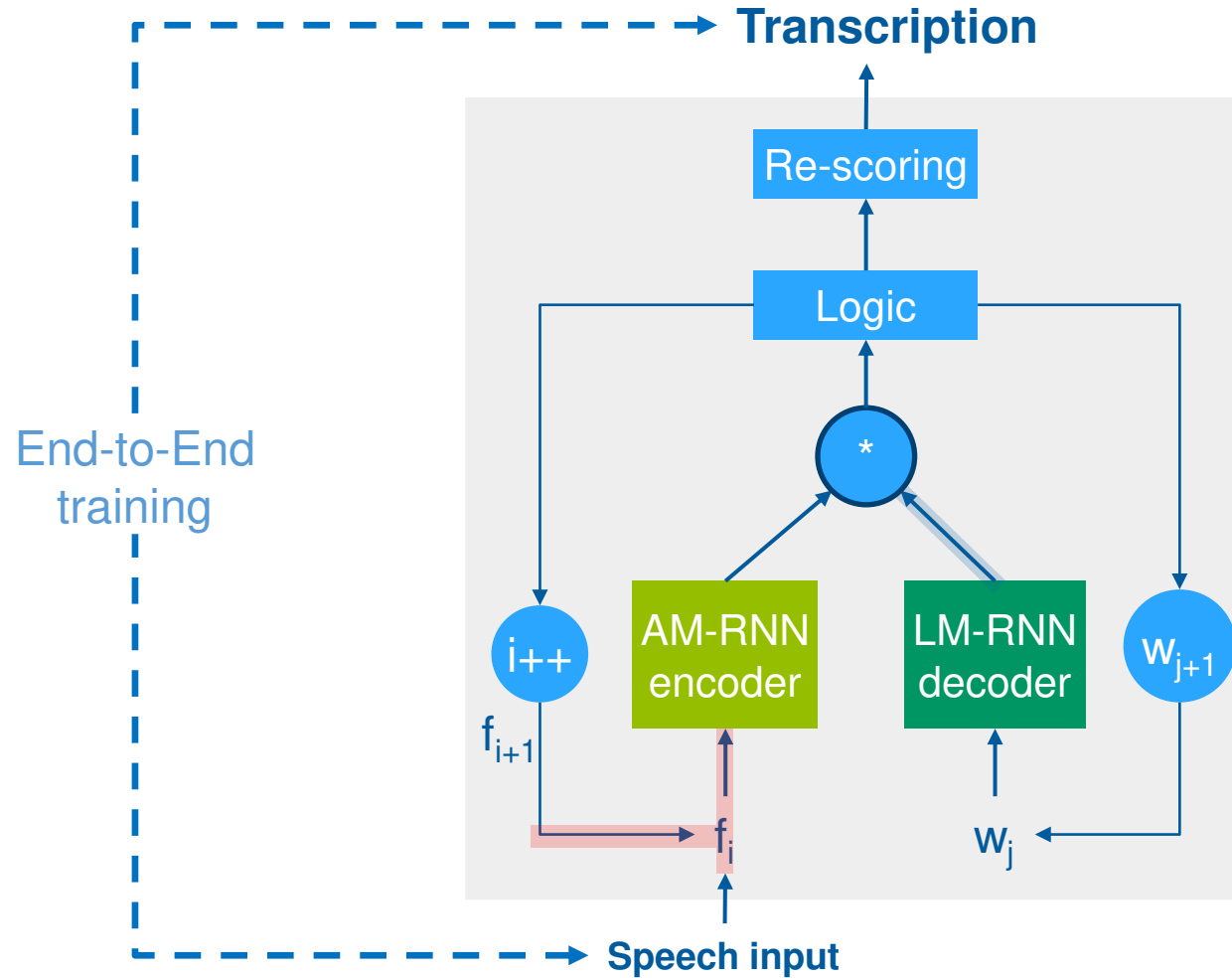
— Transcription

— LM Output



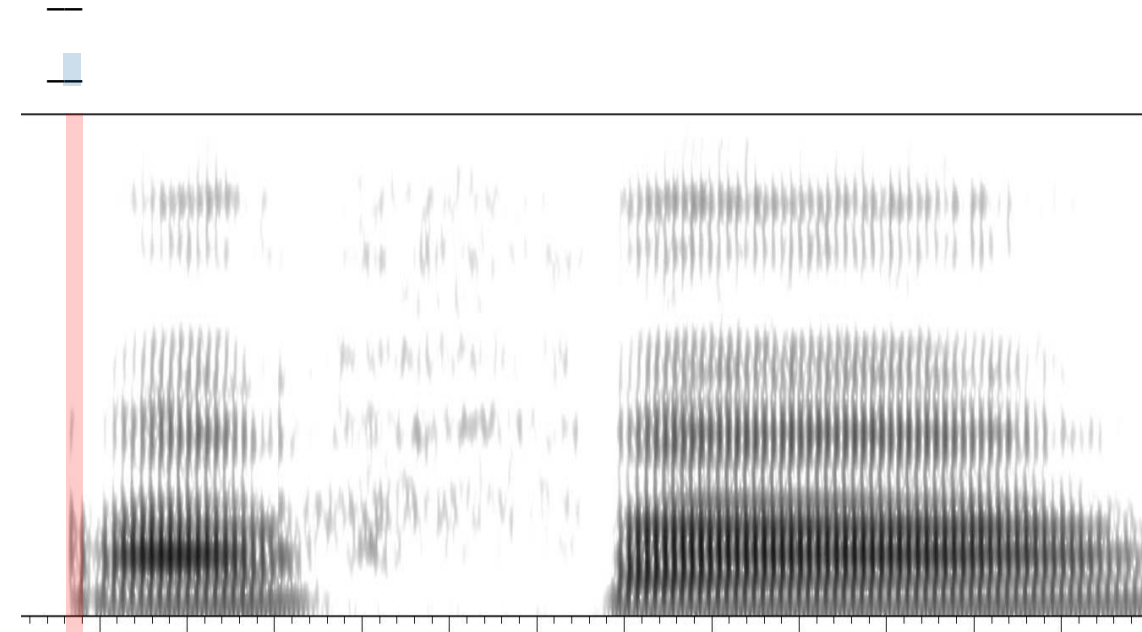
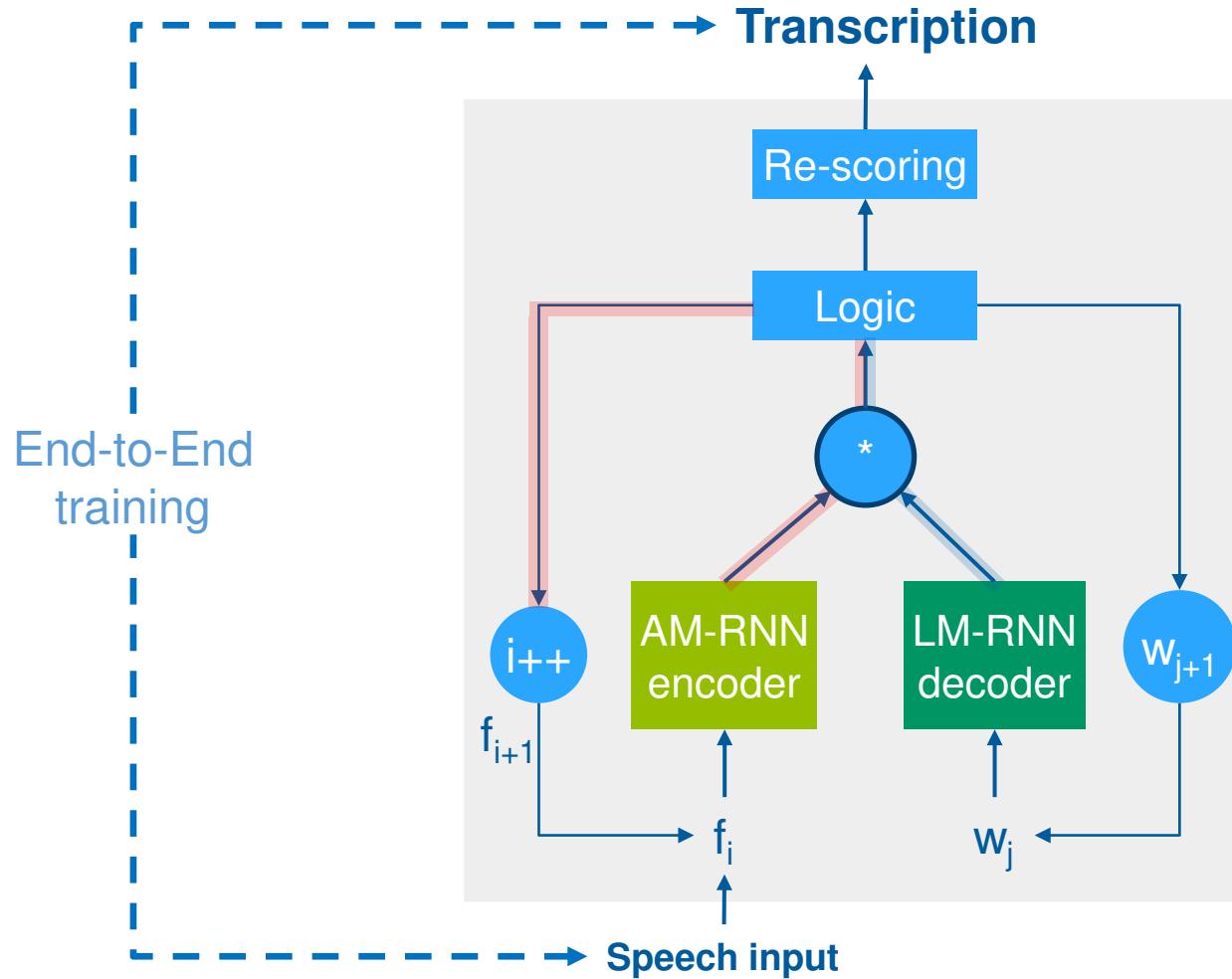
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



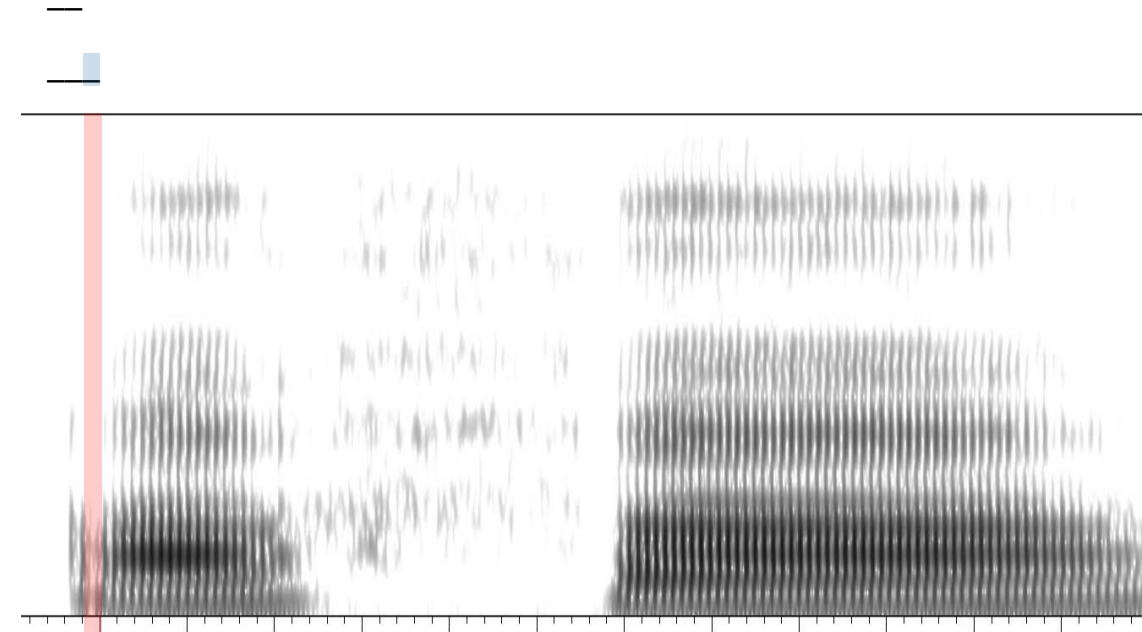
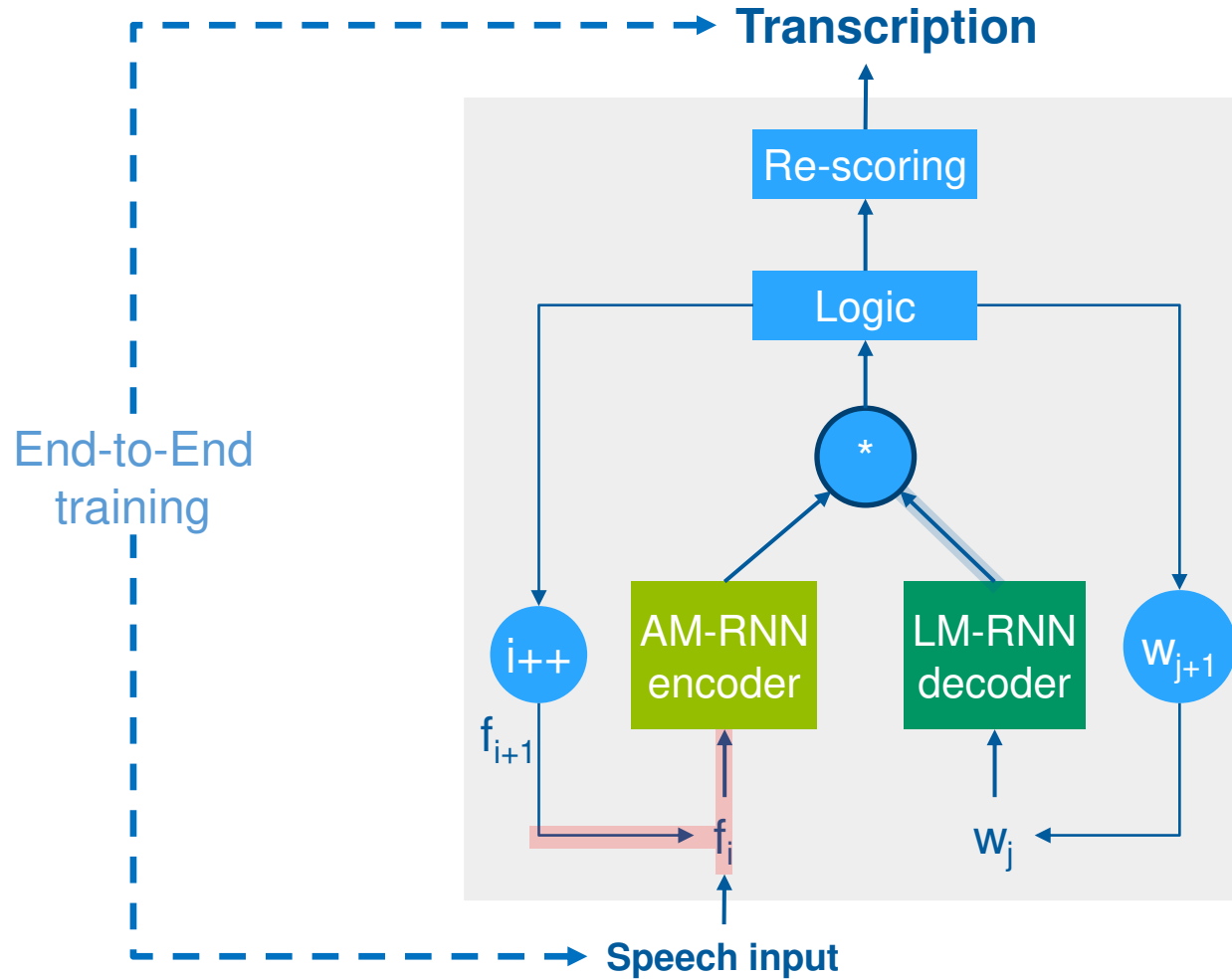
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



