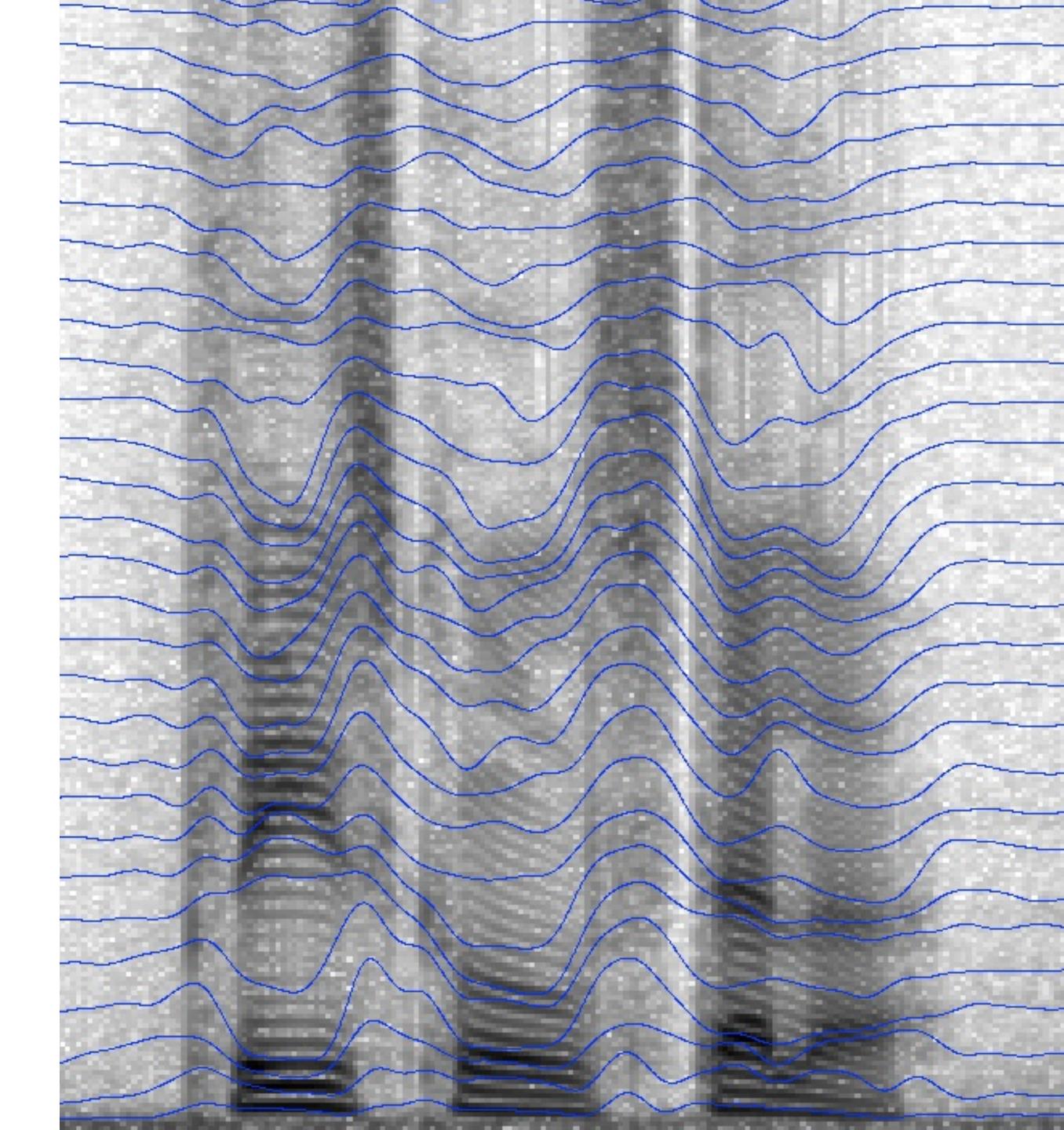
Speech Synthesis

Simon King & Korin Richmond University of Edinburgh



Introduction to the course (2024-25 version)

- learning outcomes
- delivery
- timetable
- course outline
- introduction to the coursework

Learning outcomes

- Understand the speech synthesis proce required to convert text to speech.
- Be familiar with the **different speech synthesis methods** currently used by speech synthesis systems and understand the advantages and disadvantages of each.
- Have a detailed understanding of the principles of unit selection speech synthesis, and the issues involved with choosing suitable candidate units to match a given target sequence.

• Understand the **speech synthesis process**, and be familiar with the processing steps

Learning outcomes (continued)

- selection voice.
- **Practical experience** of building a synthetic voice yourself.
- Be familiar with the different **speech coding** techniques that can be used for speech synthesis, and understand how these can be used to aid the joining of individual speech segments and how using different signal processing techniques to manipulate speech synthesis output affects the speech quality.
- Be in a position to discuss current issues in speech synthesis and see where speech synthesis research is heading in the future.

• Understand the design issues associated with **recording data** suitable for building a unit

- The website speech.zone contains almost everything you will need
 - video material, slides for the videos, reading lists, forums, calendar, coursework instructions, slides for classes
- You still need to use **Learn** for submitting your coursework
- We will also use **Learn** to send class announcements

• you must have an **account** on this site, so that you can post on the forums - make sure you can log in, and email **Simon.King@ed.ac.uk** if you have any trouble

- throughout the course.
- Simon also wants feedback on speech.zone
 - is it clearly organised?
 - is the website reliable and fast enough?
 - is it obvious what relates to this course, and what does not?
 - does everything work correctly on your device?

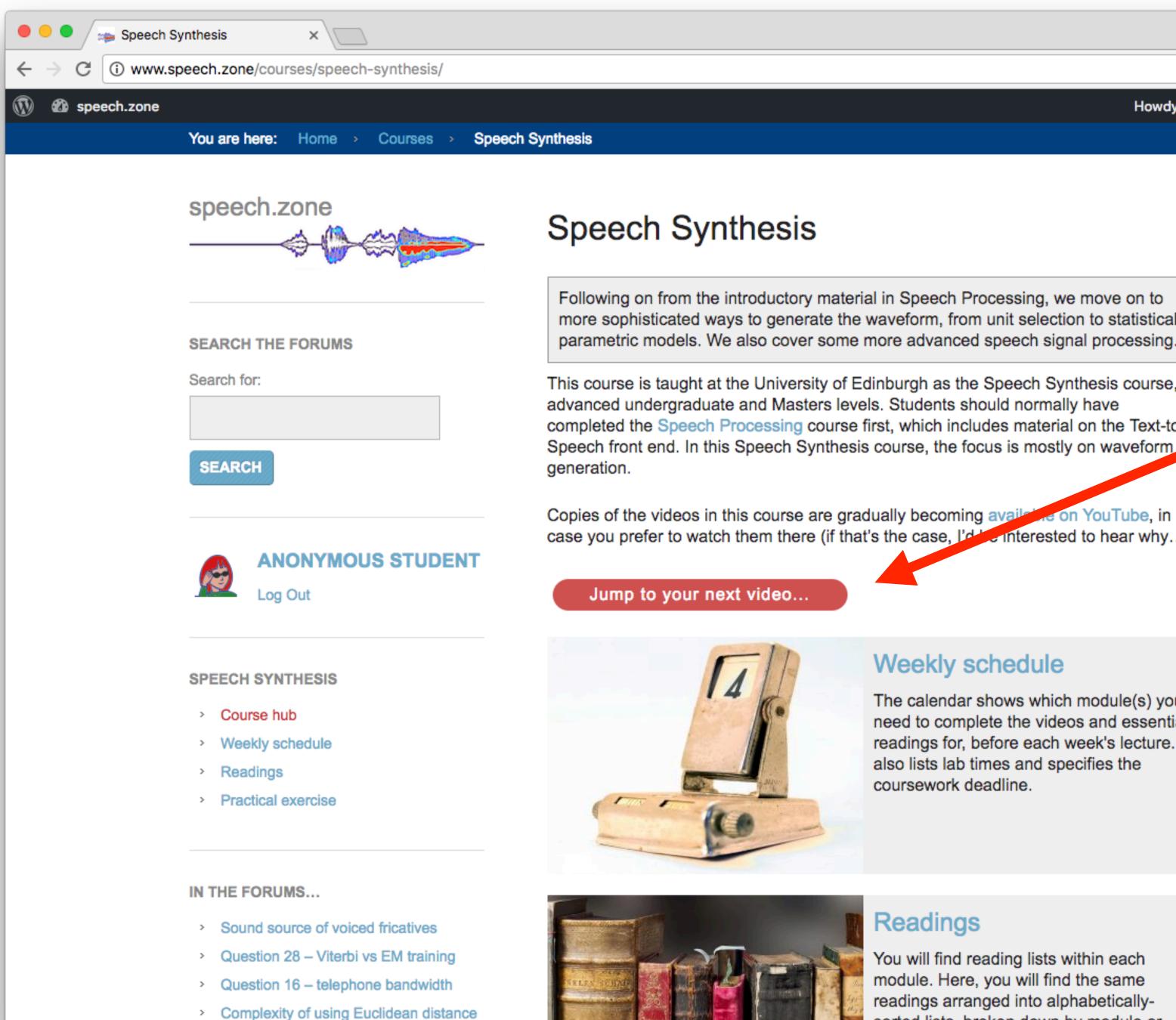
• Please give both of us **feedback** (email, forum posts, verbally, class reps, PPLS teaching) offices, notes slipped under office doors,...) about course structure and delivery mode,



- Lectures will cover the most popular current speech synthesis methods
 - unit selection
 - statistical parametric speech synthesis (SPSS) using HMMs or Neural Networks
 - the current state of the art: sequence-to-sequence models
- **Coursework** a single major assignment
 - build and evaluate a unit selection speech synthesiser, using your own recordings
- **Readings** lists provided on speech.zone
- Background assumed

• most of you will have taken Speech Processing - if you have not taken this course, then please speak to the lecturer as soon as possible (if Simon gives you permission to enrol, you'll need to catch up on Speech Processing content, including some videos and readings)

- The material on speech.zone is divided into **modules**
 - the video content provides only the *bare bones*
 - this is especially true of the more advanced material towards the end of the course • you need to flesh out the details by taking full advantage of
 - <u>readings</u>
 - active participation in <u>classes</u>
 - <u>labs</u> (including discussion with other students, the tutors, and the lecturer) forums (please attempt to answer each other's posts - I will correct any errors and
 - provide definitive answers)



Person 1
☆ 🎯 f? :
Howdy, Anonymous Student 🧟 🔍

Following on from the introductory material in Speech Processing, we move on to more sophisticated ways to generate the waveform, from unit selection to statistical parametric models. We also cover some more advanced speech signal processing.

This course is taught at the University of Edinburgh as the Speech Synthesis course, at completed the Speech Processing course first, which includes material on the Text-to-Speech front end. In this Speech Synthesis course, the focus is mostly on waveform

case you prefer to watch them there (if that's the case, l'd' a interested to hear why...)

Jump to your next video:

Requires you to rate videos, which marks them as completed.