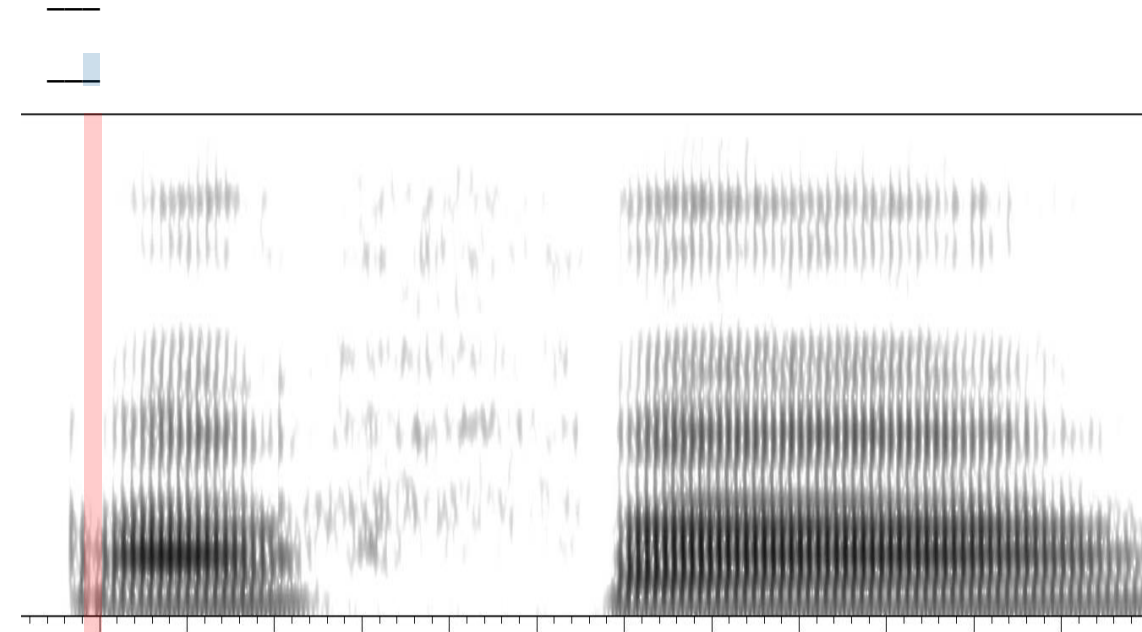
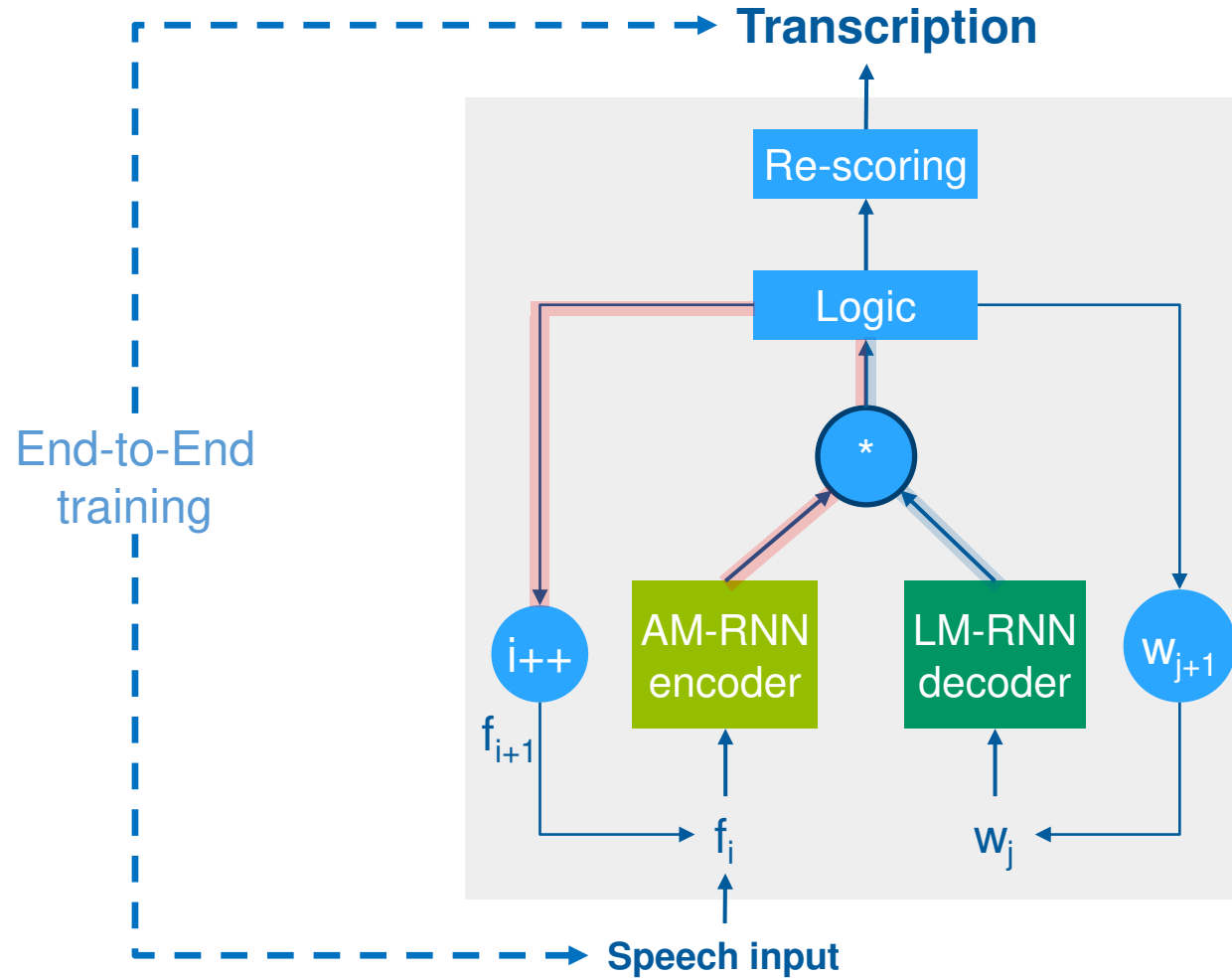
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



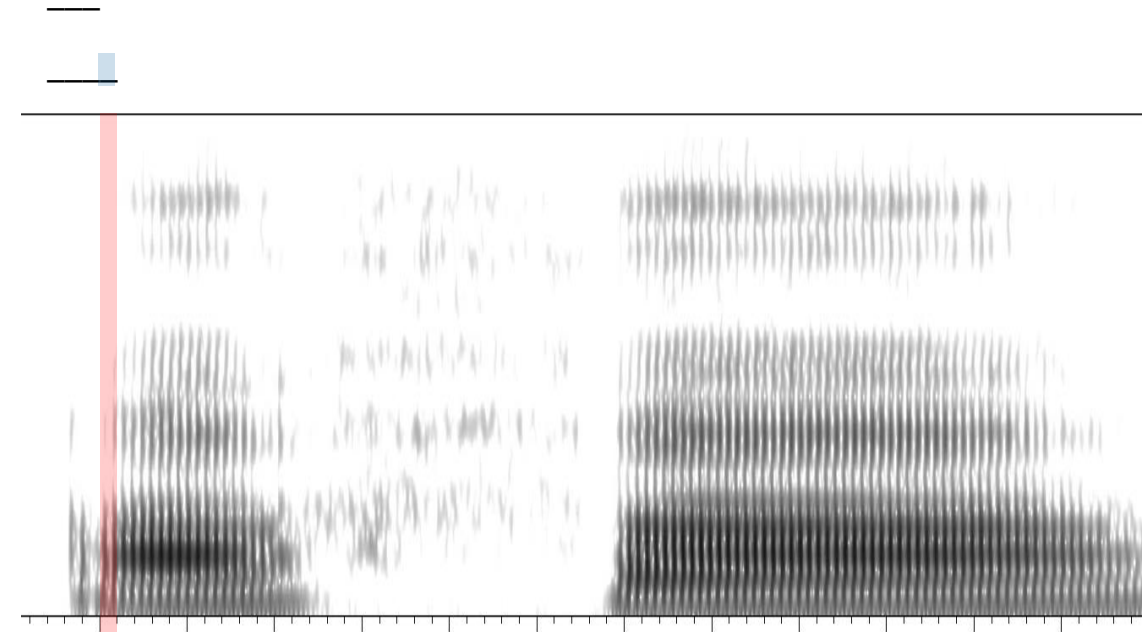
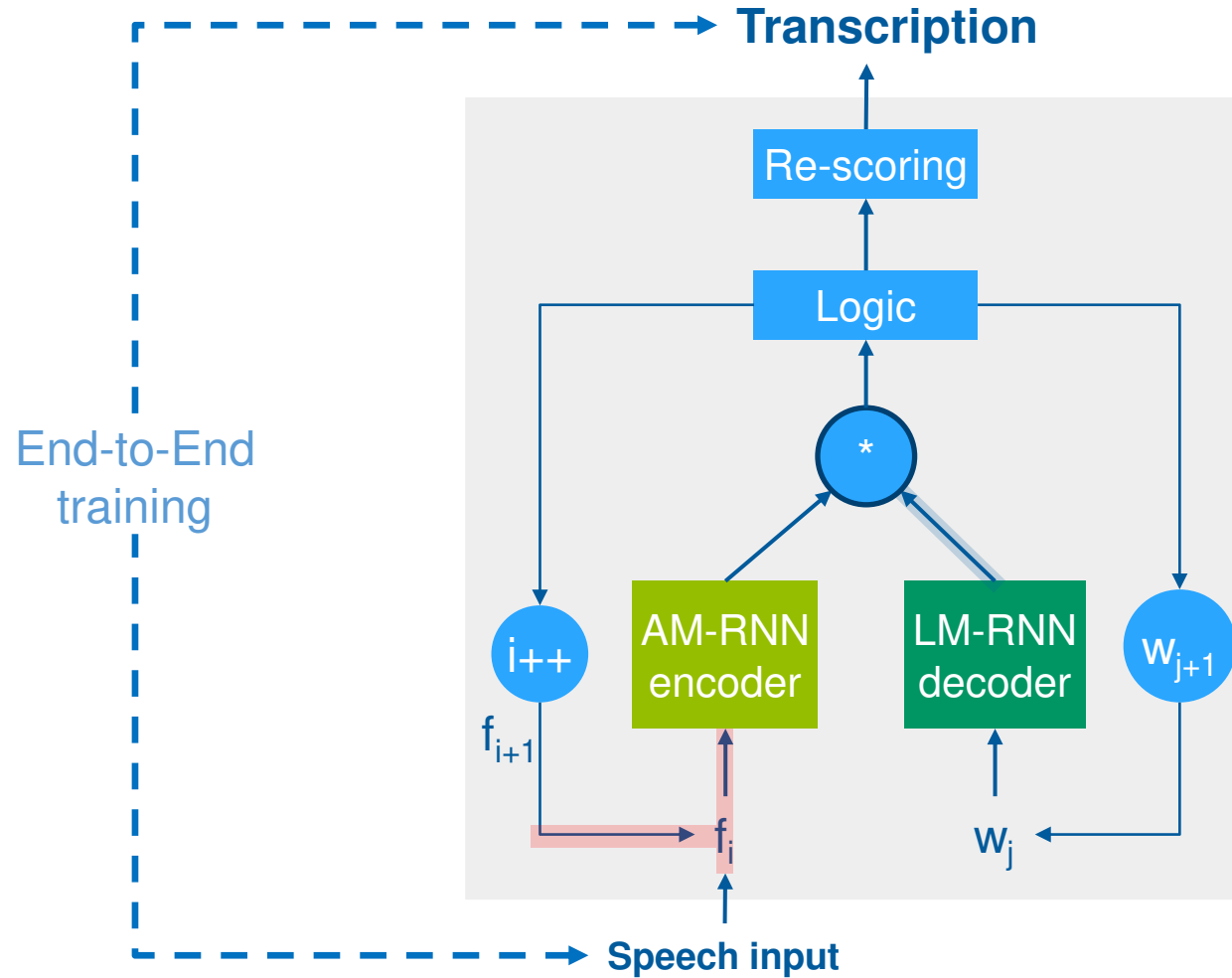
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



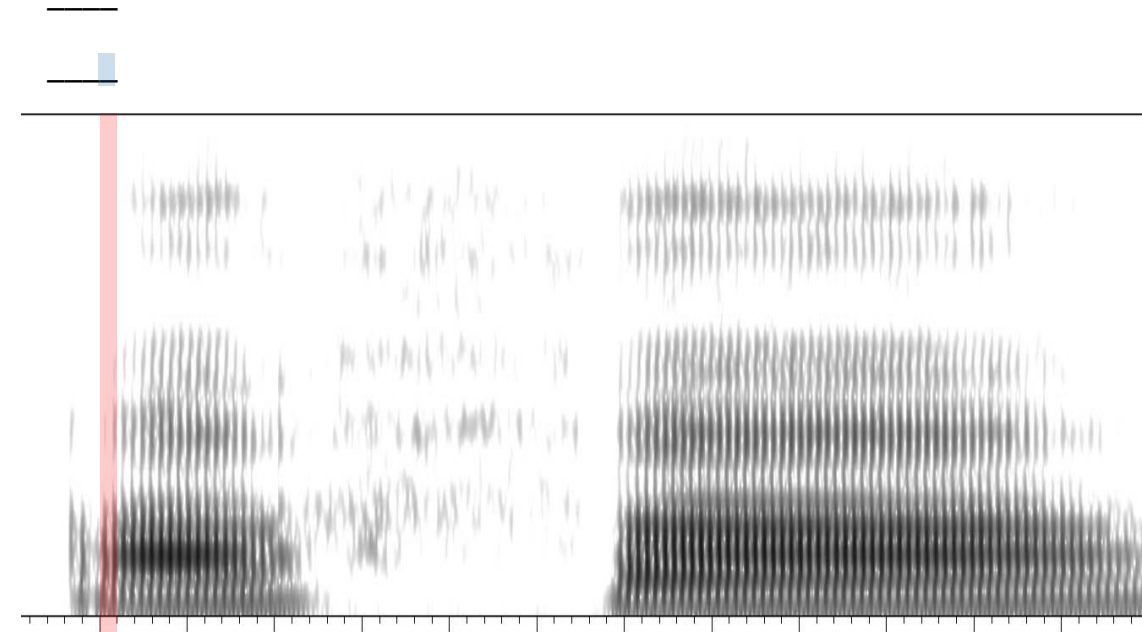
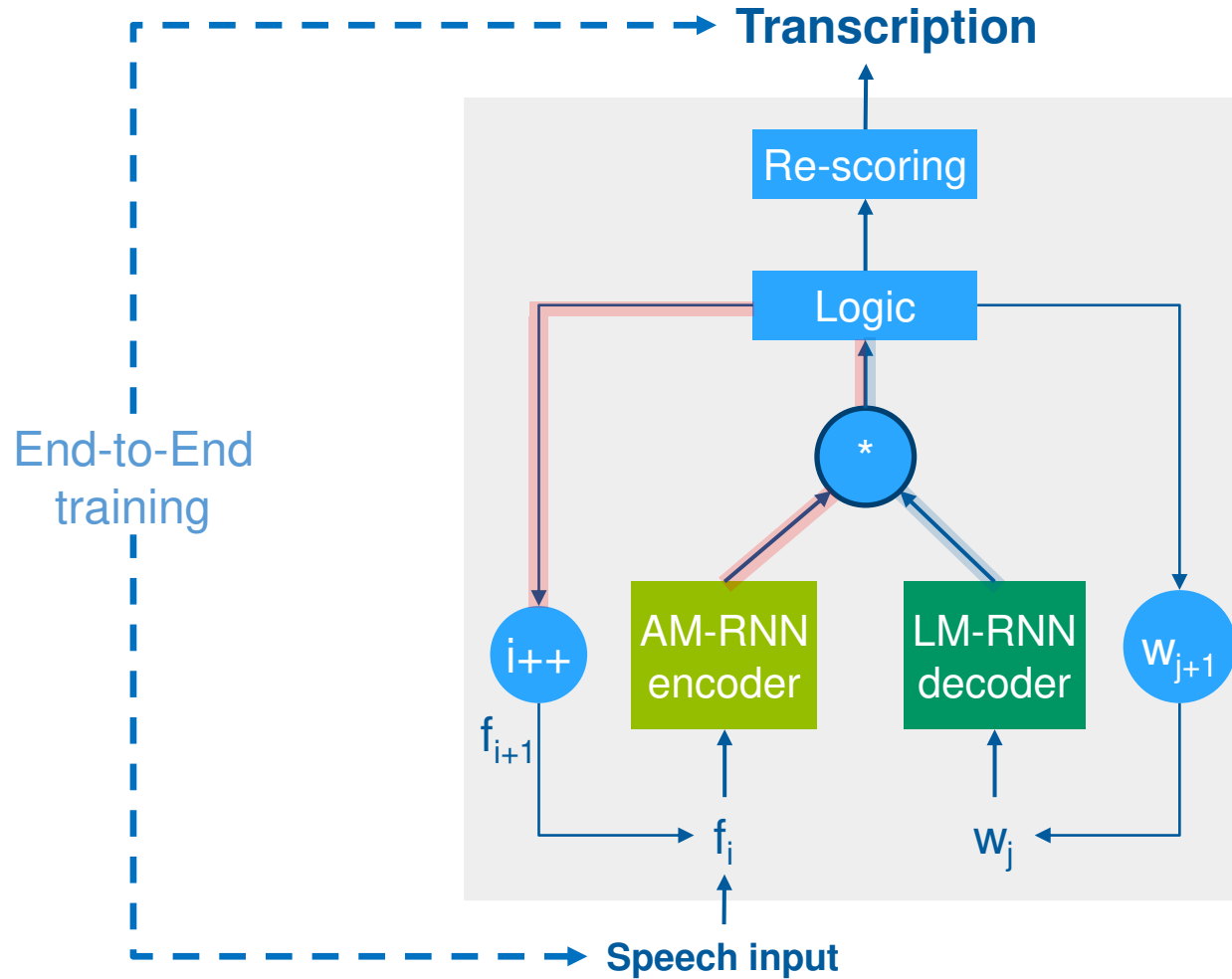
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



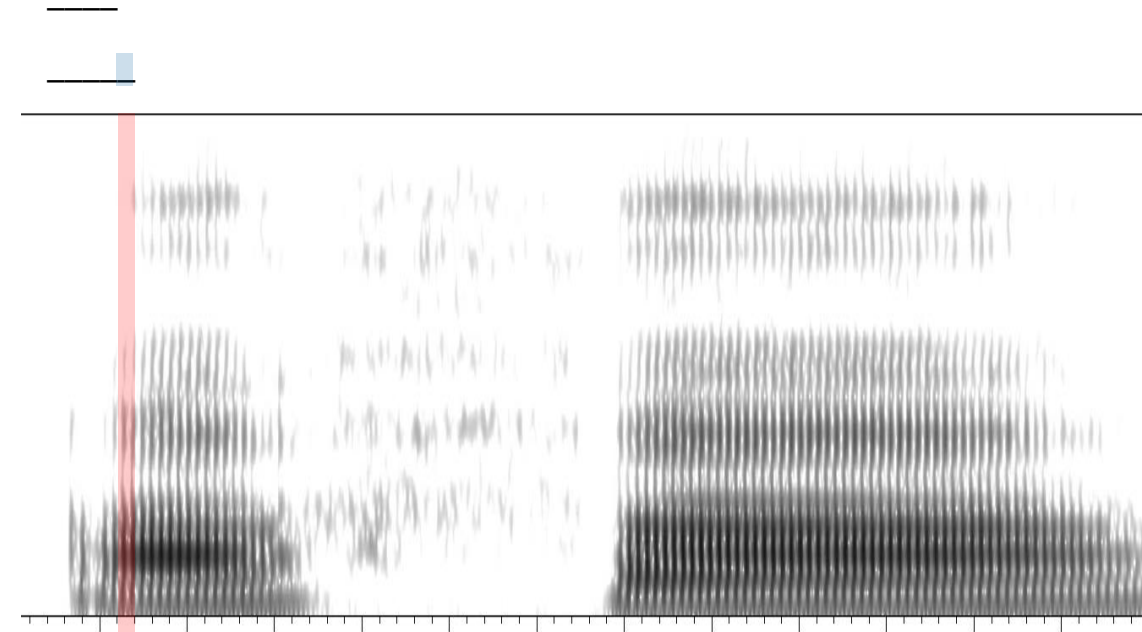
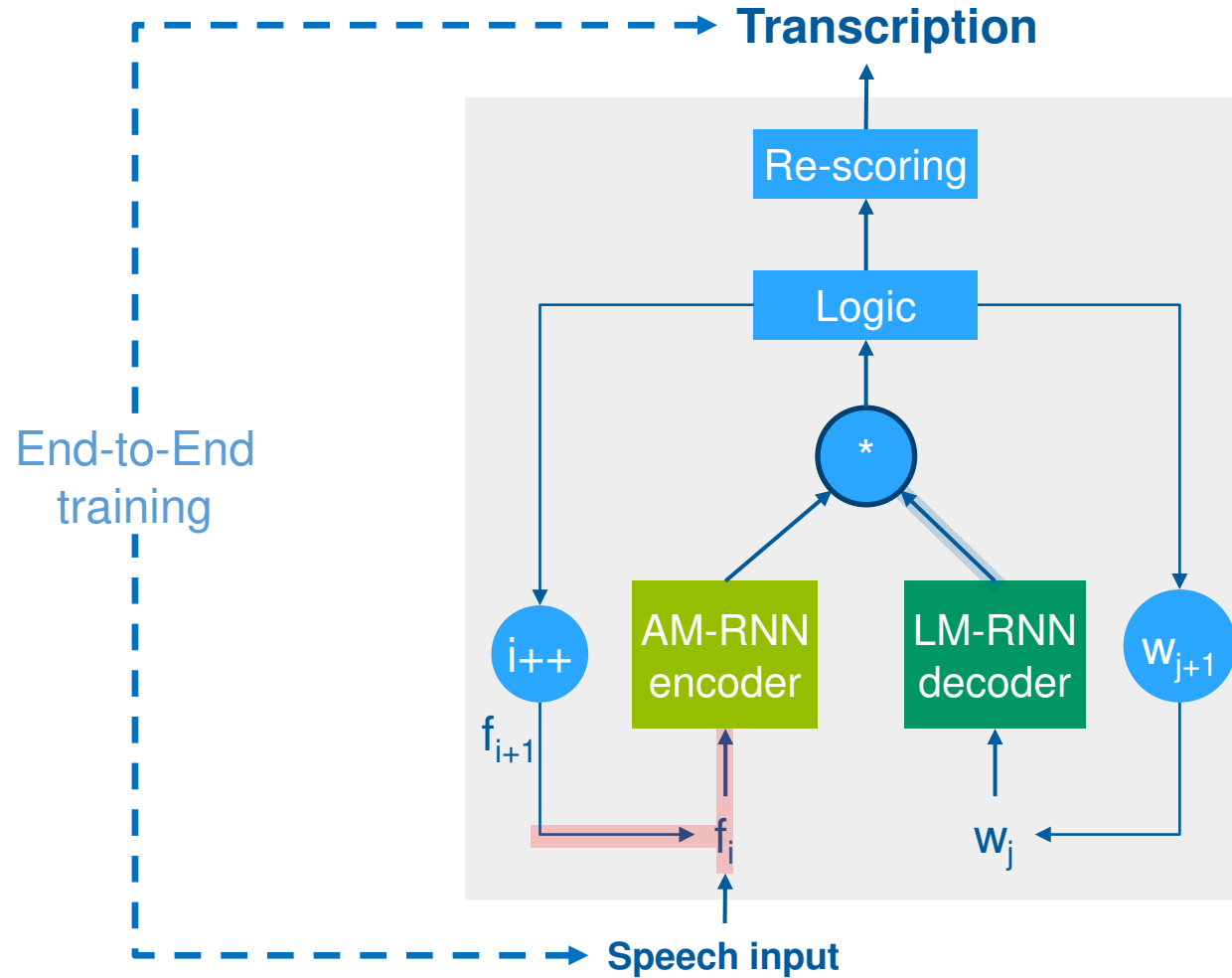
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



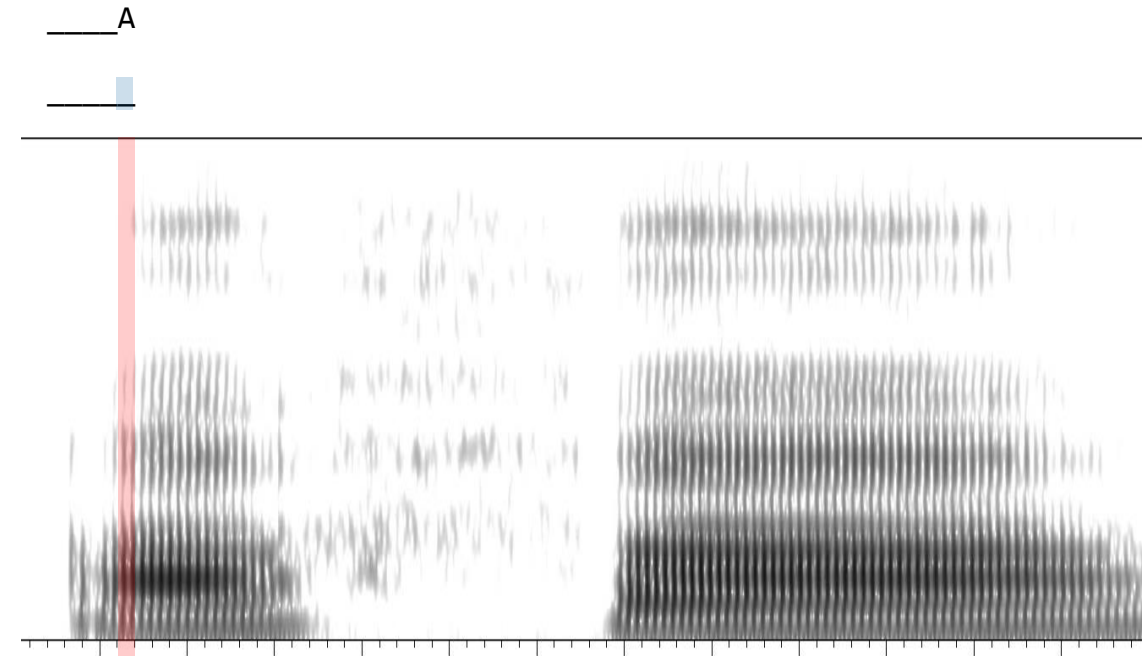
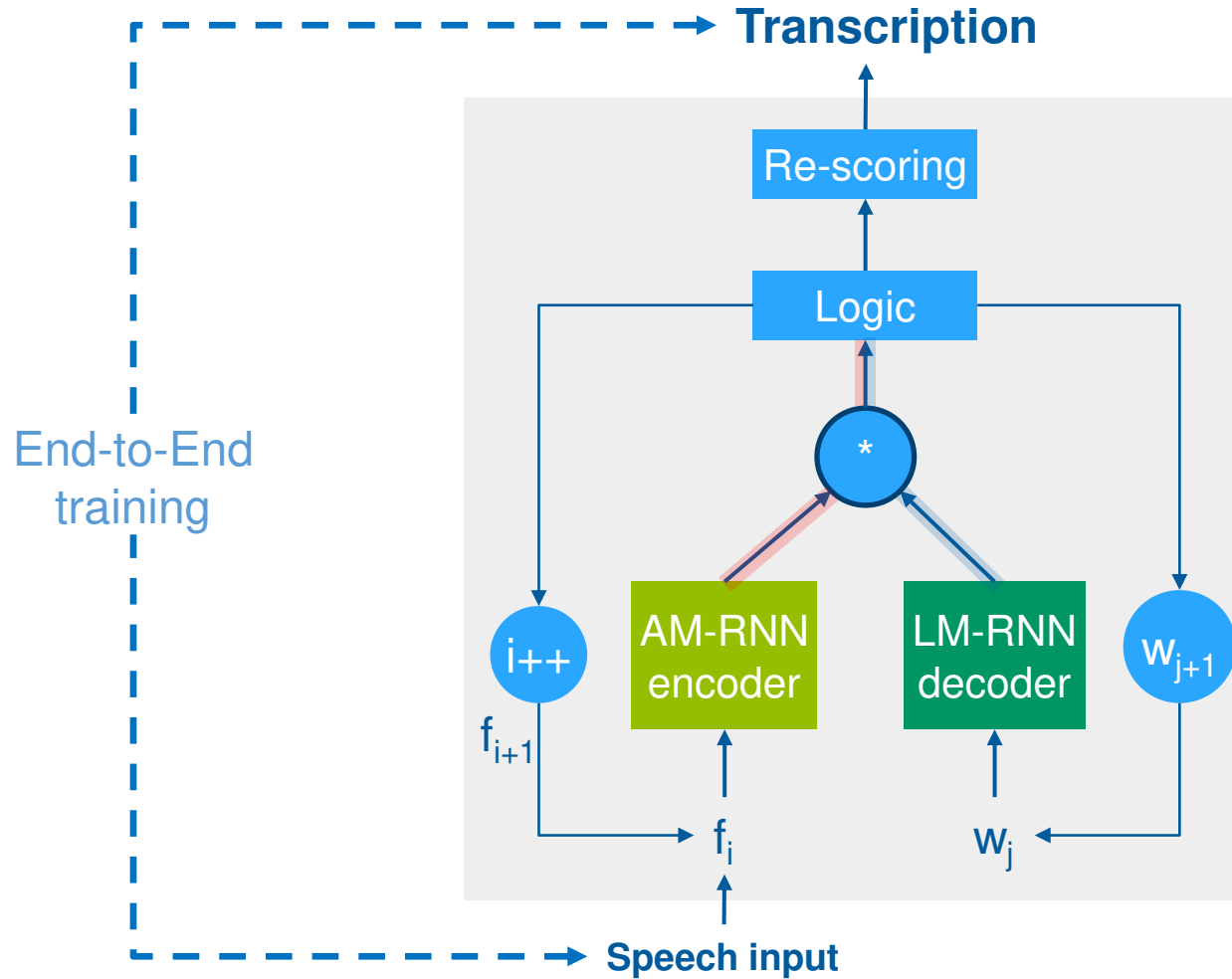
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