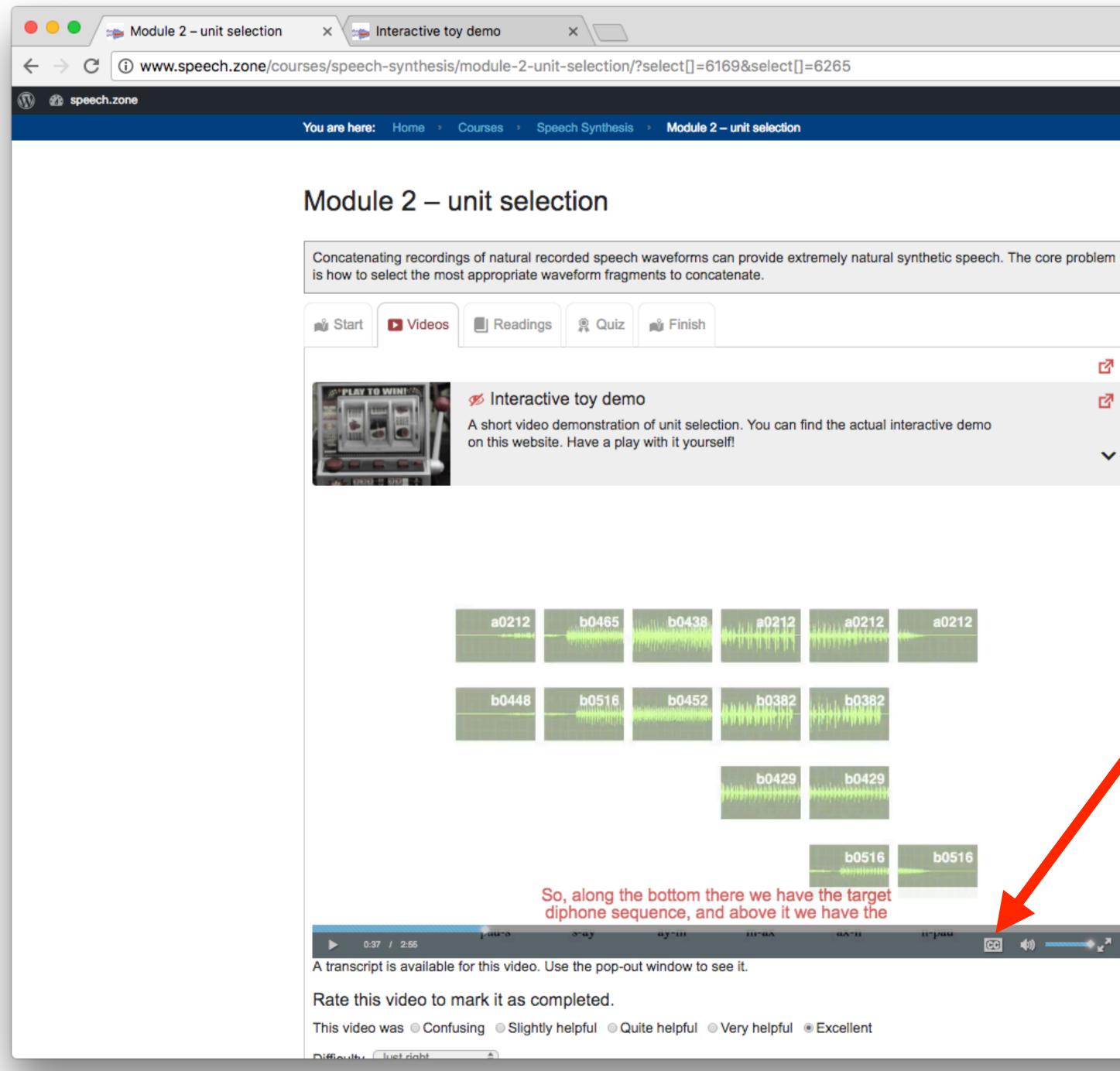
Weekly schedule

The calendar shows which module(s) you need to complete the videos and essential readings for, before each week's lecture. It also lists lab times and specifies the coursework deadline.

Readings

You will find reading lists within each module. Here, you will find the same readings arranged into alphabeticallysorted lists, broken down by module or





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Howdy, Anonymous Studen	t 🙍	Q,

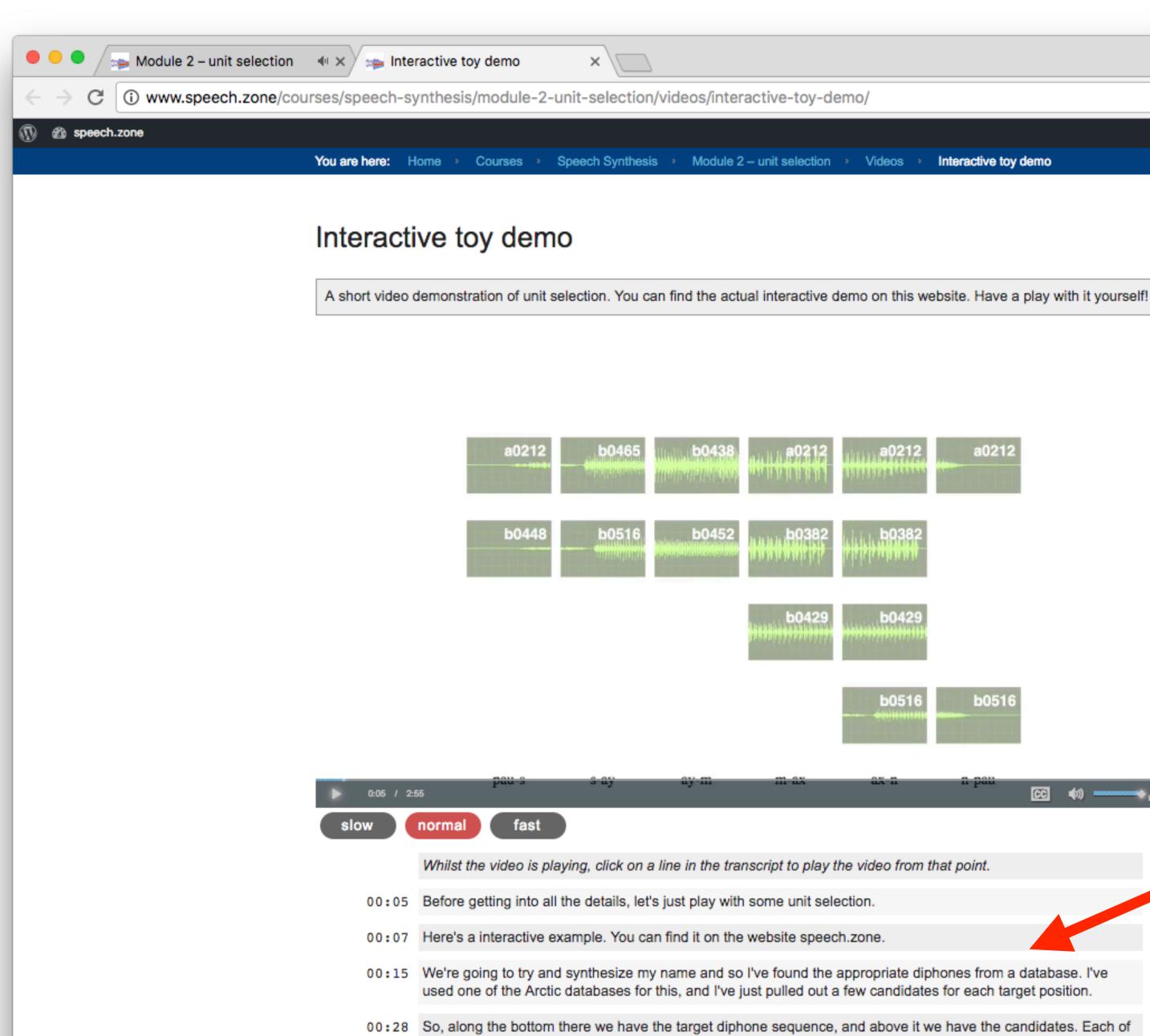
đ ₫ \mathbf{v} a0212 b0516

CC (1) ------ 2

Subtitles

are being rolled out gradually across the modules.





00:56 So, for example ... This is interactive, so if we select ... those will synthesize the waveform. This one this one all of those ones.

pick one candidate from each column.

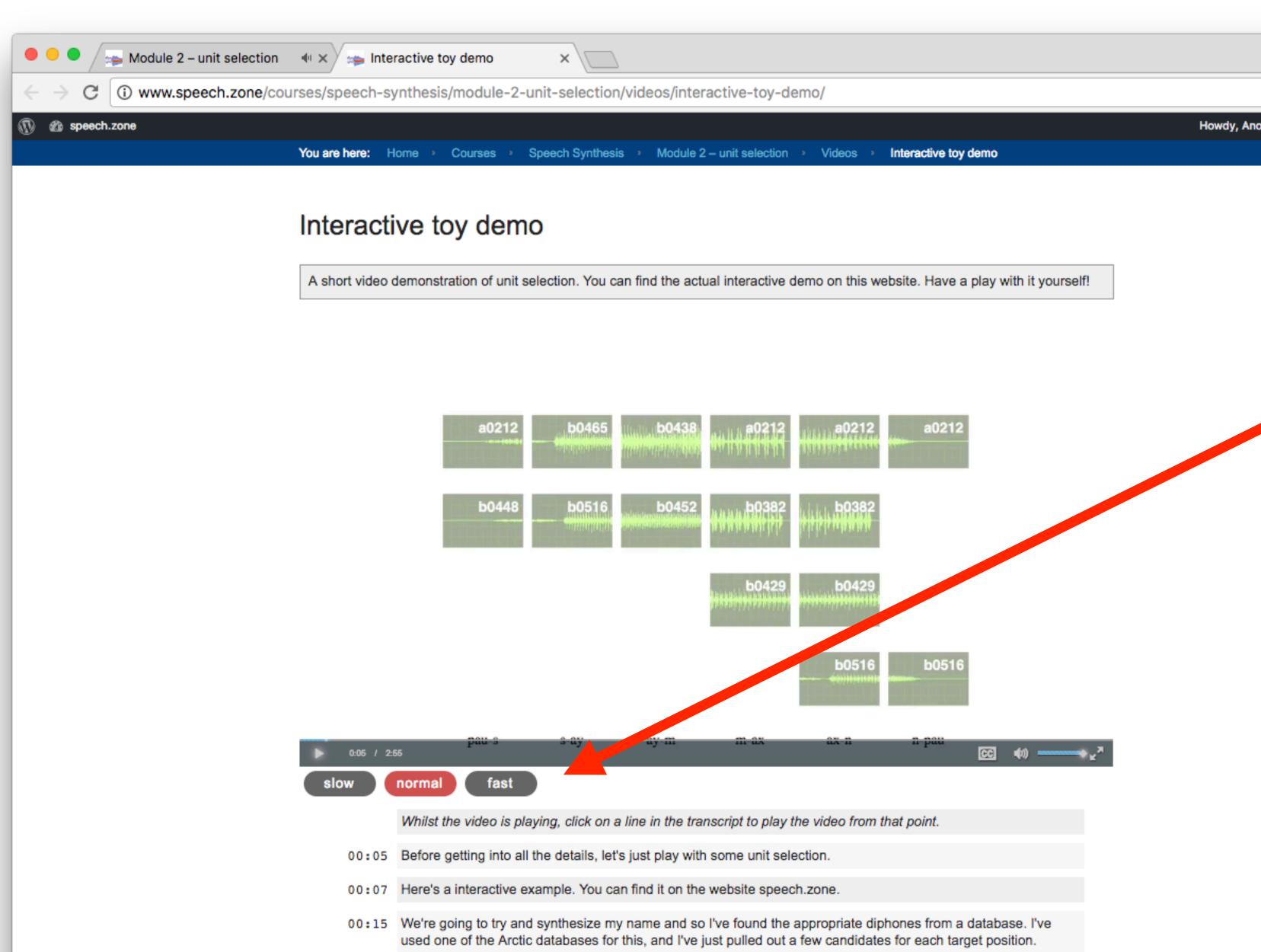
		Person 1
		Q ☆ 🎯 f? :
		Howdy, Anonymous Student 🧖 🔍
3	Interactive toy demo	

these candidates is just a little waveform fragment. So, we can listen to those and, to say my name, we need to

Transcripts

Open the video in the pop-out window. After starting playback, click on the transcript to jump to that point in the video.