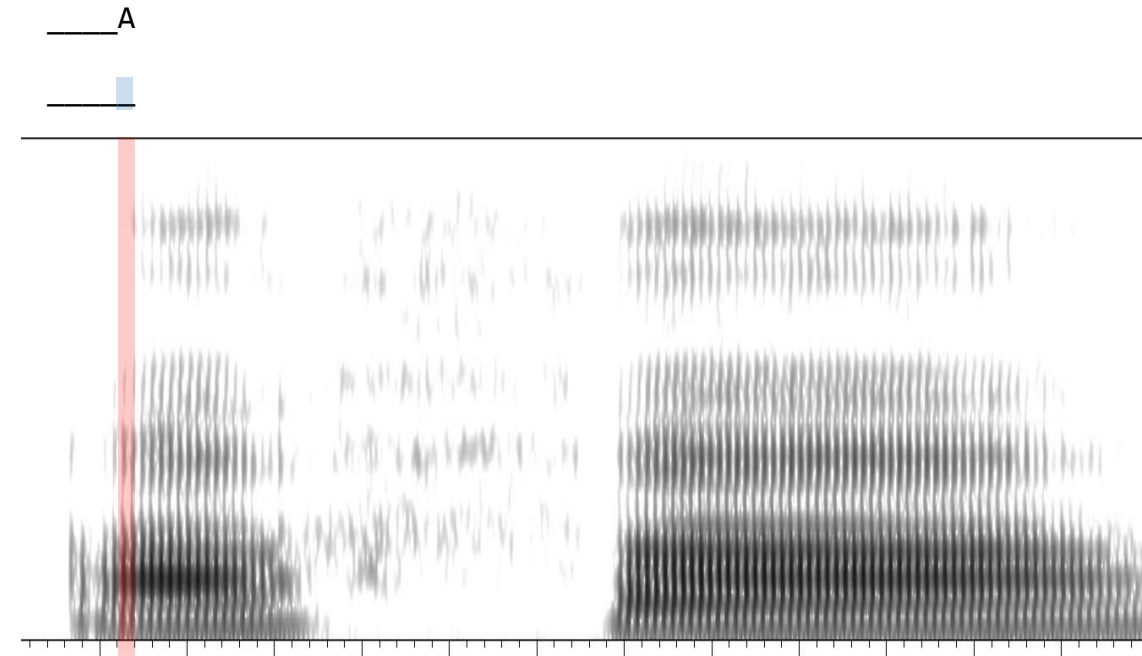
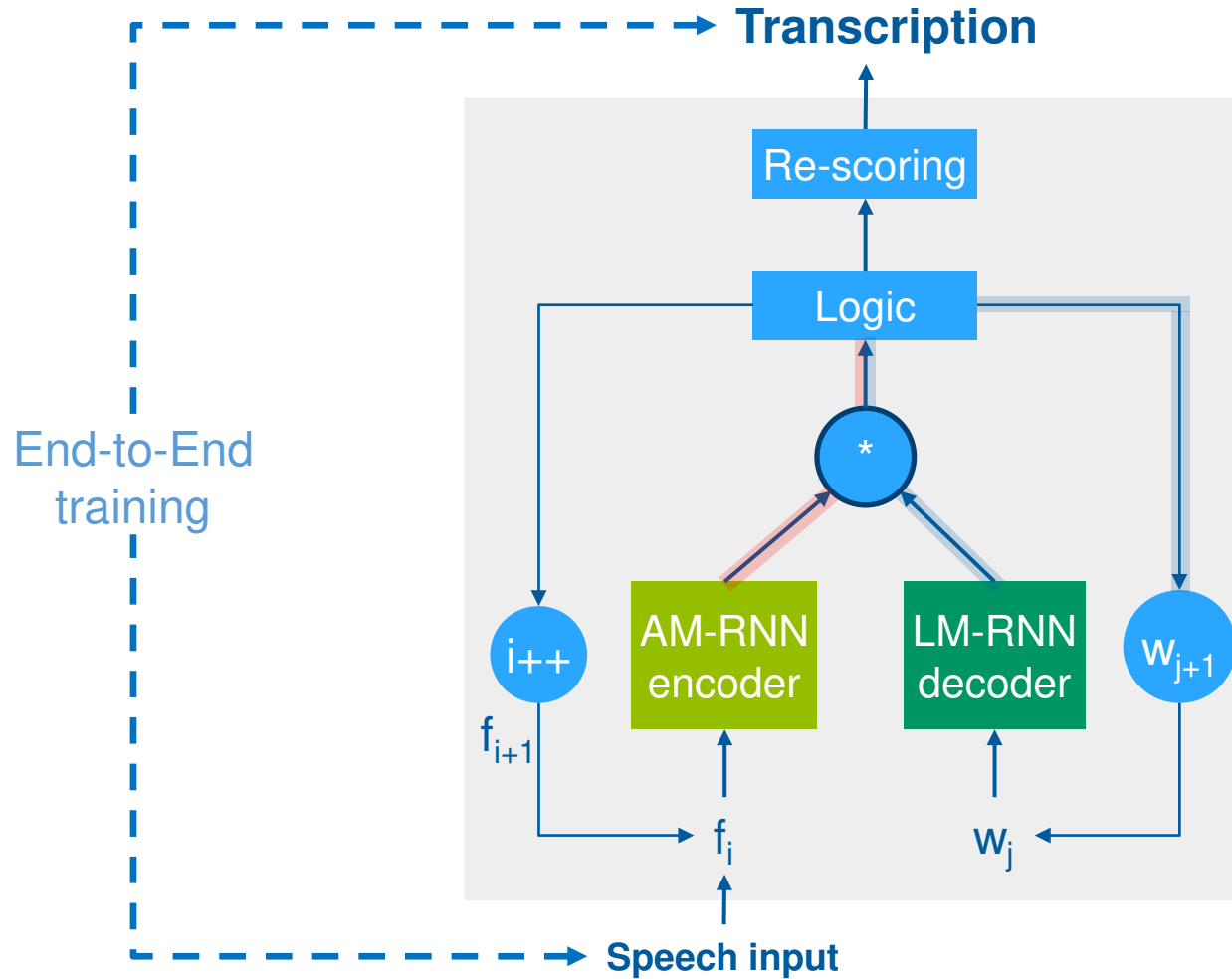
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



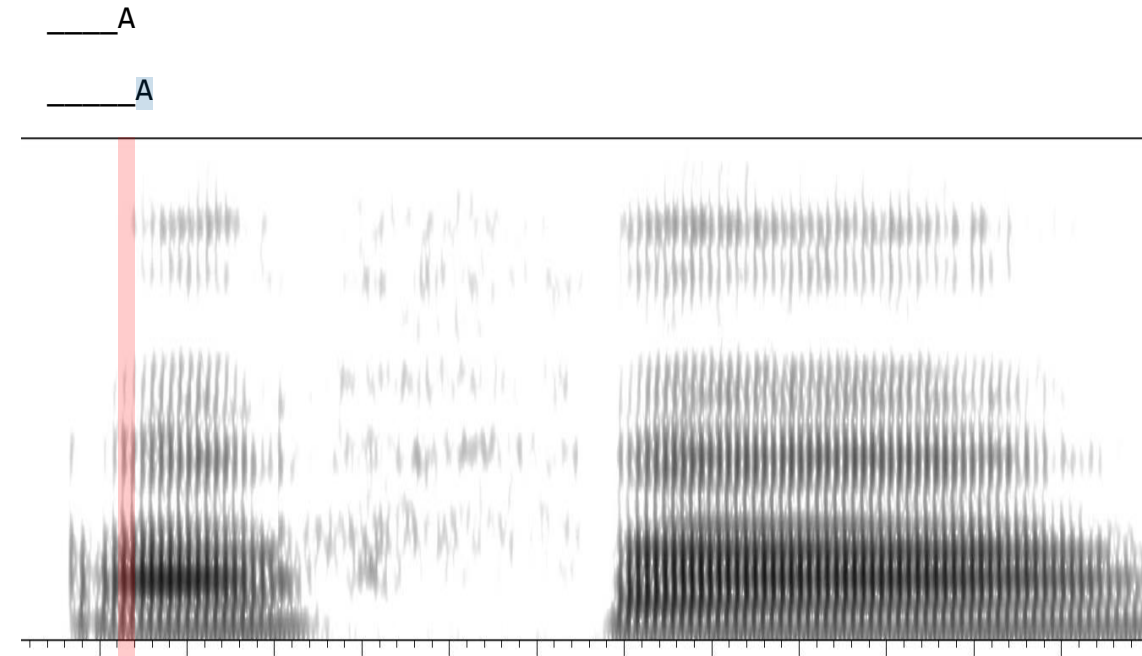
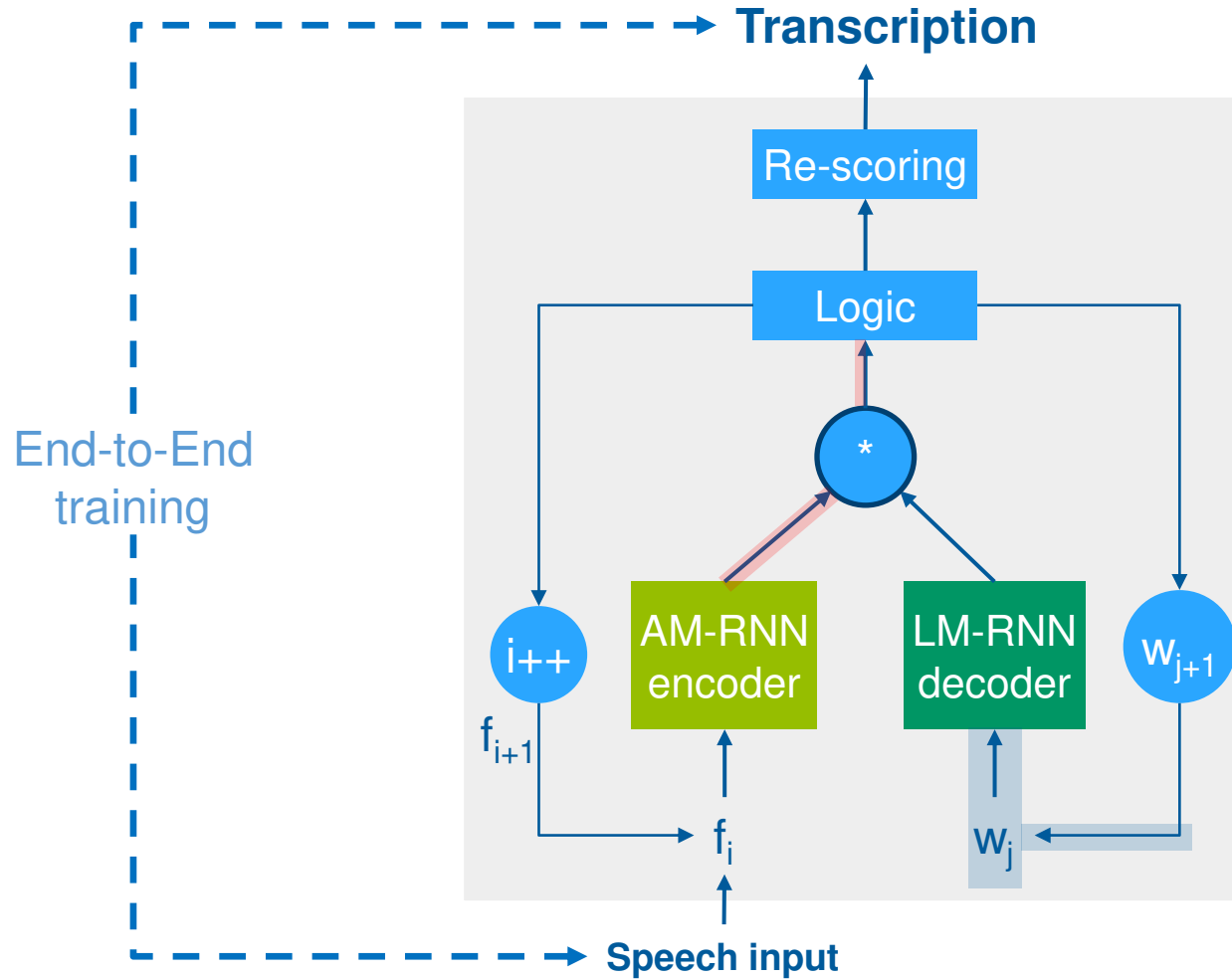
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