00:28 So, along the bottom there we have the target diphone sequence, and above it we have the candidates. Each of these candidates is just a little waveform fragment. So, we can listen to those and, to say my name, we need to pick one candidate from each column.

00:56 So, for example ... This is interactive, so if we select ... those will synthesize the waveform. This one this one all of those ones.

	Person 1
	Q ☆ 🎯 f? :
	Howdy, Anonymous Student 🧖 🔍
Interactive toy demo	

Speed controls

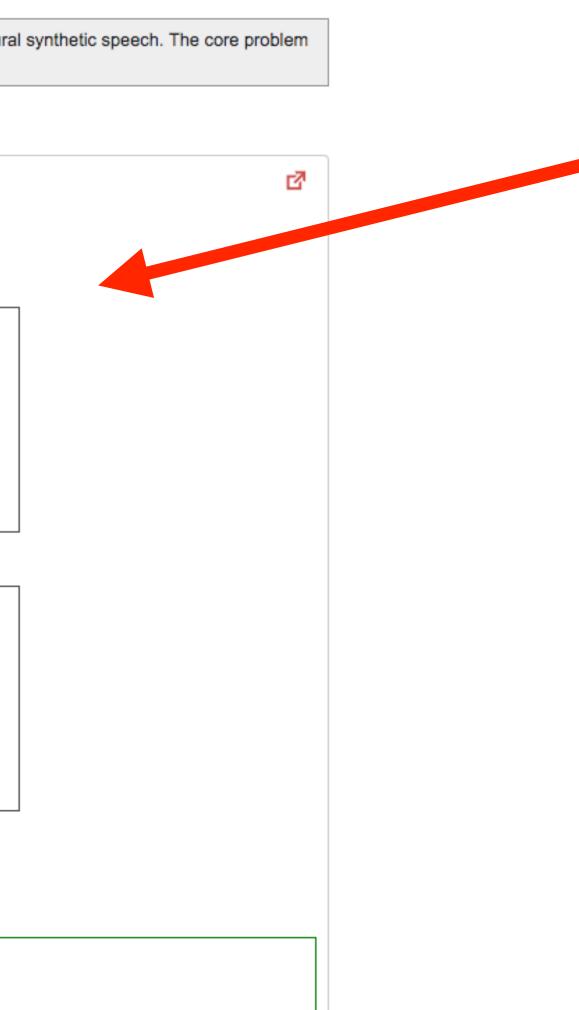
Chrome is recommended for best quality audio.

Please give feedback: are the speed settings right?

(You can also install a browser plugin to provide variable speed control for all videos on all websites.)



Module 2 - unit selection × Interactive toy demo ×	Person 1
$\leftarrow \rightarrow C$ (i) www.speech.zone/courses/speech-synthesis/module-2-unit-selection/?select[]=6169&select[]=6265	Q ☆ 🎯 f? :
M speech.zone You are here: Home > Courses > Speech Synthesis > Module 2 – unit selection	nonymous Student 📓 🔍
Module 2 – unit selection	
Concatenating recordings of natural recorded speech waveforms can provide extremely natural synthetic speech. The core problem is how to select the most appropriate waveform fragments to concatenate.	Flip
🔊 Start 🗈 Videos 🗐 Readings 😤 Quiz	
I'm going to be adding some quizzes – here's a teaser	Que
What does HMM stand for?	Ansv
	Click card
Would you like more of these?	
	If yo
There will also be short multiple-choice quizzes, like this:	
1. What does HMM stand for? It doesn't stand for anything	



card quizzes

estion on one side.

wer on the other.

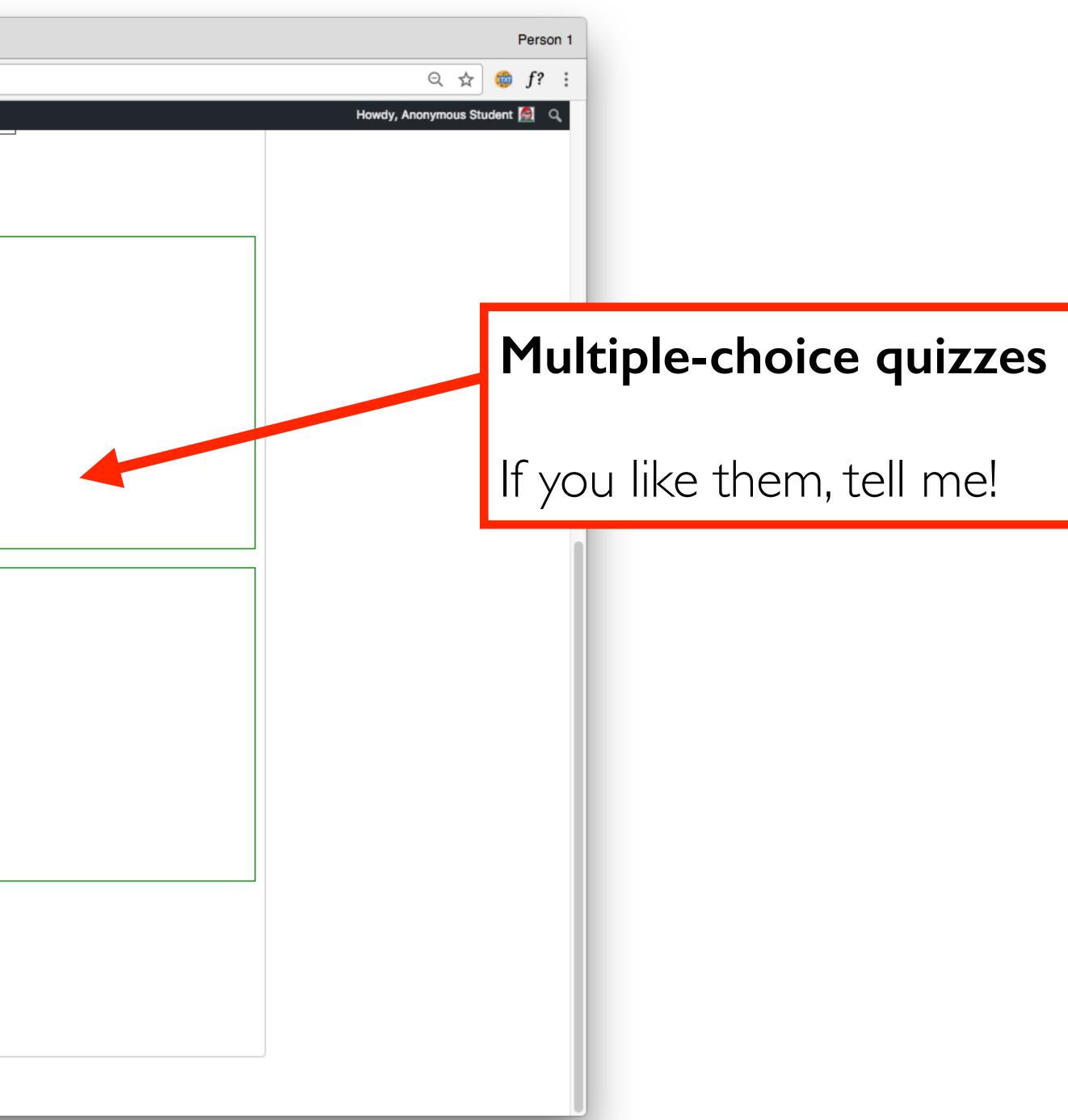
k anywhere on the to flip it over.

bu like them, tell me!



•••/=	Module 2 – unit selection	× Interactive toy demo	×
$\epsilon \rightarrow c$ () www.speech.zone/cours	ses/speech-synthesis/module-2-	unit-selection/?select[]=6169&select[]=6265

🕅 🍘 speech.zone	
	There will also be short multiple-choice quizzes, like this:
	1. What does HMM stand for? It doesn't stand for anything
	I have no idea
	Hidden Markov Model
	2. How old are you?
	21 (again)
	Old enough to know better
	Too old to remember
	Povoal all answers
	Reveal all answers
	Hide answers
	Show your score
	Clear your answers and try again

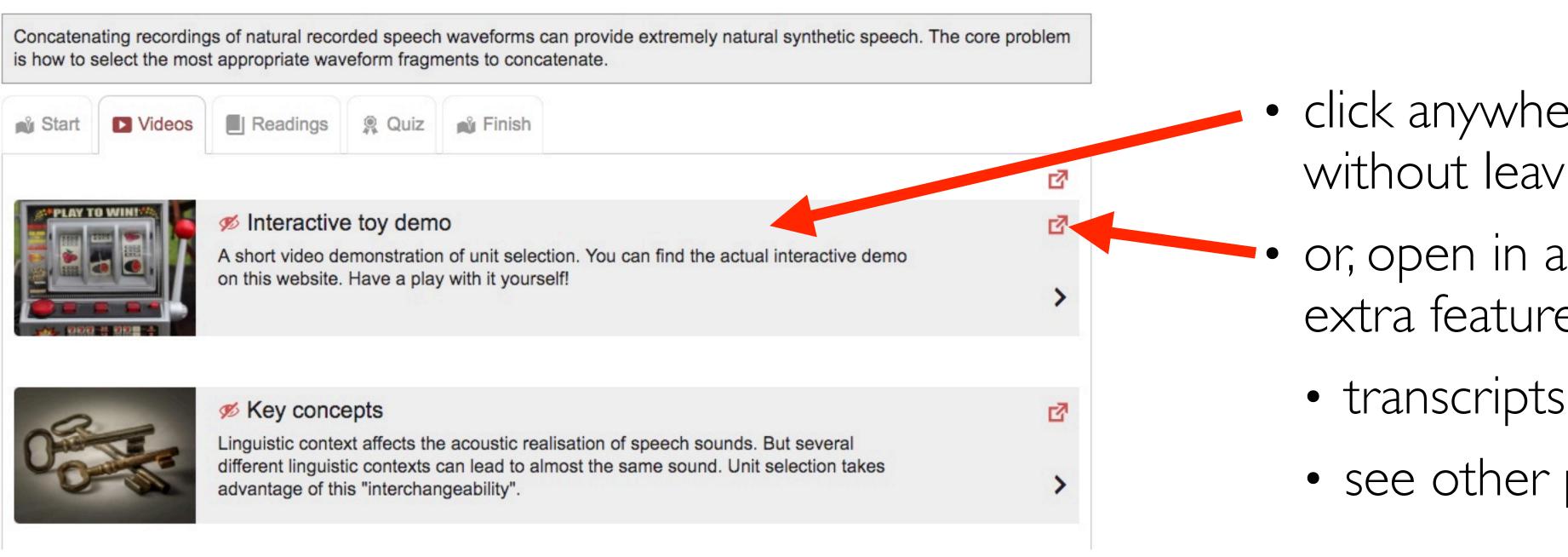




speech.zone tips

 check 'Remember me' to stay logged in for a year, in the current browser (otherwise, it's 2 days)

Module 2 – unit selection





Username: student	
Password:	
Remember Me Lost Password	Log In

- click anywhere to open videos without leaving this page
- or, open in a new window to get extra features
- see other people's ratings



What **you** have to do (*)

Before each class

- complete the module specified in the course calendar, including
 - the videos + all <u>Essential</u> readings
- post your questions on the forum

• In each class

- actively participate in discussion of the course content
- ask questions

(*) if you don't like this, then this course probably will not suit your learning style

Timetable

- Class
 - Tuesday 14:10 16:00
- Coursework
 - deadline is in the course calendar on speech.zone
- Exam (UG only)
 - during the April/May exam period ; date to be announced later

• Lab (you need to attend one session each week, but come to both if you fall behind) • group 1 : Wednesday 11:10 - 13:00 group 2 : Thursday 11:10 - 13:00 • additional booked lab time (talking & discussion encouraged!) : Friday 10:00 - 11:50

• examinable content = videos + Essential readings + class content for Module 8 onwards

Lecturers

Modules I to 5

Korin Richmond

Module 6 onwards

• Simon King

Marking policy

- Same marking policy as Speech Processing
 - https://www.speech.zone/courses/speech-processing/marking-policy
- Please read the **Common Marking Scheme** 60-69% = a good understanding of the video content and <u>Essential</u> readings 70-79% = as above, plus most <u>Recommended</u> readings 80% + = as above, plus independent study, including further readings of your choice

Coursework: build your own unit selection voice

- Supervised lab sessions start this week
 - attendance is a required
 - you will only do well on the assignment if you attend the lab every week
- Each lab session will be led by that week's lecturer, with a tutor
- There is an introduction to the coursework within this lecture, after a course outline

Course outline

Introduction

- taster, brief history lesson, understanding the problem, list of current issues
- Unit selection
 - the method, and how to construct the speech database it relies upon
- Signal processing
 - vocoding, estimating F0 from speech signals
- Statistical parametric speech synthesis
- The latest developments (the "state of the art")
 - from Deep Neural Networks to sequence-to-sequence models
 - open issues

• the method, and its advantages over unit selection ; from HMMs to Deep Neural Networks

Text-to-speech key challenges

- We can identify four main challenges for any builder of a TTS system.
 - **1.** Semiotic classification of text
 - 2. Decoding natural-language text
 - **3.** Creating natural, human-sounding speech
 - 4. Creating intelligible speech
- We can also identify two current and future main challenges
 - **1.** Generating affective and augmentative prosody
 - 2. Speaking in a way that takes the listener's situation and needs into account

(Taylor 2009, Section 3.6, page 51)

Semiotic classification of text

- This is what we called "text normalisation" in Speech Processing
- Commercial systems do pretty good job of this
- Festival is reasonably good
 - improvements would be straightforward, but take a lot of effort

• Largely a solved problem (or at least solvable with current methods, given enough effort)

Decoding natural-language text

- In Speech Processing, we covered aspects of this, including:
 - homographs
 - disambiguate using POS tags
 - will fail for homographs with the same POS but different senses
 - shallow ("syntactic") structure
 - phrase break prediction
- We can say that parts of this problem are solved
 - POS tagging, at least for well-resourced languages
- but that it's not entirely clear how much 'decoding' is needed for speech synthesis
 - which prevents people solving the remaining problems

Creating natural, human-sounding speech

- Much to discuss here, from
 - **low-level** signal quality
 - concatenating waveforms vs. using models & classical vocoders vs. neural vocoders
 - **segmental** quality
 - pronunciation, stress, connected speech processes
 - **augmentative** prosody (text-related)
 - very much an open and important problem even hard to define the scope!
 - **affective** prosody (not necessarily text-related)
 - some methods for generating 'affective' or 'emotional' speech, but few for predicting it (from what?)



Creating intelligible speech

- Closer to a solved problem than naturalness
- Can achieve human levels of intelligibility
 - straightforward with good statistical parametric systems (example 1.6.1)
- Unit selection systems
 - generally less intelligible than natural speech (example 1.6.2)
 - but this is in lab conditions with semantically-unpredictable sentences
- differences between systems cannot be measured, and may not matter

• interestingly, the most *natural-sounding* systems are **not** always the most *intelligible*

• In real applications, with 'normal' sentences, intelligibility is often at ceiling levels anyway, so

Understanding the problem

- Input is text
 - what properties of text do we need to know about?
- Output is speech
 - what properties of speech do we need to know about?
- How hard is the conversion from text to speech?
 - Do we need to understand the text?
 - If so, how would we do that?
 - If not, what do we need to extract from the text?

What properties of text do we need to know about?

"it is not necessary to go all the way and uncover the meaning from the written signal; we have to perform just the job of text decoding, not also that of text understanding

by and large, the identity and order of the words to be spoken is all we require to synthesise speech; no higher-order analysis or understanding is necessary." (Taylor 2009, Section 3.1.2, page 29)

but Taylor adds two caveats:

- word sense disambiguation (e.g., "polish", "lead", "bass")
- prosody (a huge caveat !!)



What properties of speech do we need to know about?

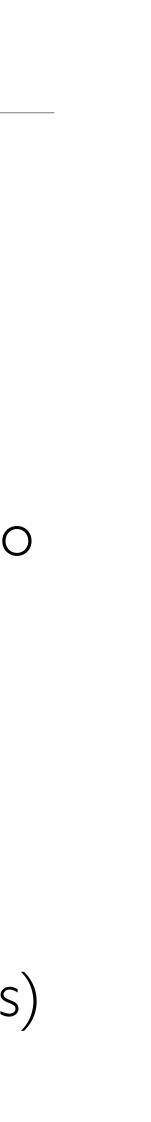
- about what speech is "made of", because
 - database)
 - the speech we have to recognise
- It is convenient to think about speech as a **linear** sequence of units
 - enables a concatenative approach to speech synthesis
 - to make models of larger units (e.g. words)

• To start us thinking about the issues involved in creating synthetic speech, let's think first

• in speech synthesis, we need to say **new** things (i.e., utterances not in our recorded

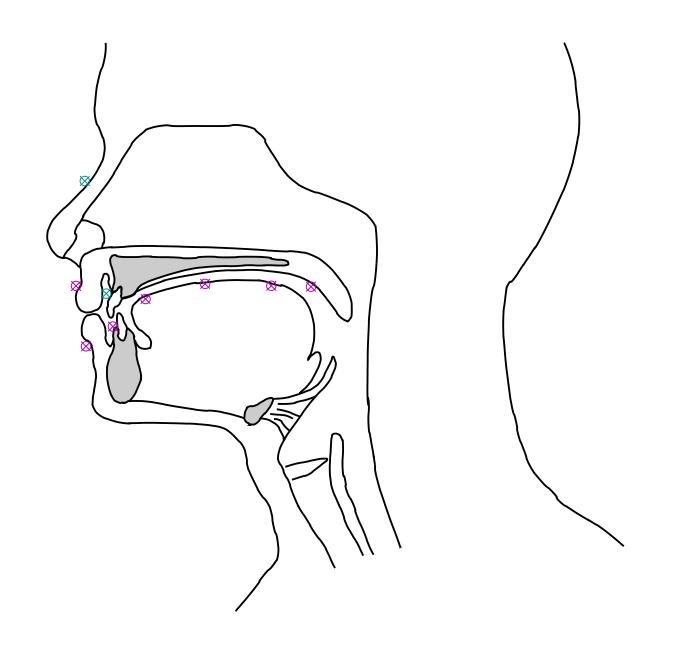
• in speech recognition, we need to generalise from the examples in the training data to

• in speech recognition, allows us to string together models of small units (e.g. phonemes)



Speech production

- Observed signal is result of several interacting processes
- The context in which a speech sound is produced affects that sound
 - articulatory constraints: where the articulators are coming from / going to
 - phonological effects
 - prosodic environment



Units of speech

- operating at different time scales
 - other aspects of the context in which it occurs
 - the context is complex it's not just the preceding/following sounds
- How can we reconcile this conflict, when we want to simultaneously:
 - model speech as a simple string of units
 - current moment in time

• The speech signal we observe (the waveform) is the product of interacting processes

• at any moment in time, the signal is affected not just by the current phoneme, but many

• take into account all the long-range effects of context, before, during and after the



Context is the key

- Context-dependent units offer a solution
 - engineer the system in terms of a simple linear string of units
 - then account for context by having a different version of each unit for every different context
- But, how do we know what all the different contexts are?
- If we enumerate all possible contexts, they will be practically infinite
 - there are an infinite number of different sentences in a language
 - context potentially spans the whole sentence (or further)
- However, what is important is the effect that the context has on the current speech sound - so next we can think about reducing the number of effectively different contexts

Current issues

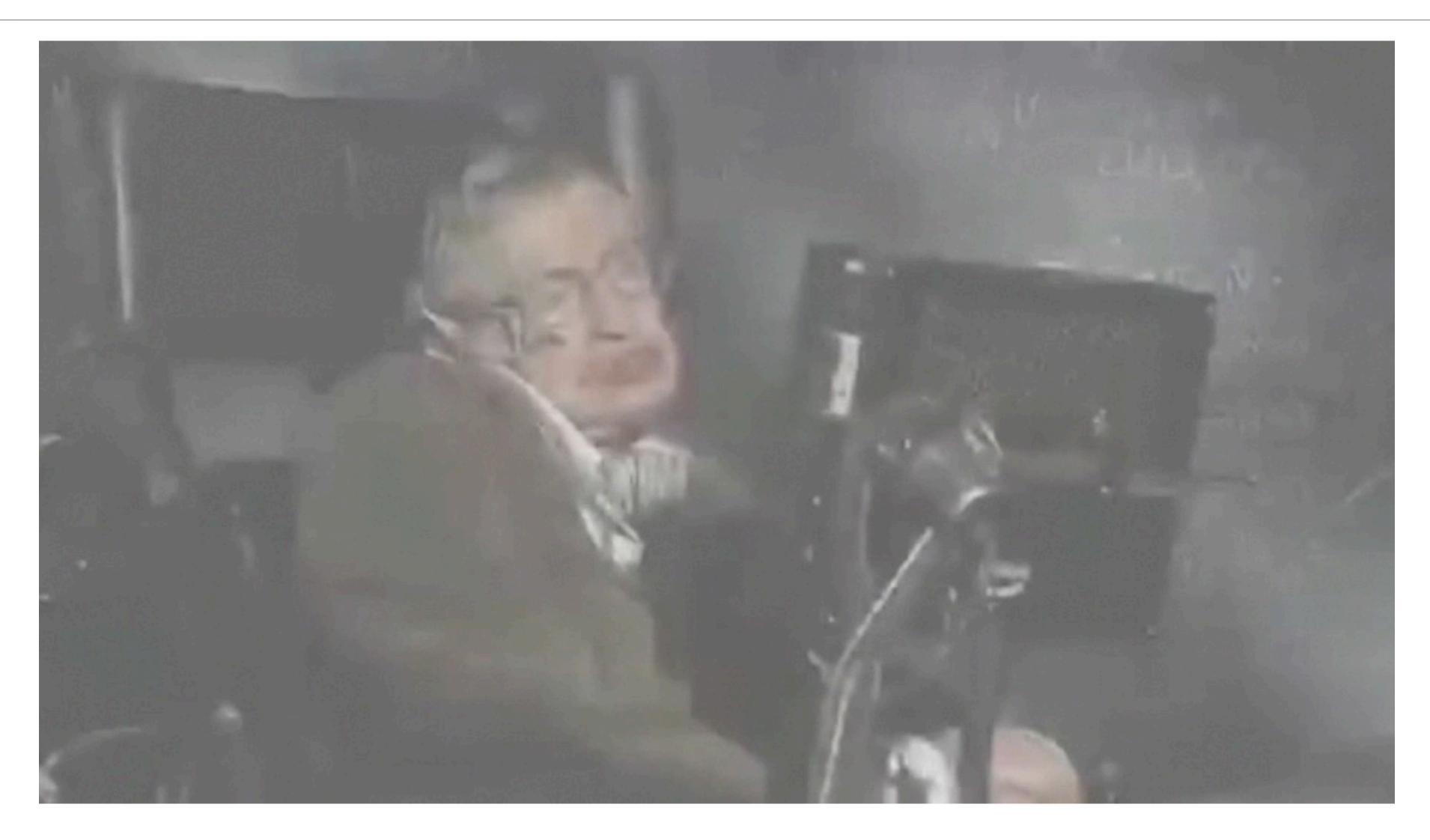
Deployed commercial systems

- heavily reliant on high-quality speech, professionally recorded in a studio
- multiple languages (but typically fewer than 50)
- speaking styles from a fixed set (e.g., newscaster, narrator)
- adaptation and control using markup, speech exemplars, Human-in-the-Loop