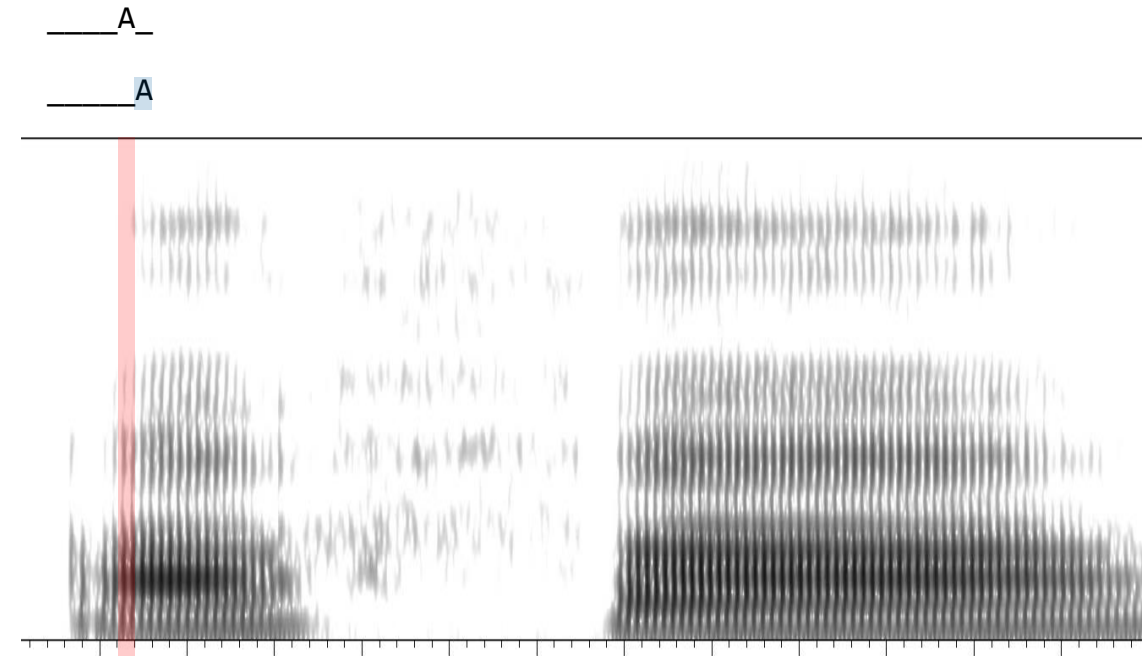
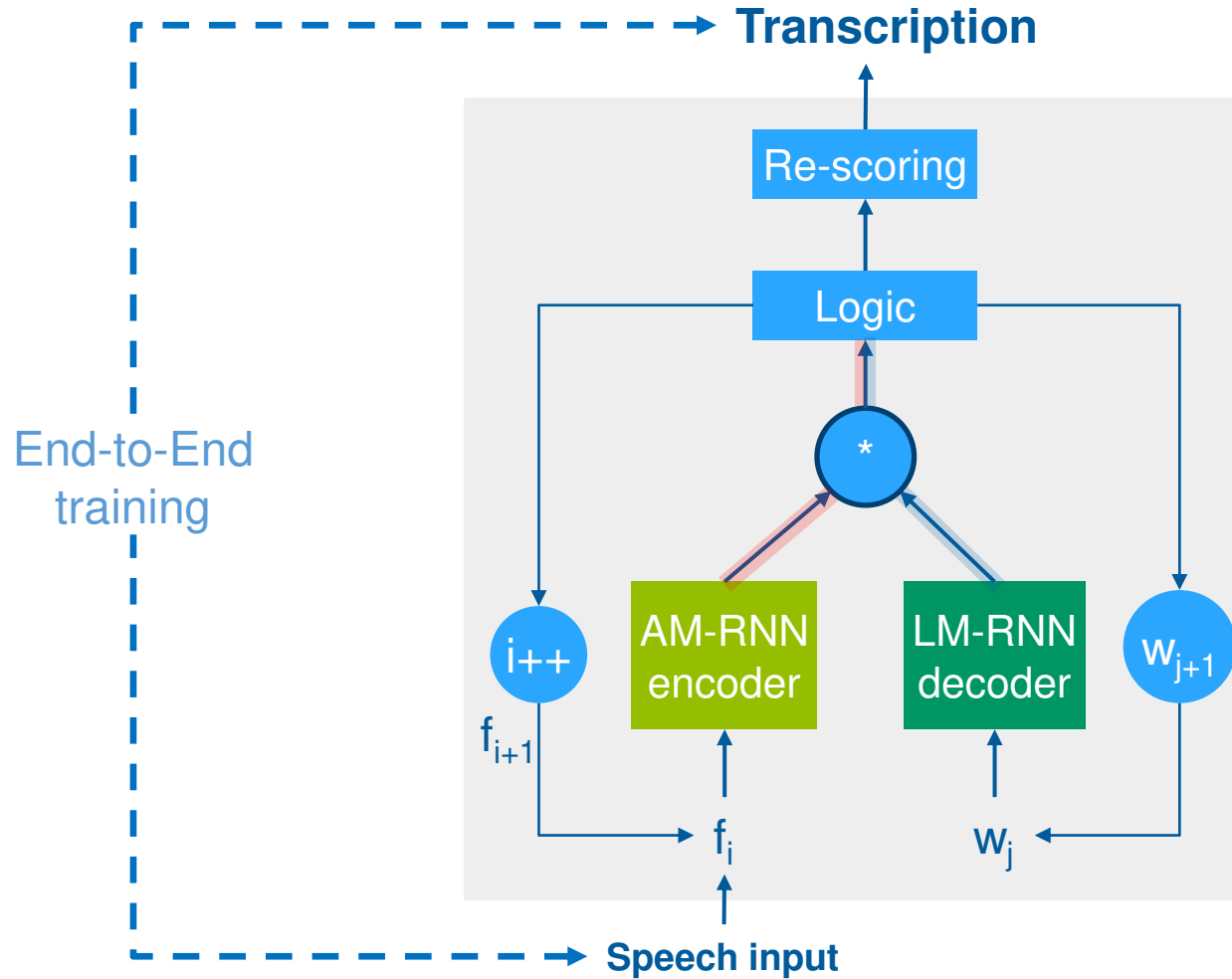
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



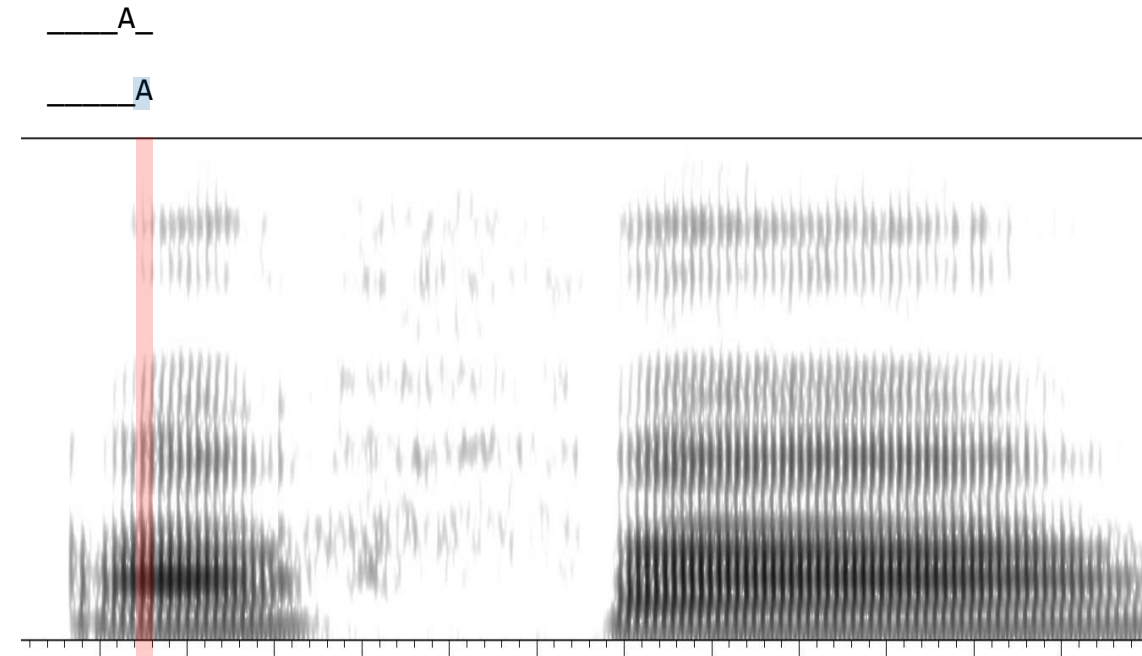
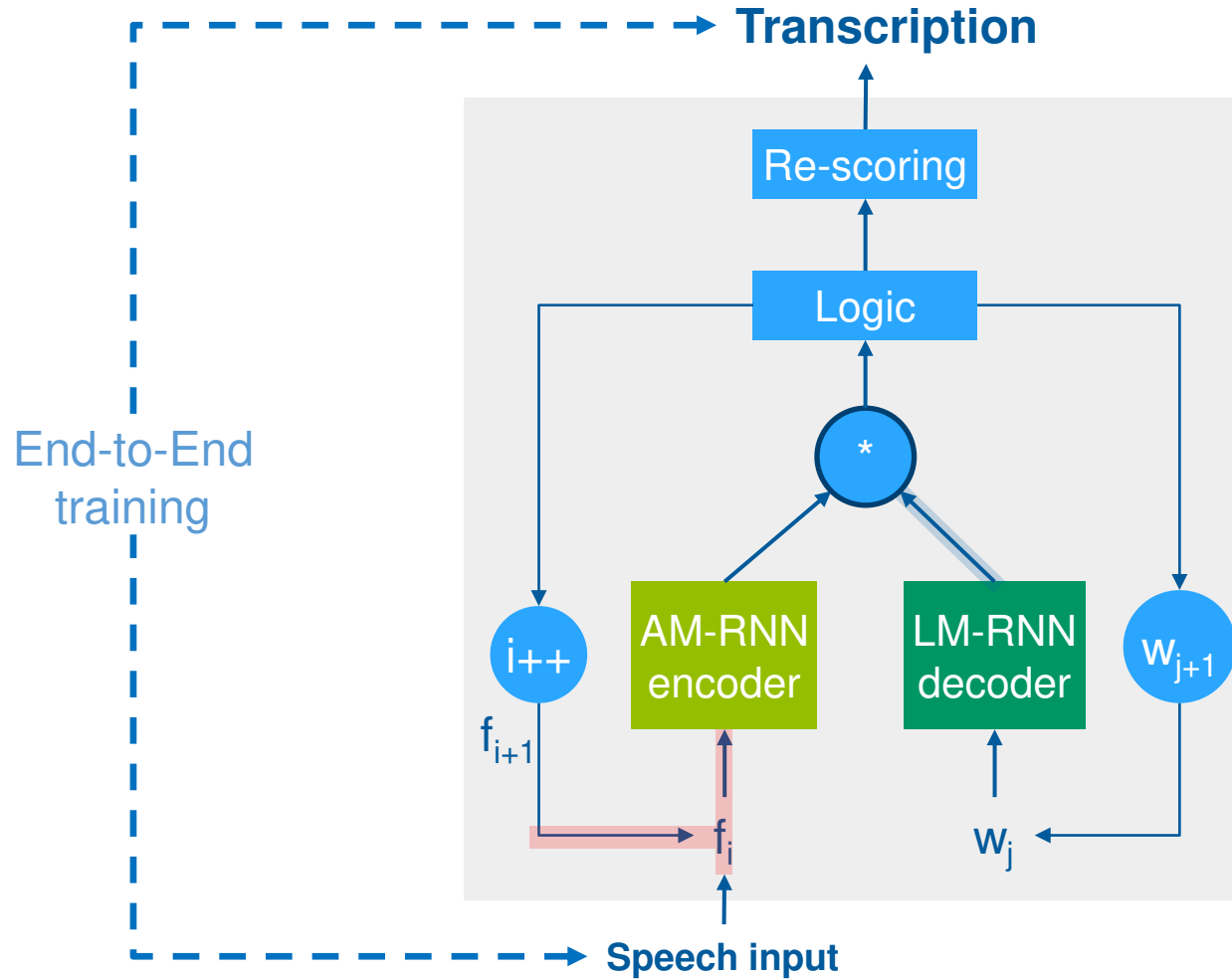
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



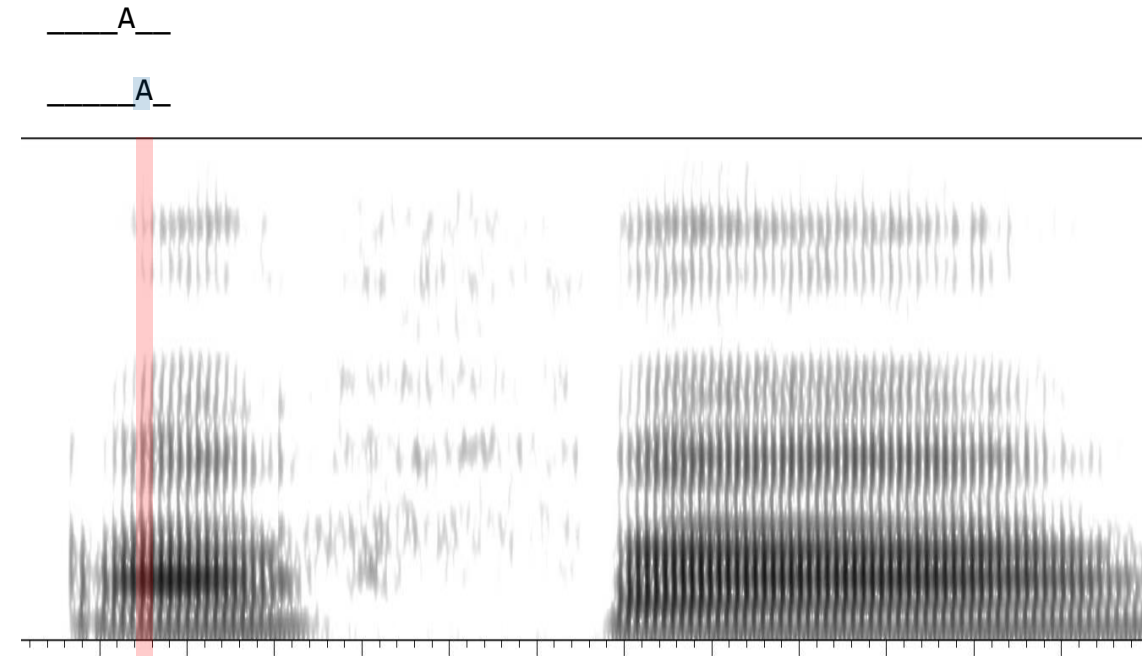
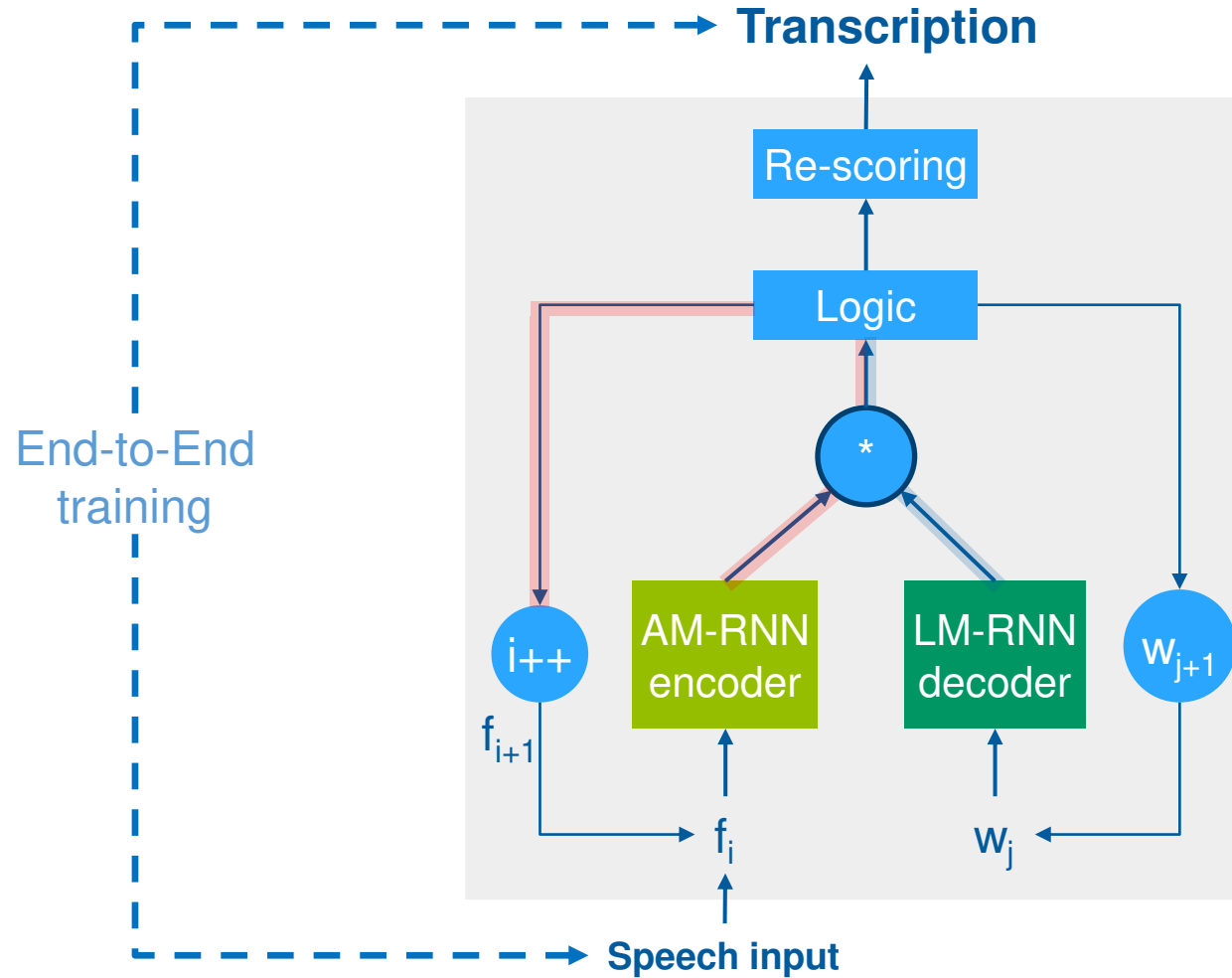
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