Recent and emerging techniques

- extensive use of found data' (necessitated by models that require a lot of data) • more powerful forms of control over speaking style
- rapid expansion to more languages

Assistive communication devices



Module I - introduction Class



Voices easy. Languages harder!



Module I - introduction Class

LibriVox acoustical liberation of books in the public domain

Listen

LibriVox provides free audiobooks from the public domain. There are several options for listening. The first step is to get the mp3 or ogg files into your own computer:

LibriVox Catalog

Podcast

Read

Would you like to record chapters of books in the public domain? It's easy to volunteer. All you need is a computer, some free recording software, and your own voice.

LibriVox volunteers record chapters of books in the public domain and publish the audio files on the Internet. Our goal is to record all the books in the public domain.

Module I - introductio Class



Volunteer

Visit the Forums



LibriVox: free audiobooks

LibriVox volunteers record chapters of books in the public domain and release the audio files back onto the net. Our goal is to make all public domain books available as free audio books.

- » More info
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LibriVox Links

- » Our catalog
- » How to listen
- » How to volunteer
- » Thank a reader
- » LibriVox forums
- » LibriVox wiki

Synthetic speech created from audiobooks



Class

Module I - introduction 1 paragraph example

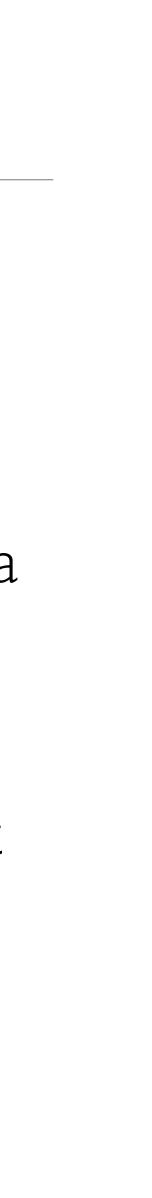
Current issues

What is being actively researched

- neural deep learning approaches mostly sequence-to-sequence models
- better and faster neural **vocoders**
- **prosody**, including its relationship to the meaning of the text (what Taylor calls "Generating affective and augmentative prosody")
- takes the listener's situation and needs into account.")
 - for impaired listeners
 - for speech-to-speech translation, including dubbing

• semi-, self-, and **un-supervised learning**, to reduce reliance on expensive labelled data

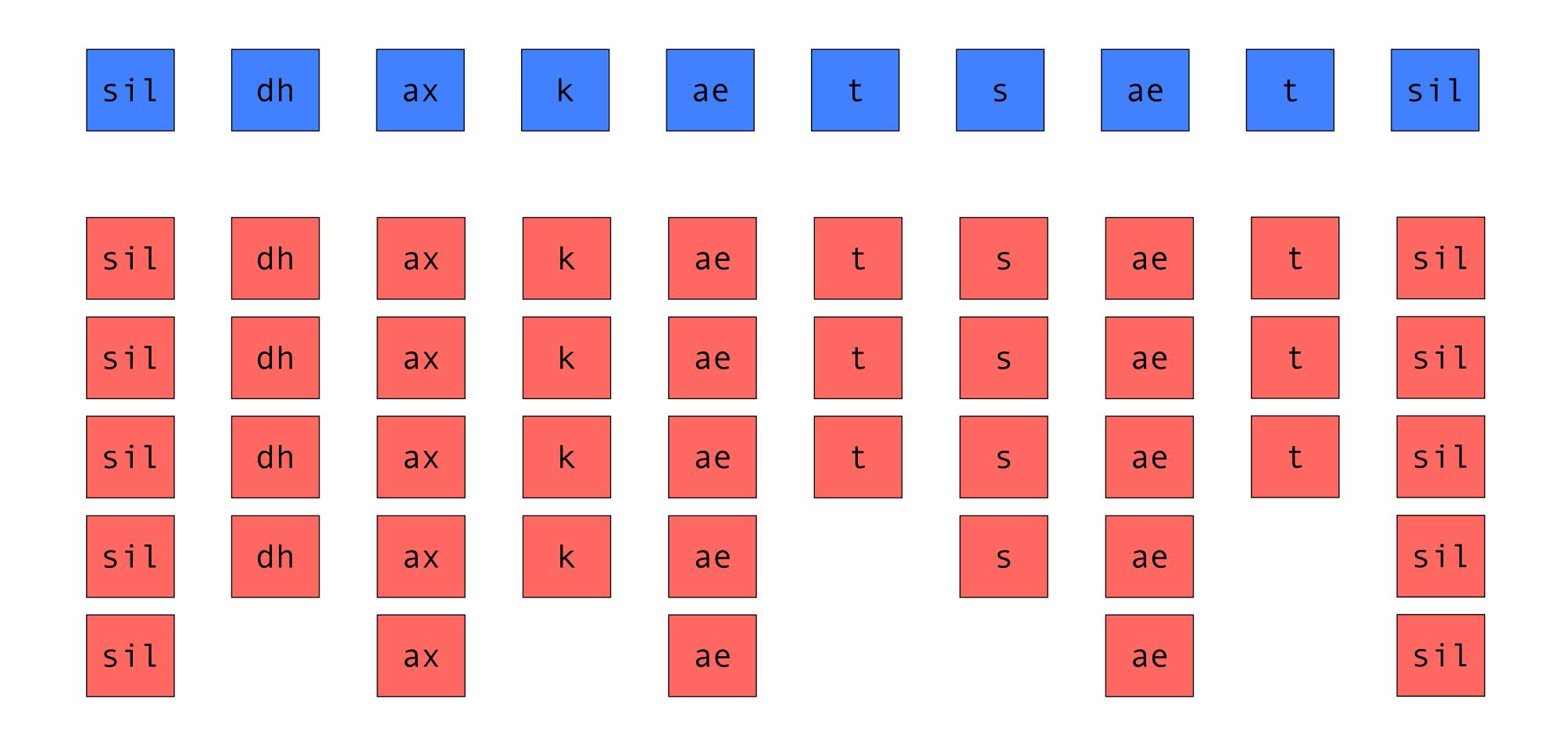
• listener and situation-**appropriate synthesis** (what Taylor calls "Speaking in a way that



A tour of the remaining modules

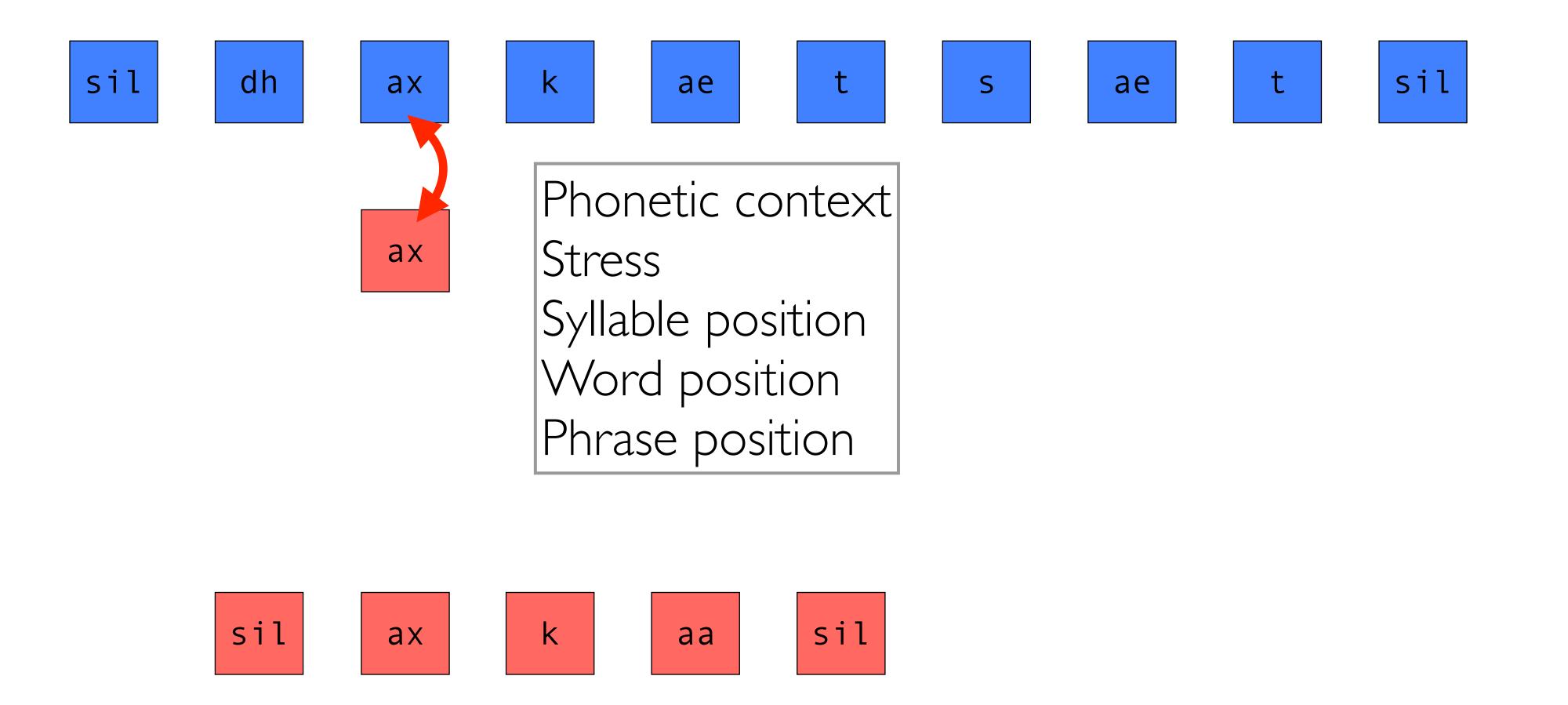
Module I - introduction Class

Module 2 - unit selection



Module 2 - unit selection Video 4 - Target cost and join cost

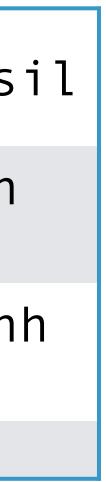
Module 3 - unit selection target cost functions



Module 3 - unit selection target cost functions Video I - Independent Feature Formulation

Module 4 - the database

So I came here.	sil_s	s_ow ow_ay ay	_k k_ey ey_m m_	_hh hh_ih i	h_r r_s
Now we have final	y heard her. sil_n n_ax a	n_aw_aw_w_w_iy ax_l l_iy iy_h	y iy_hh hh_ae a a hh_er er_d d	ae_v v_f f_ _hh hh_er e	ay ay_n r_sil
Those chefs know are.		— — — —	_sh_sh_eh_eh_f y_aa_aa_r_r_si		ow ow_h
etc					
aa_aa aa_f aa_ae aa_g aa_ah aa_hh aa_ao aa_ih aa_aw aa_iy aa_ay aa_jh aa_b aa_k aa_ch aa_l aa_d aa_m aa_dh aa_n aa_eh aa_ng aa_er aa_ow aaheyatabase aa_oy 2 - Script design	ay_ey ay_f ay_g ay_hh ay_ih ay_iy ay_jh ay_l ay_l ay_n ay_ng ay_ow	ey_f ey_g ey_hh ey_ih ey_iy ey_jh ey_k ey_l ey_n ey_n ey_ng ey_ow ey_oy	<pre>hh_f hh_g hh_hh hh_ih hh_iy hh_jh hh_k hh_l hh_l hh_n hh_n hh_n hh_ng hh_ow hh_oy</pre>	zh_f zh_g zh_hh zh_ih zh_iy zh_jh zh_j zh_l zh_l zh_n zh_n zh_ng zh_ow zh_oy	zh_p zh_r zh_s zh_sh zh_t zh_t zh_u zh_u zh_u zh_v zh_v zh_y zh_z zh_z



Module 5 - evaluation

Section 2: Part 1 / 13

In this section, after you listen to each sentence, you will choose a score for the audio file you've just heard.

This score should reflect your opinion of how natural or unnatural the sentence sounded.

Note that you should not judge the grammar or content of the sentence, just how it sounds.

Listen to the example below.

4 b

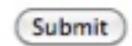
Then choose a score for how natural or unnatural the sentence sounded.

The scale is from 1 [Completely Unnatural] to 5 [Completely Natural].

4 : Mostly N

Module 5 - evaluation Video 2 - Subjective evaluation

atural	-	
atural		



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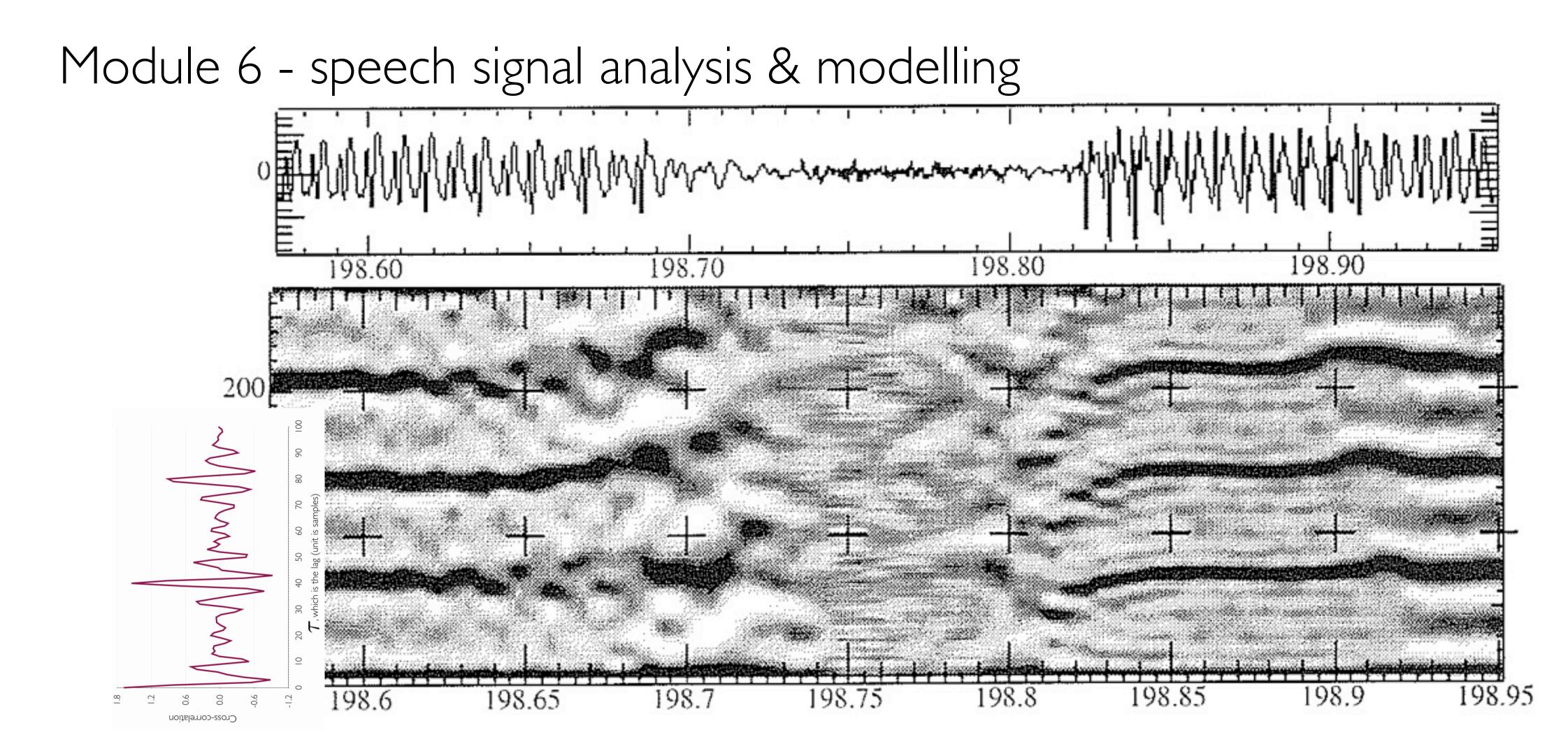
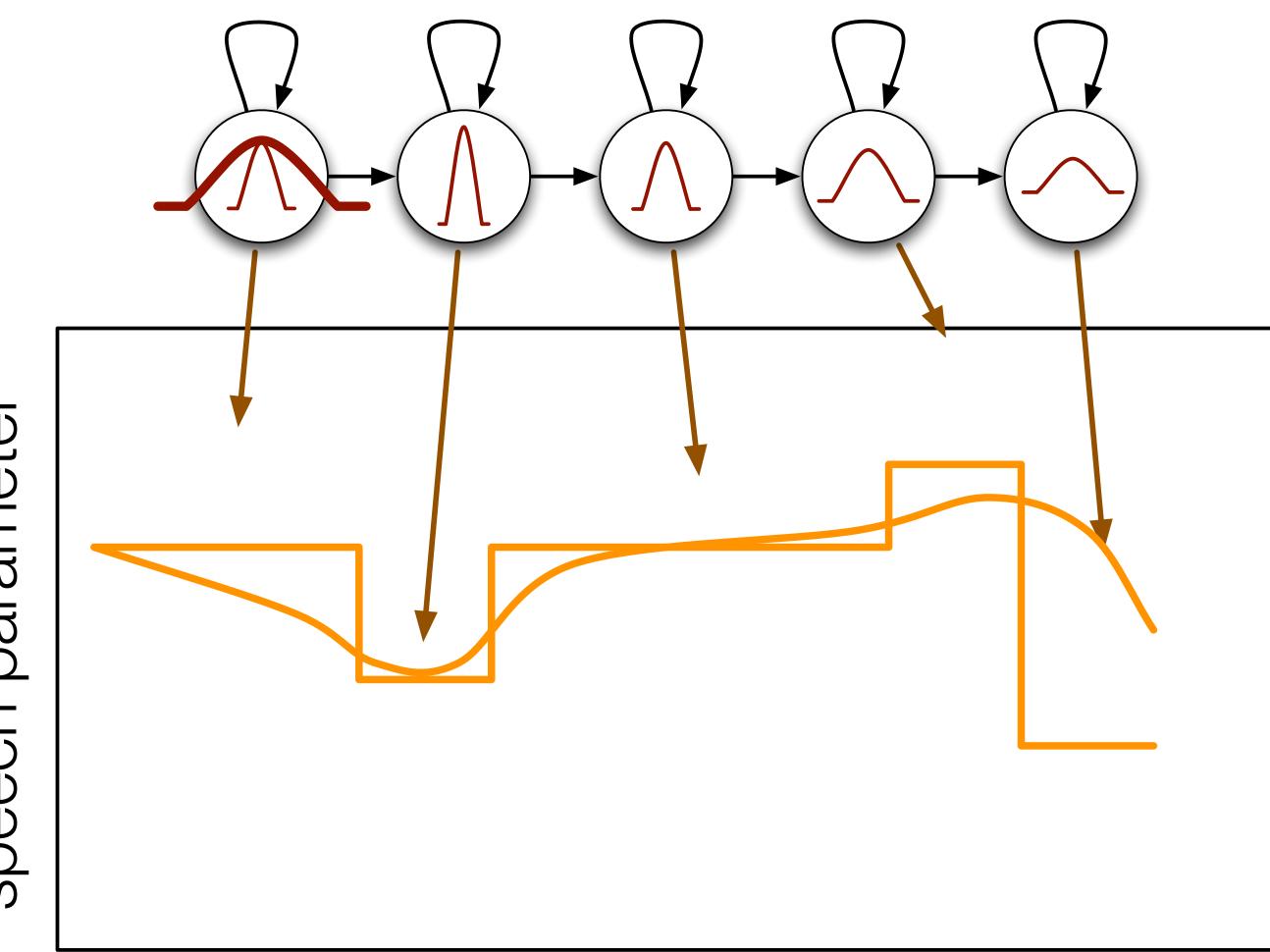


Figure 2 from David Talkin "A Robust Algorithm for Pitch Tracking (RAPT)" in Speech Coding and Synthesis, W. B. Kleijn and K. K. Palatal (eds), pages 497-518 Elsevier Science B.V., 1995

Module 6 - speech signal analysis & modelling Video 4 - F0 estimation (part 2)



Module 7 - Statistical Parametric Speech Synthesis (SPSS)