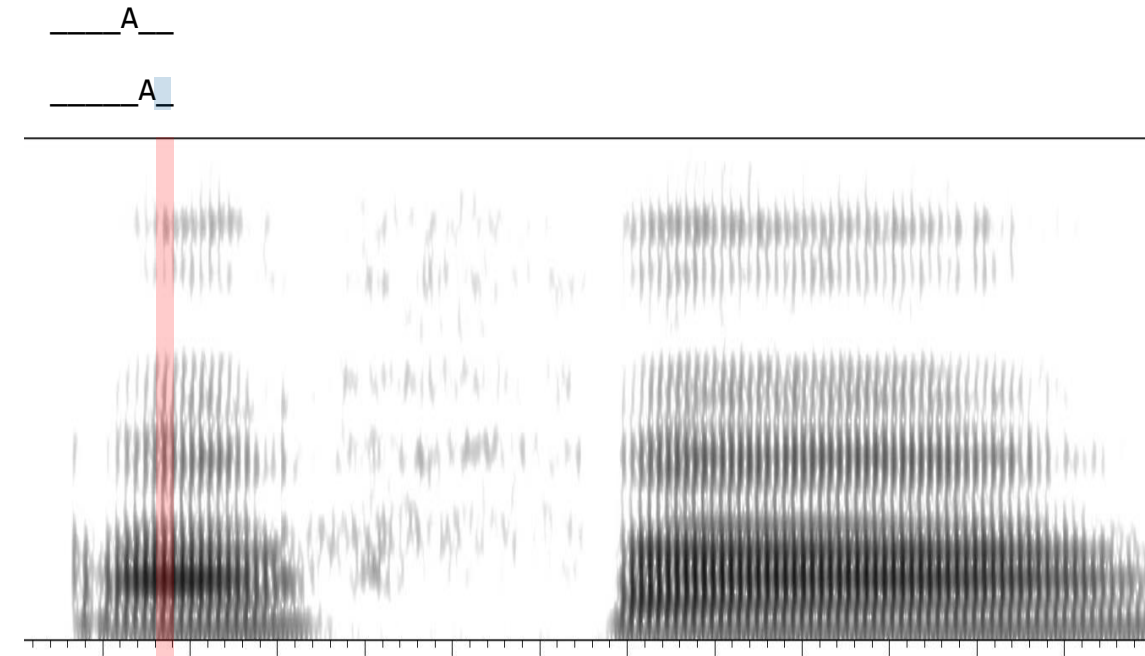
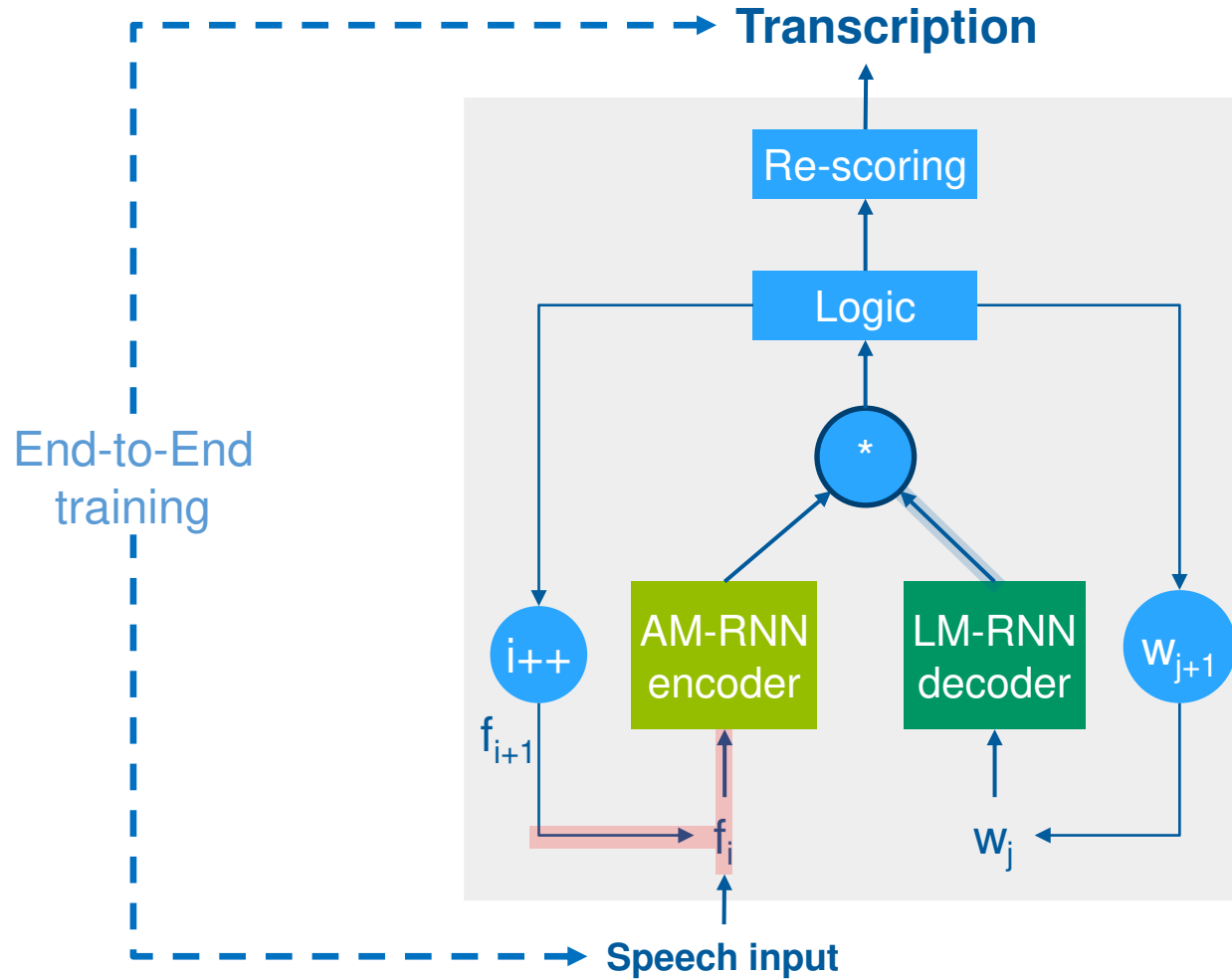
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



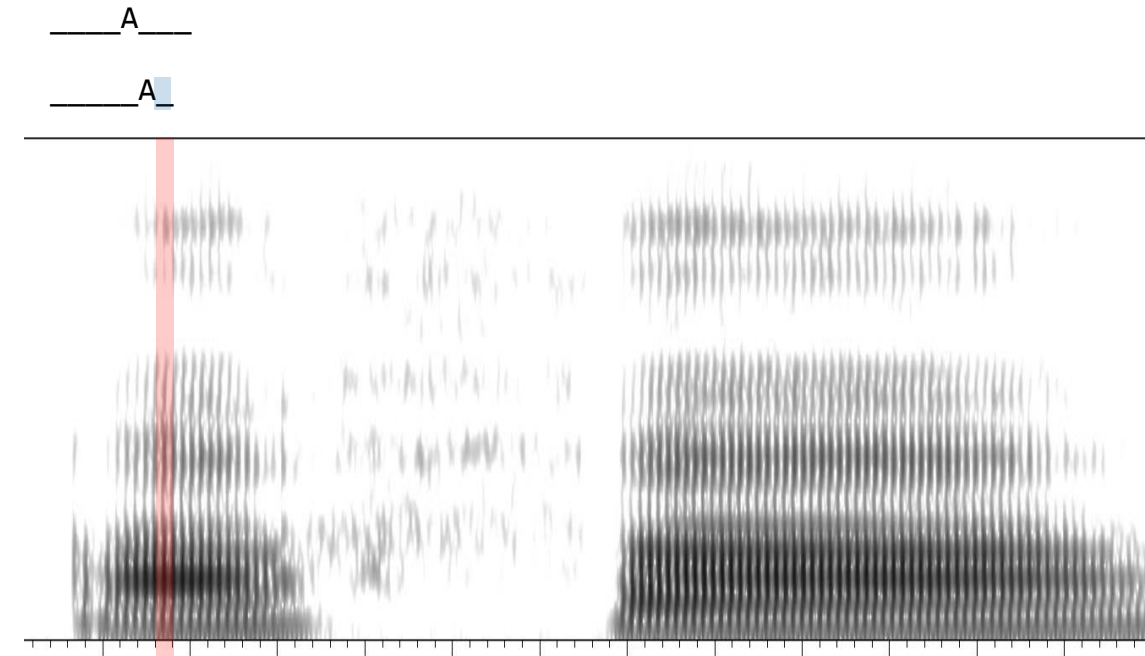
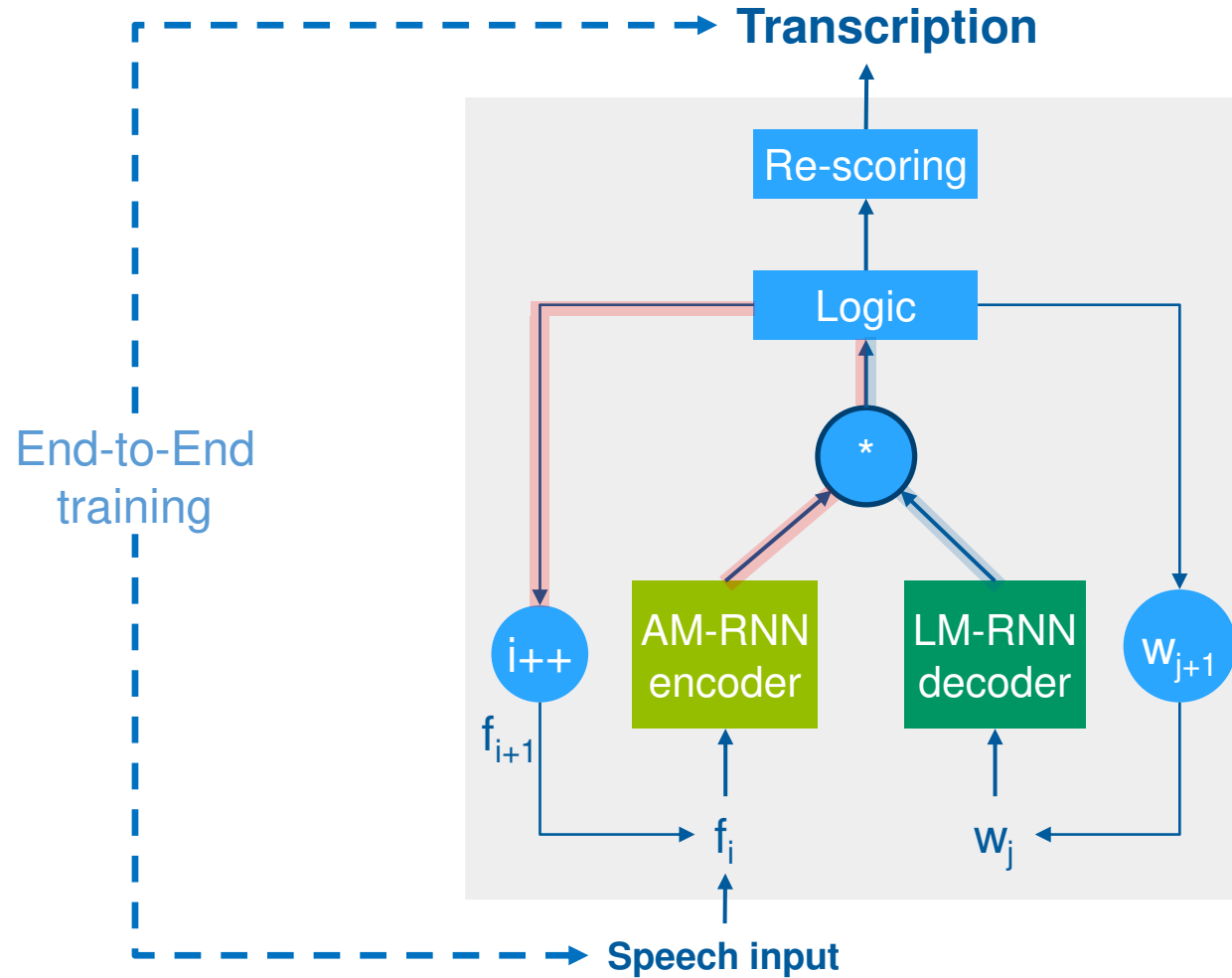
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



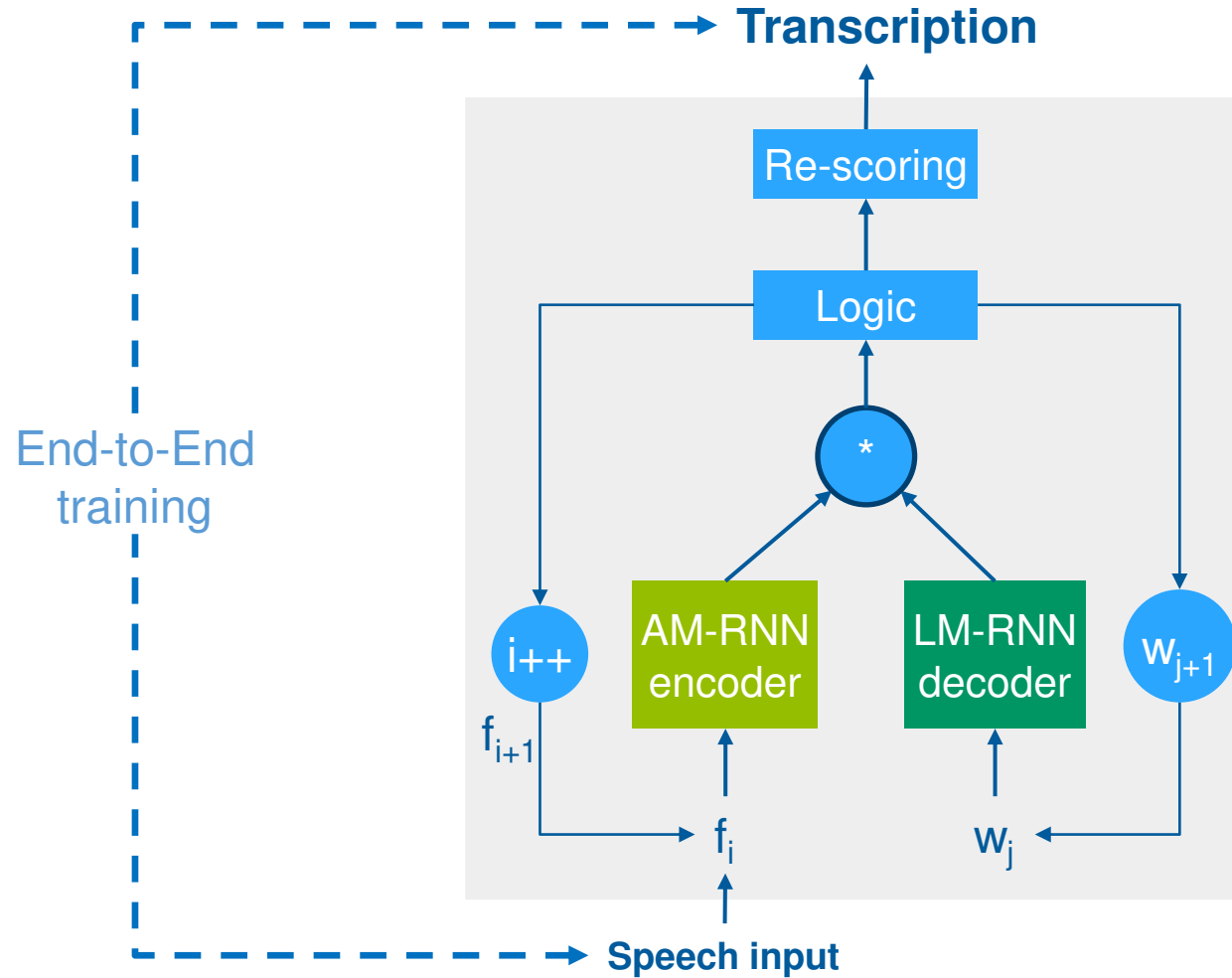
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