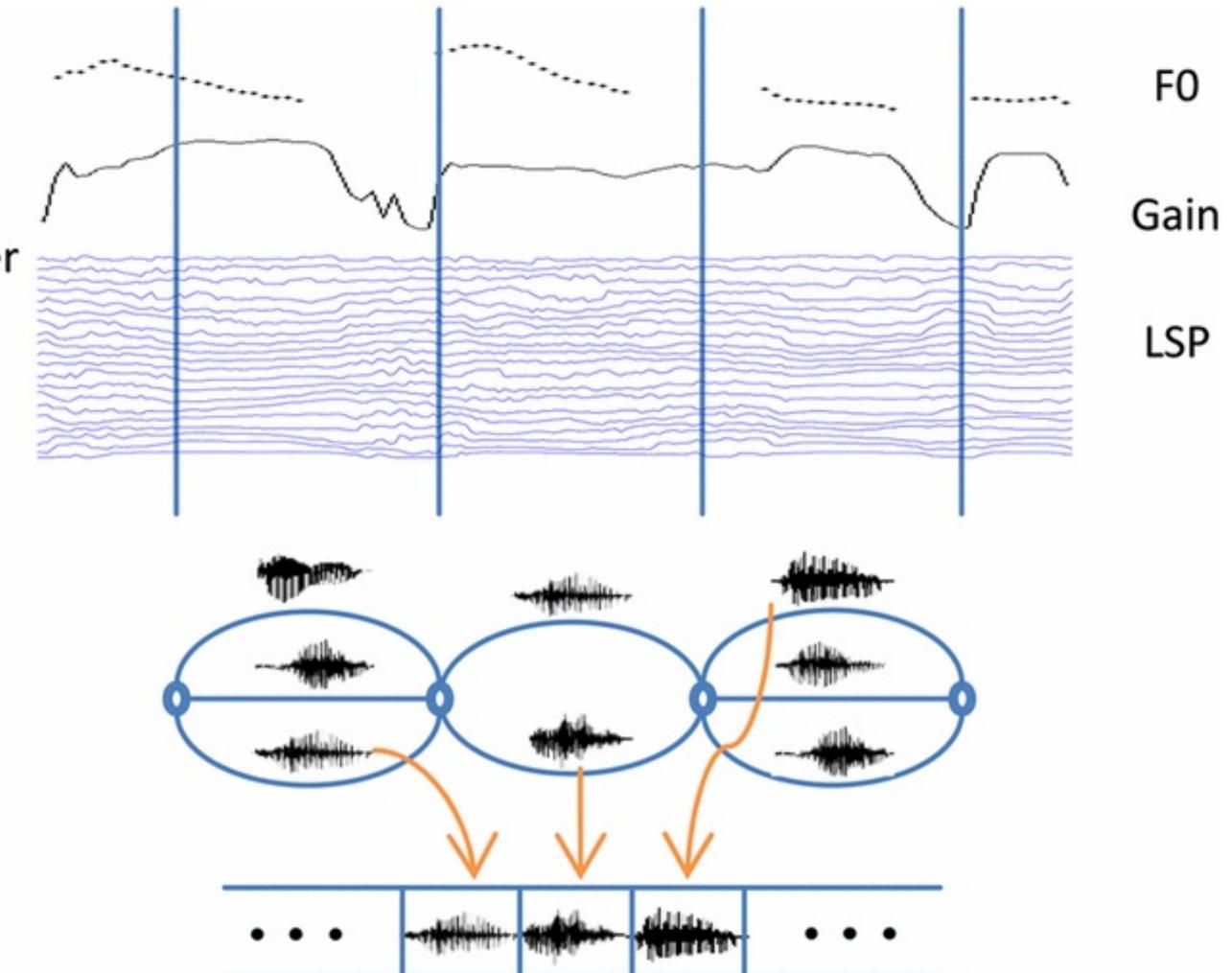
speech parameter

Module 7 - statistical parametric speech synthesis Video 4 - Wrap up

time

Module 7 bonus material (non-examinable) - hybrid speech synthesis

Guiding parameter trajectories



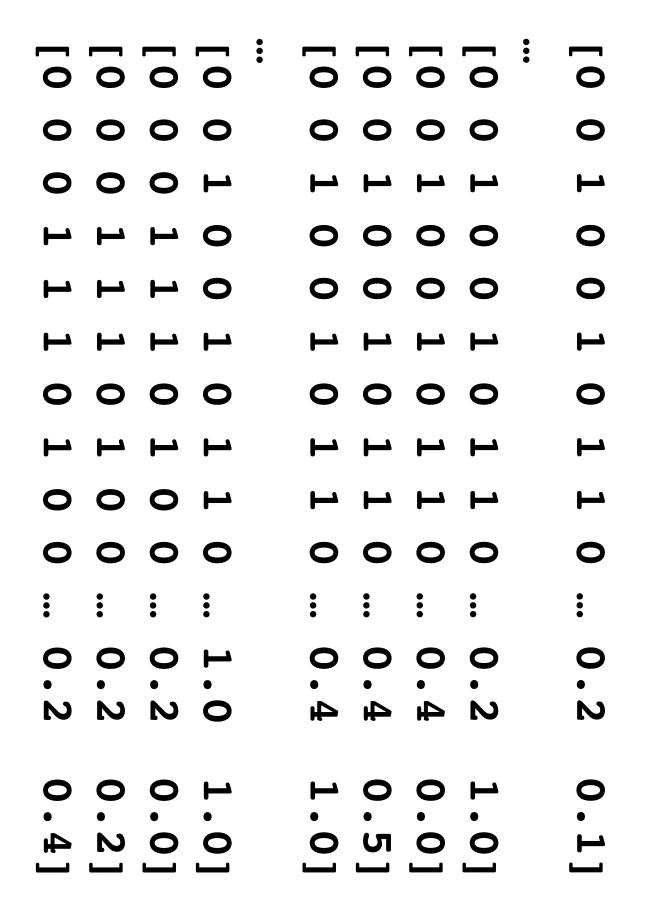
"Sausage" of waveform tiles

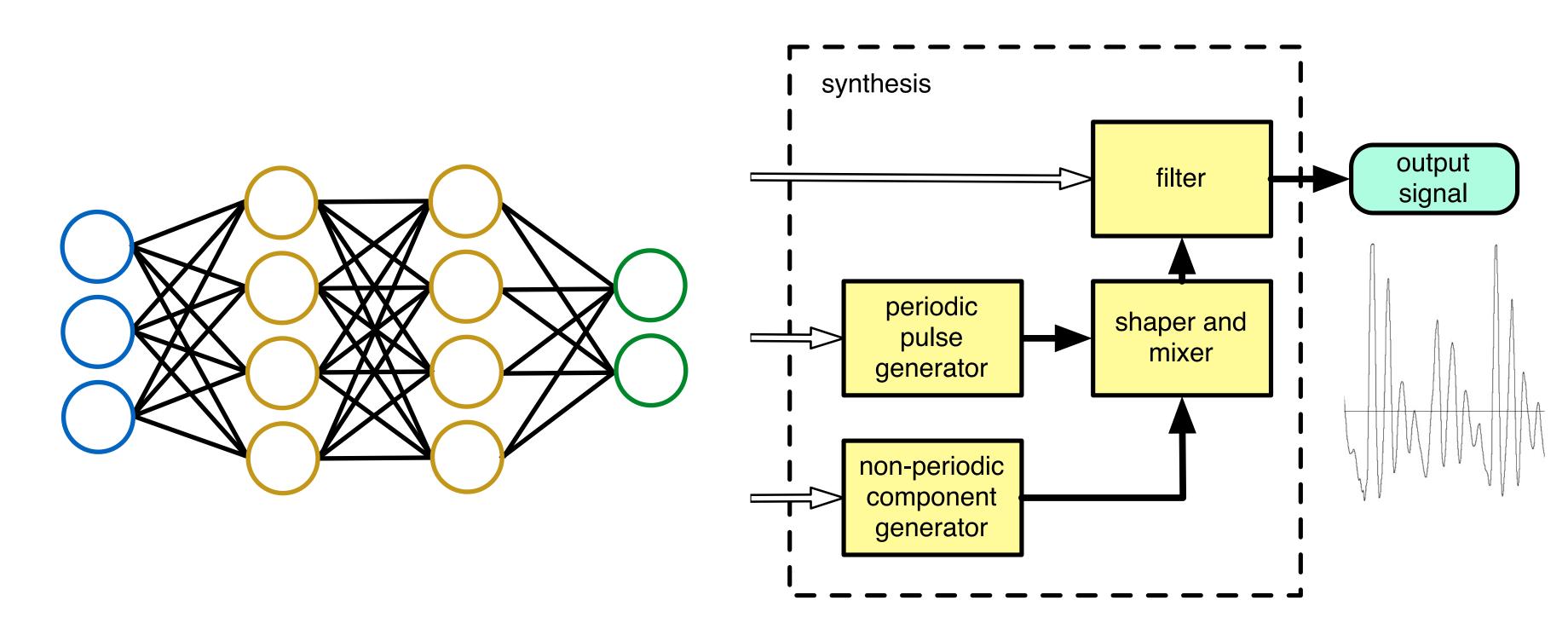
Waveform tile concatenation

Module 9 - hybrid speech synthesis Video 2 - Trajectory tiling

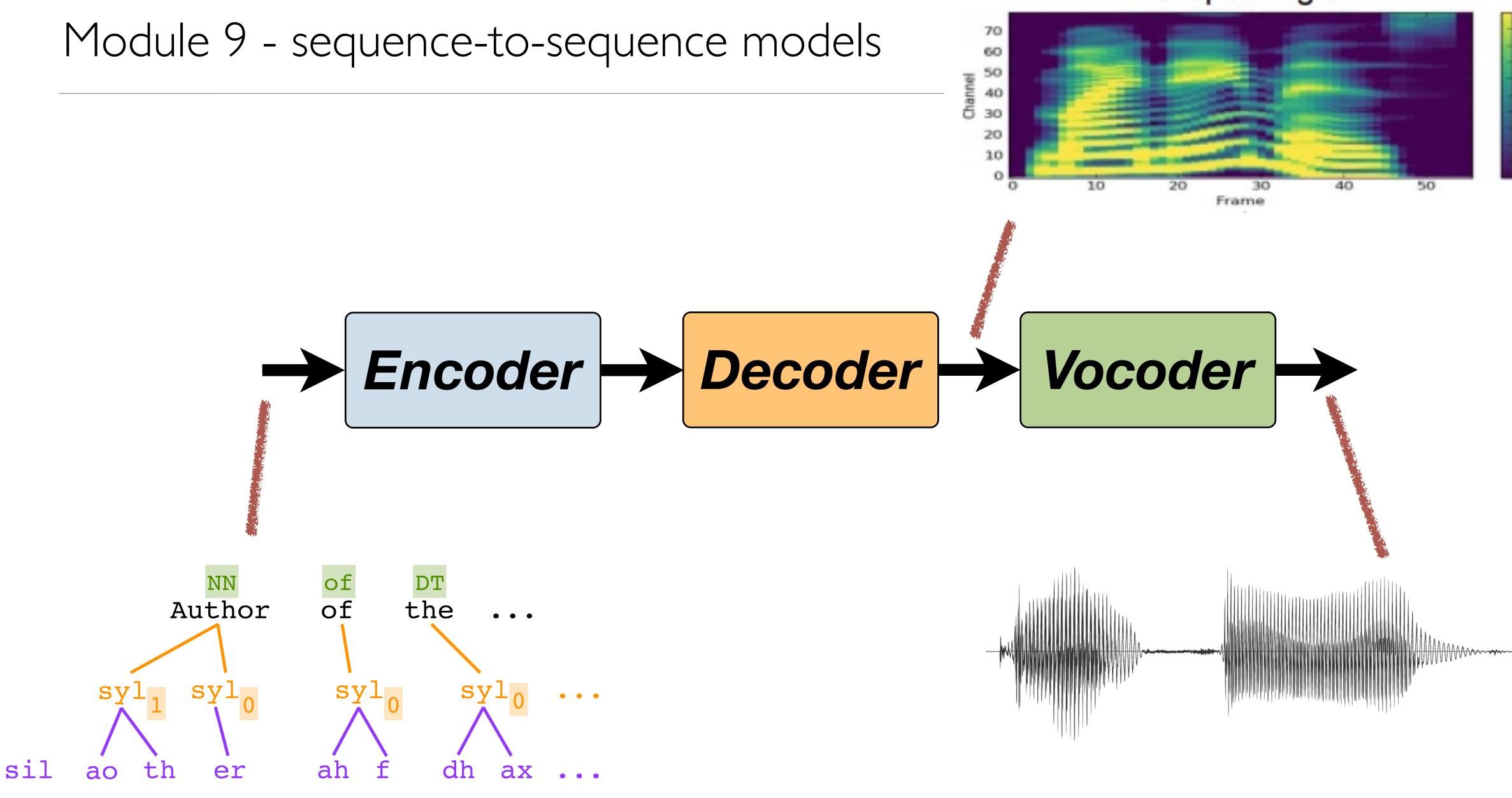
Module 8 - Deep Neural Networks (DNNs)