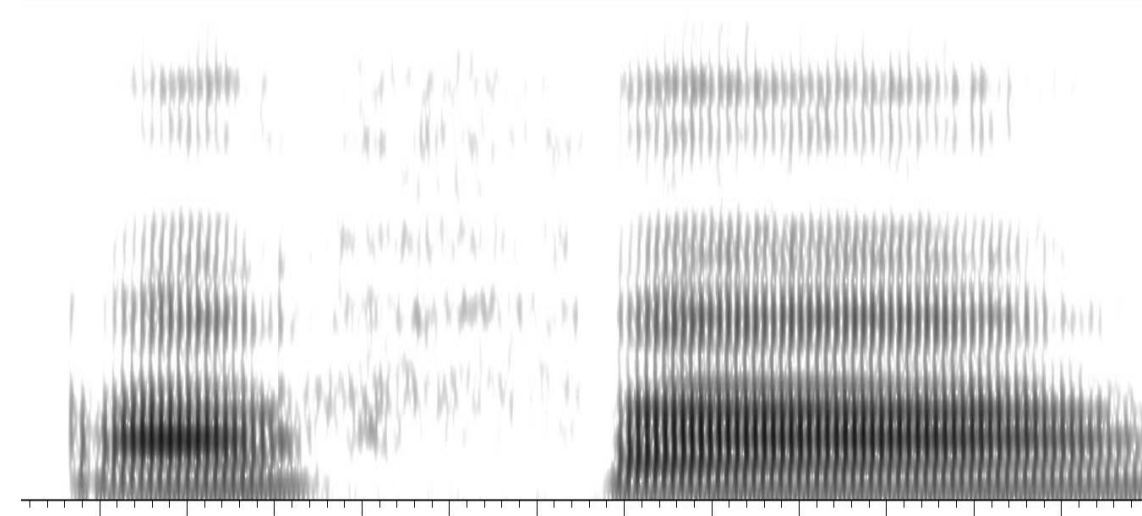
End-to-End Trained Automatic Speech Recognition

Recurrent Neural Network Transducer



_____A_____FFFF_____AAAA_____

_____A_____F_____A_____

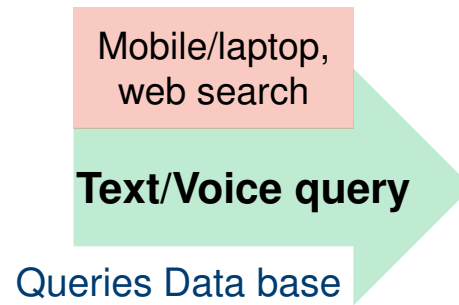


Spoken Language Understanding

Where we are, where we go.



Supports the user to find solutions



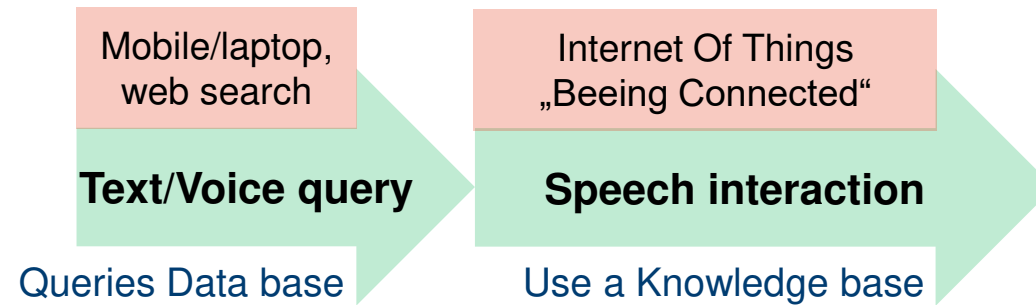
Spoken Language Understanding

Where we are, where we go.



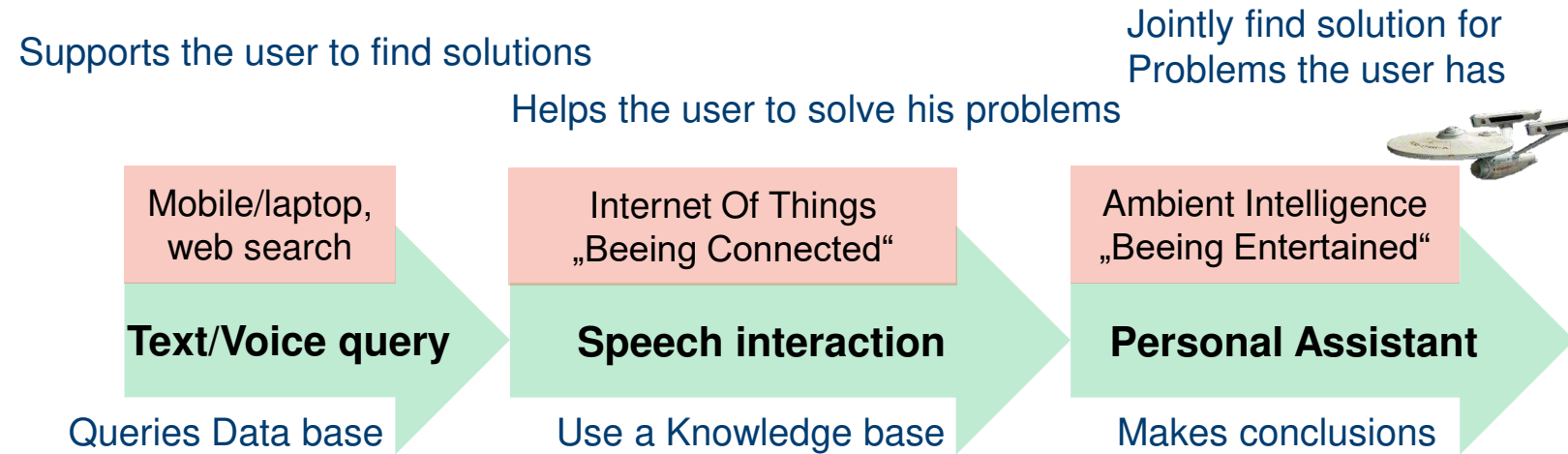
Supports the user to find solutions

Helps the user to solve his problems



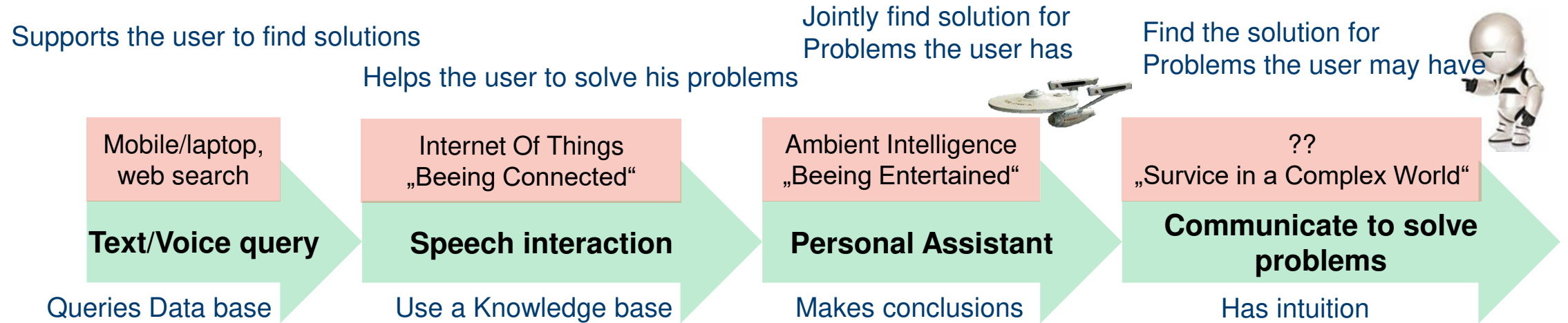
Spoken Language Understanding

Where we are, where we go.



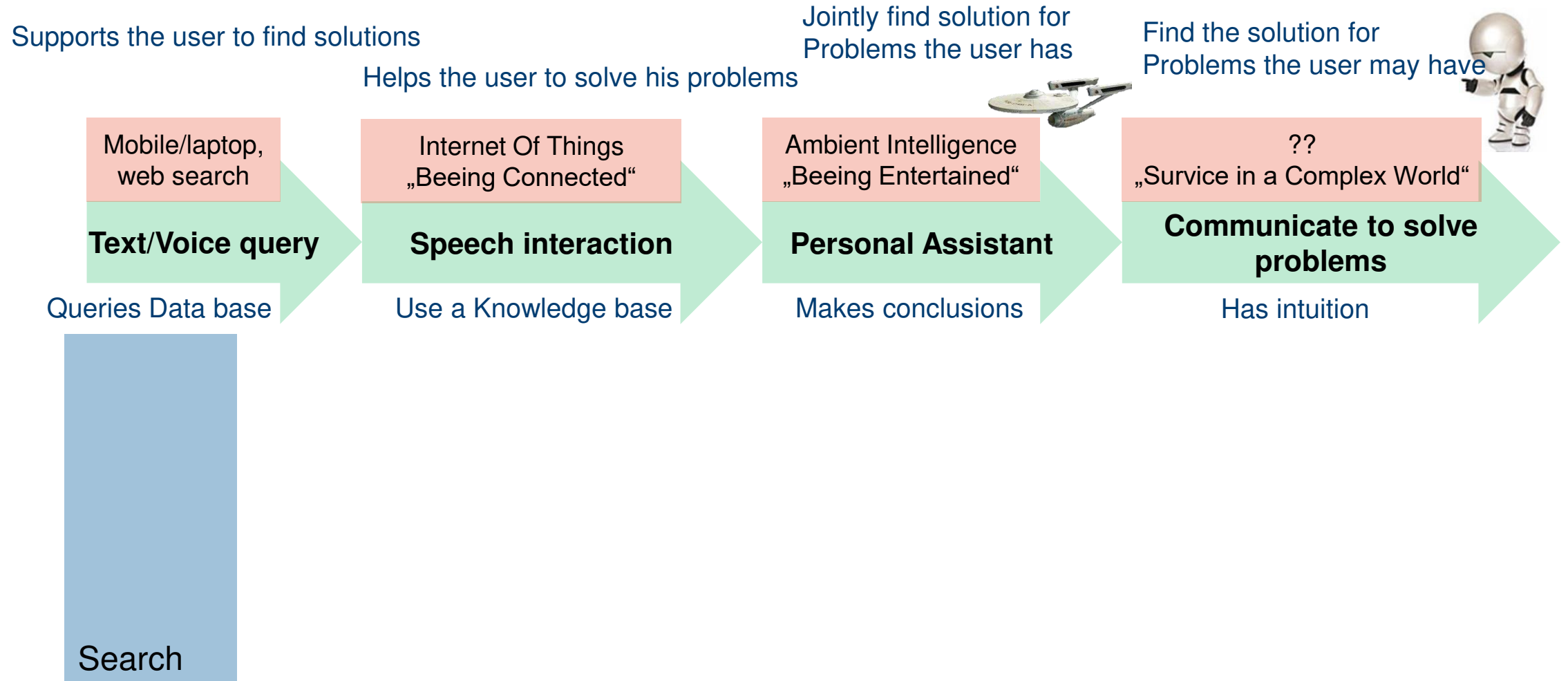
Spoken Language Understanding

Where we are, where we go.



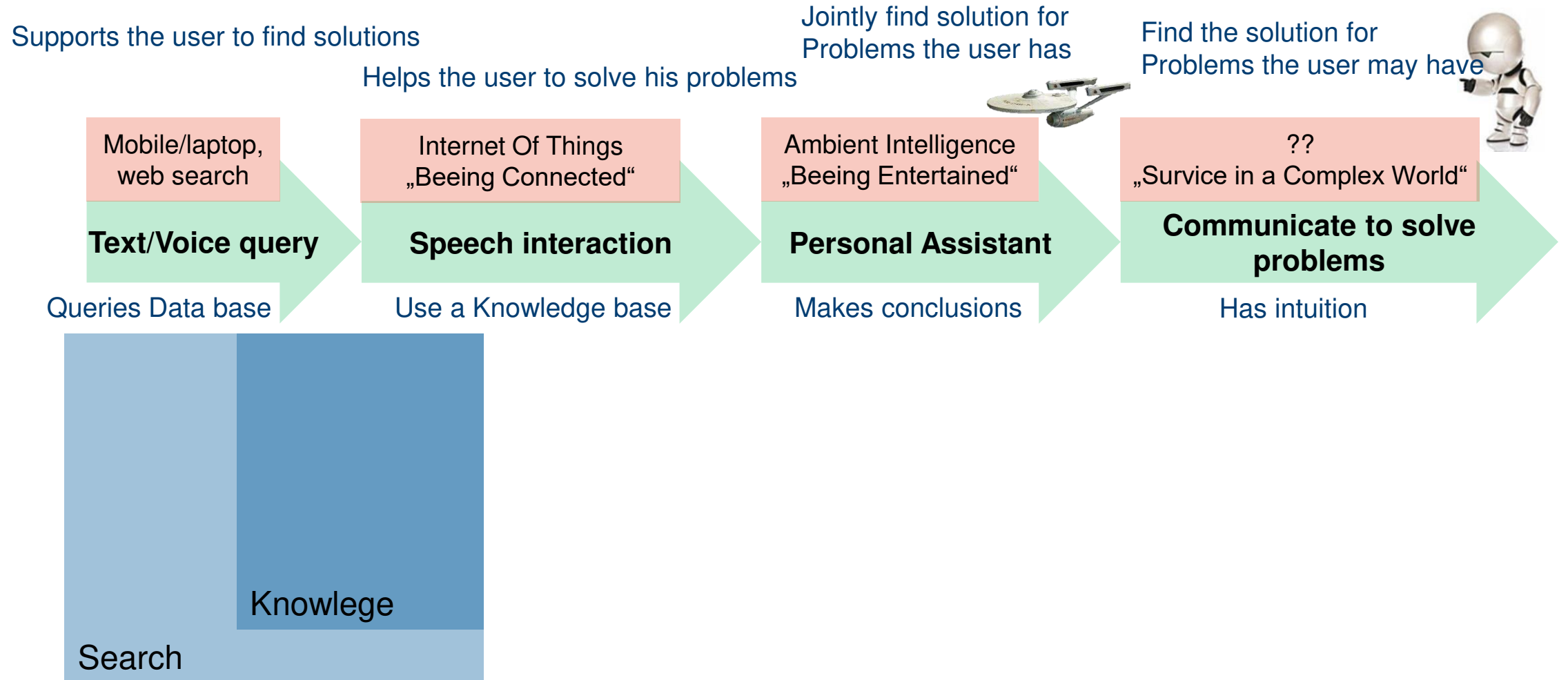
Spoken Language Understanding

Where we are, where we go.



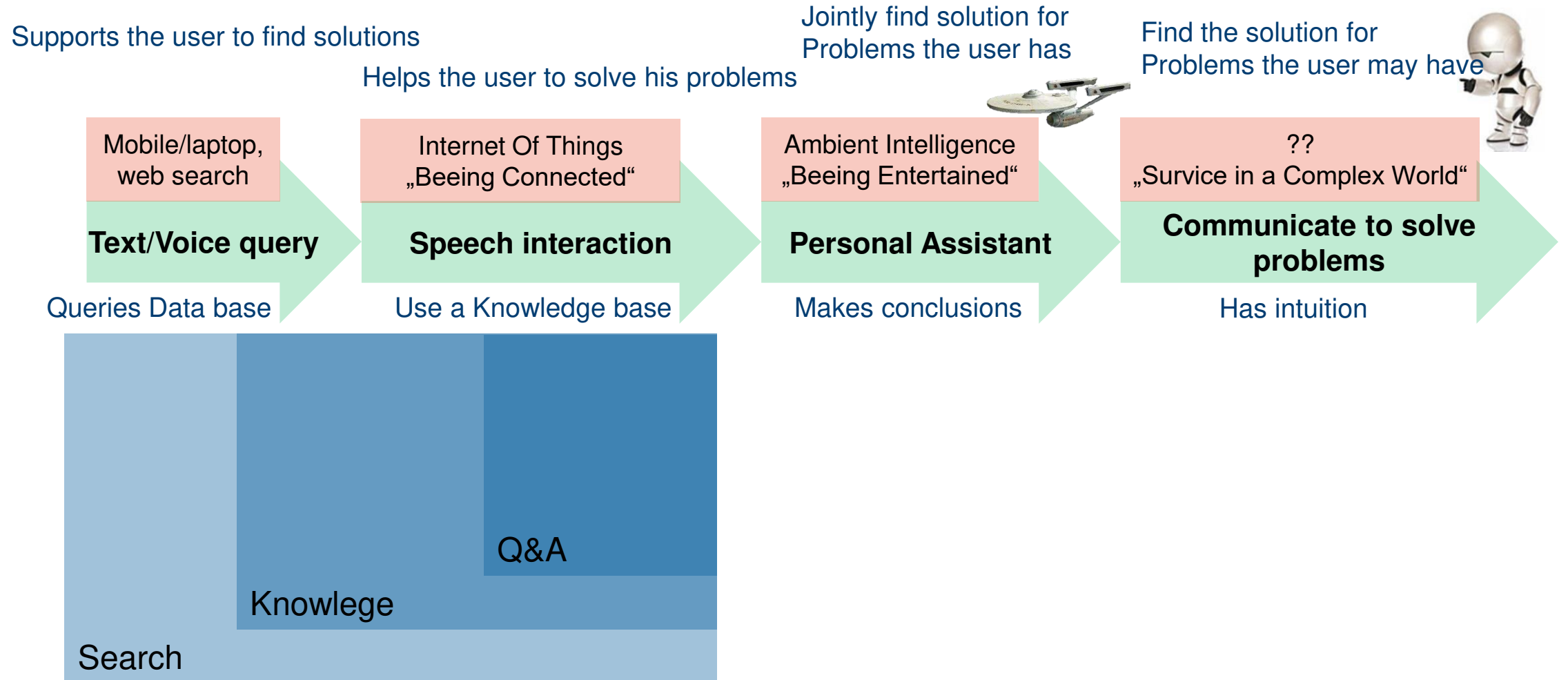
Spoken Language Understanding

Where we are, where we go.



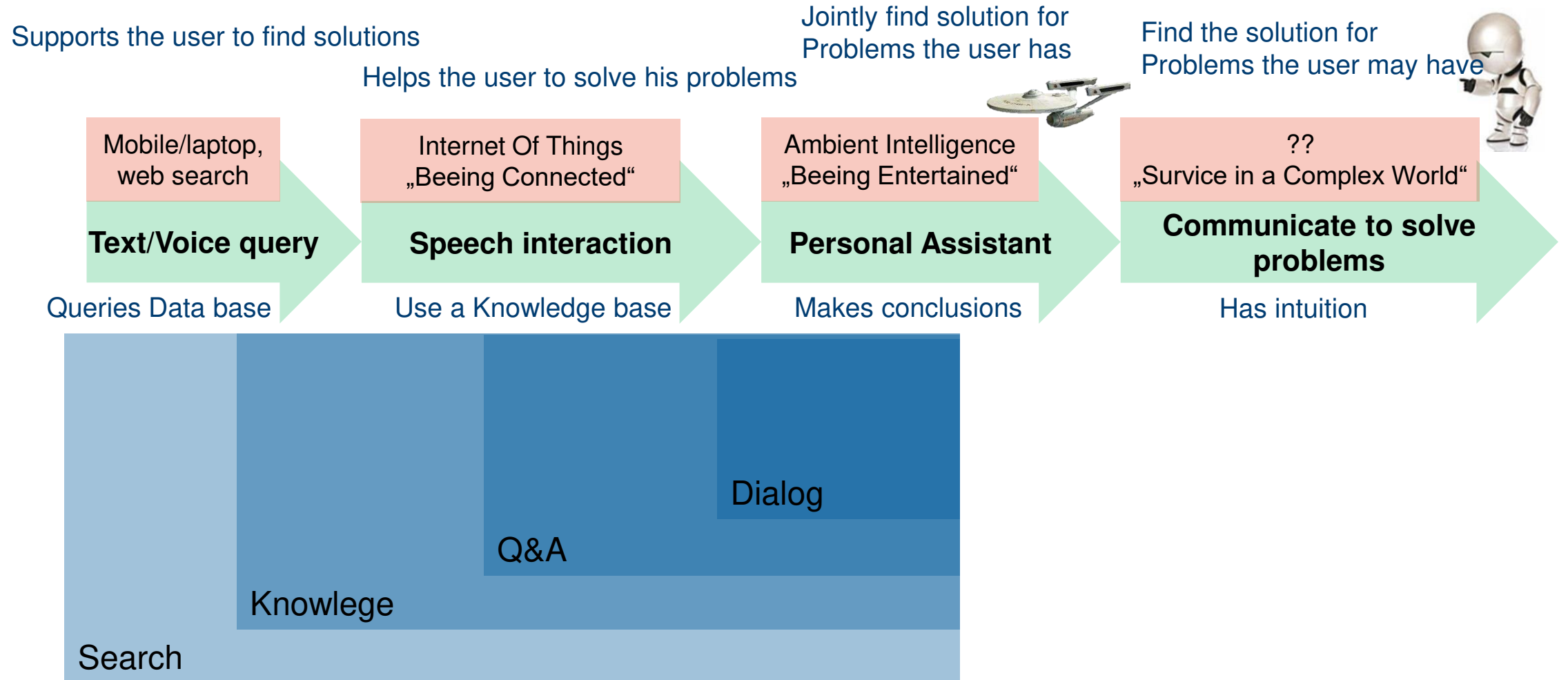
Spoken Language Understanding

Where we are, where we go.



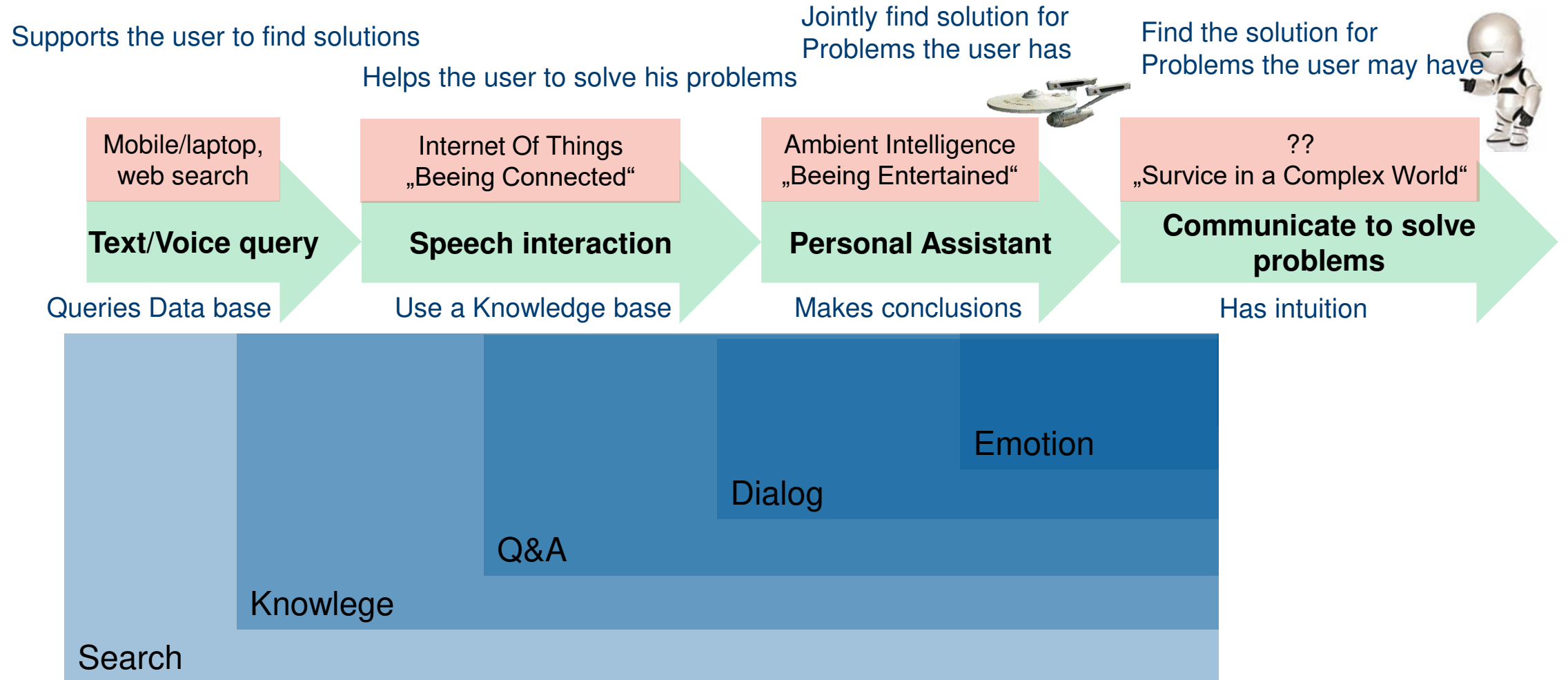
Spoken Language Understanding

Where we are, where we go.



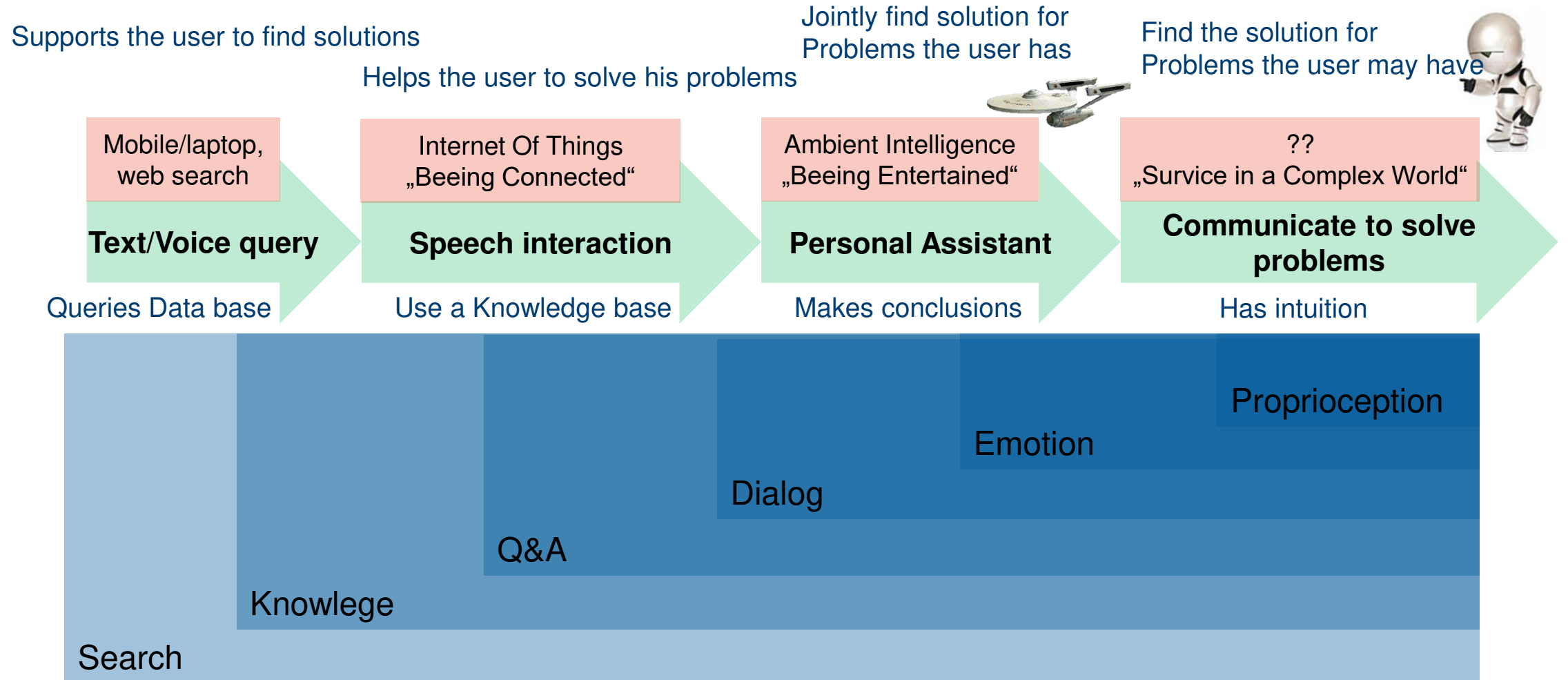
Spoken Language Understanding

Where we are, where we go.



Spoken Language Understanding

Where we are, where we go.



Spoken Language Understanding

Where we are, where we go.