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Module 8 - speech synthesis using Neural Networks Video 2 - Doing Text-to-Speech



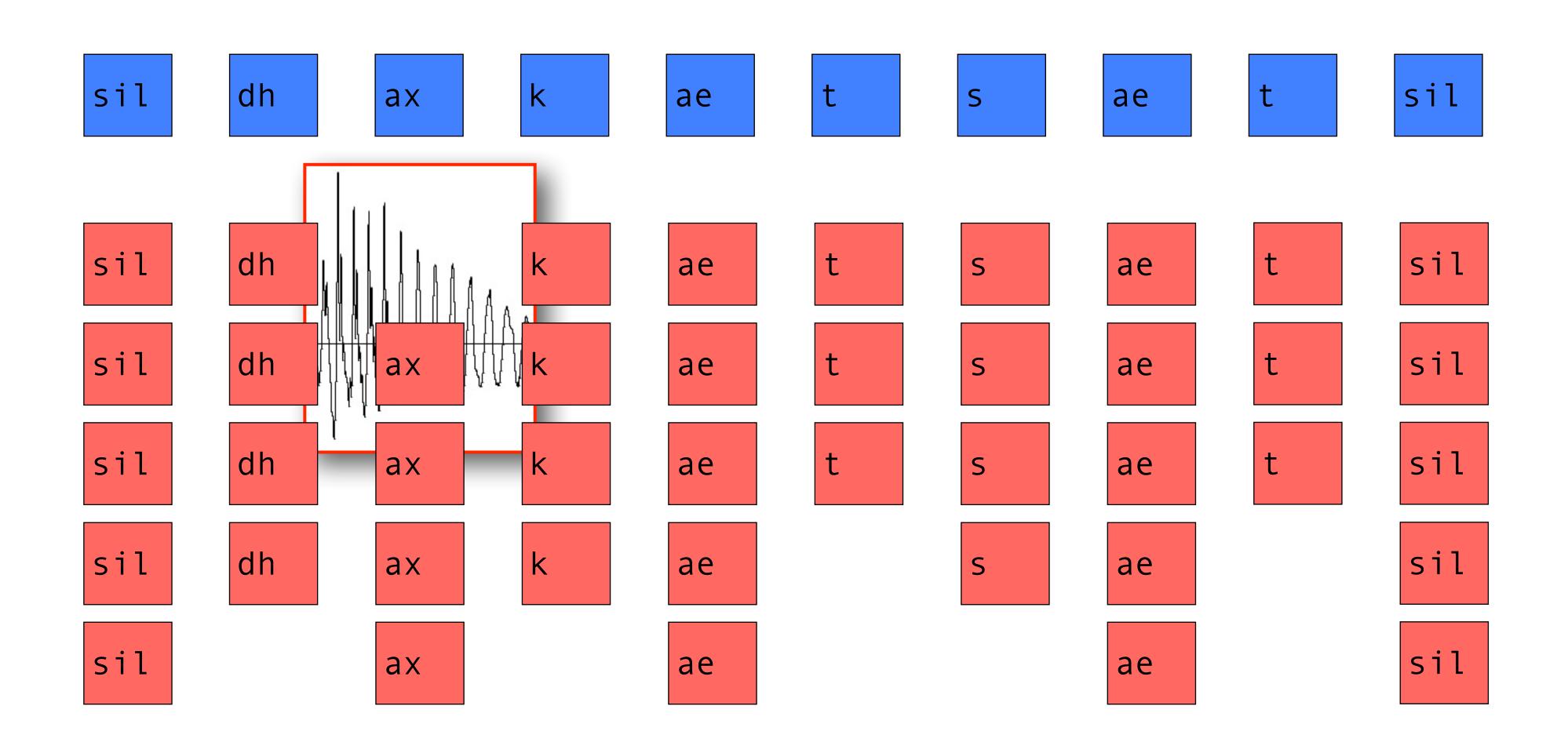
mel spectrogram

4.	0	
з.	2	
2	4	
1.	6	
0.	8	
0.	0	
-	0.	8
-	1	6
-	2	4
-	з.	2
-	4	0

The state of the art (2 further lectures after Module 9)

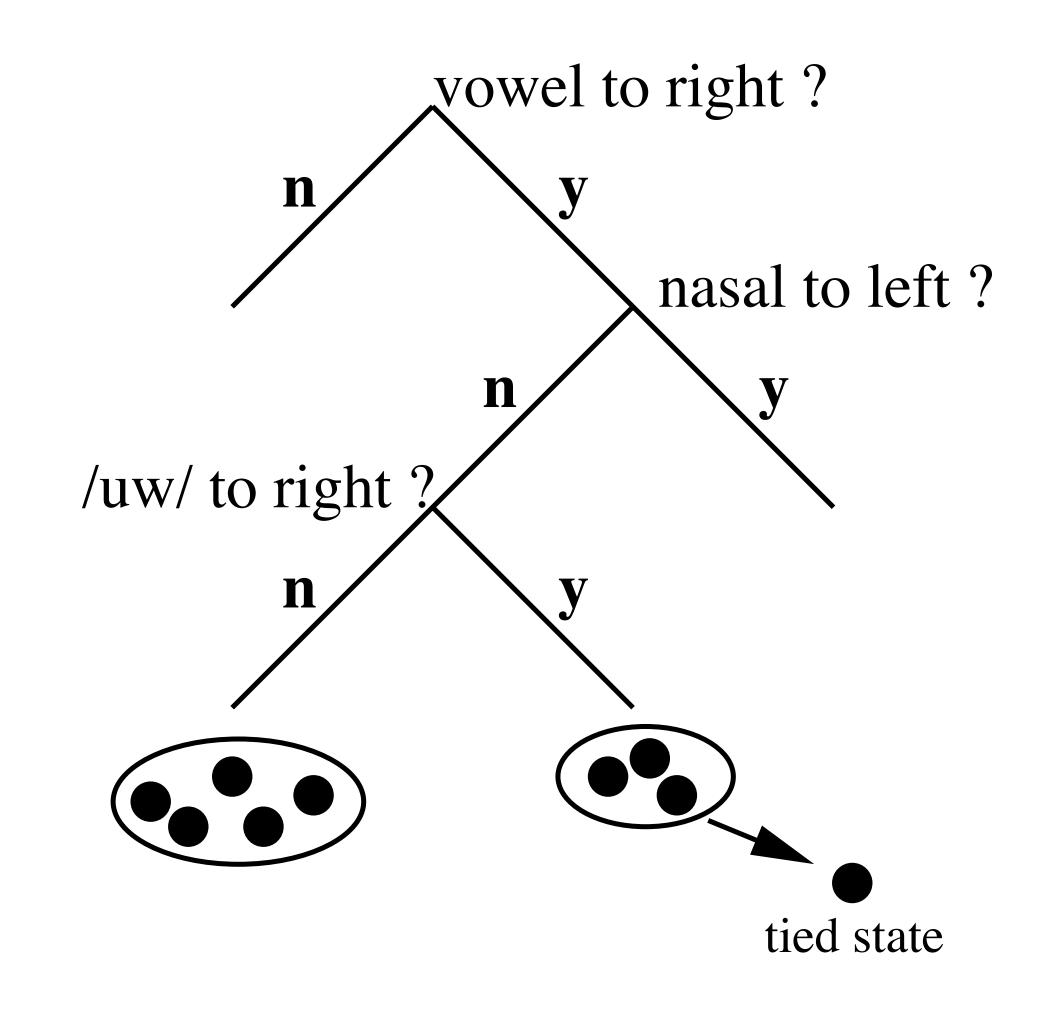
We will write these lectures "just in time" !

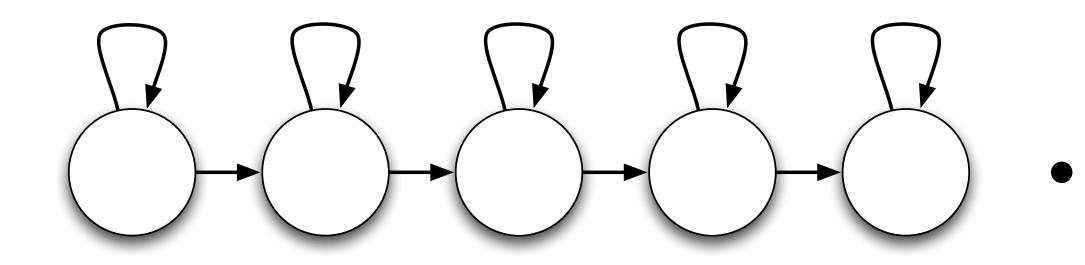
Context is everything : unit selection



Module 2 - unit selection Video 3 - Target and candidate units

Context is everything : HMMs (+ regression trees)

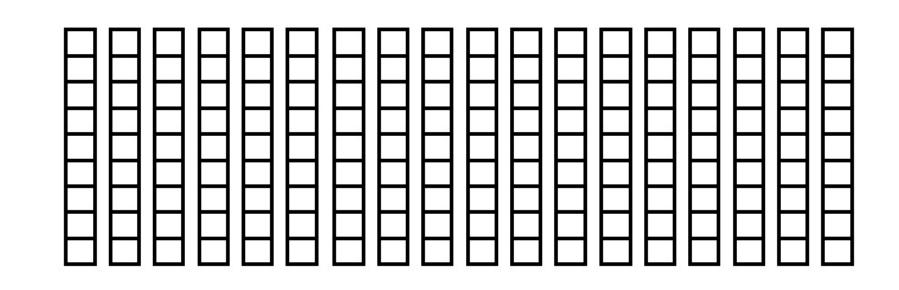


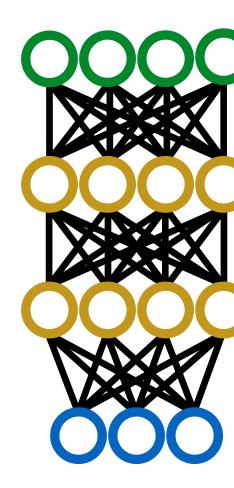


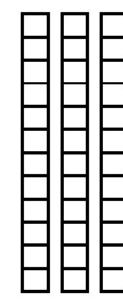


Context is everything : first attempts using Deep Neural Networks

output sequence

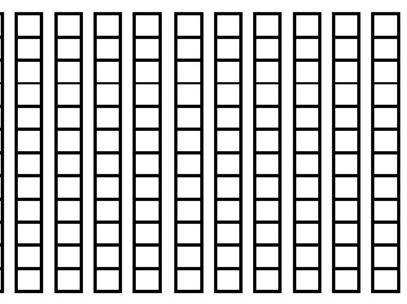




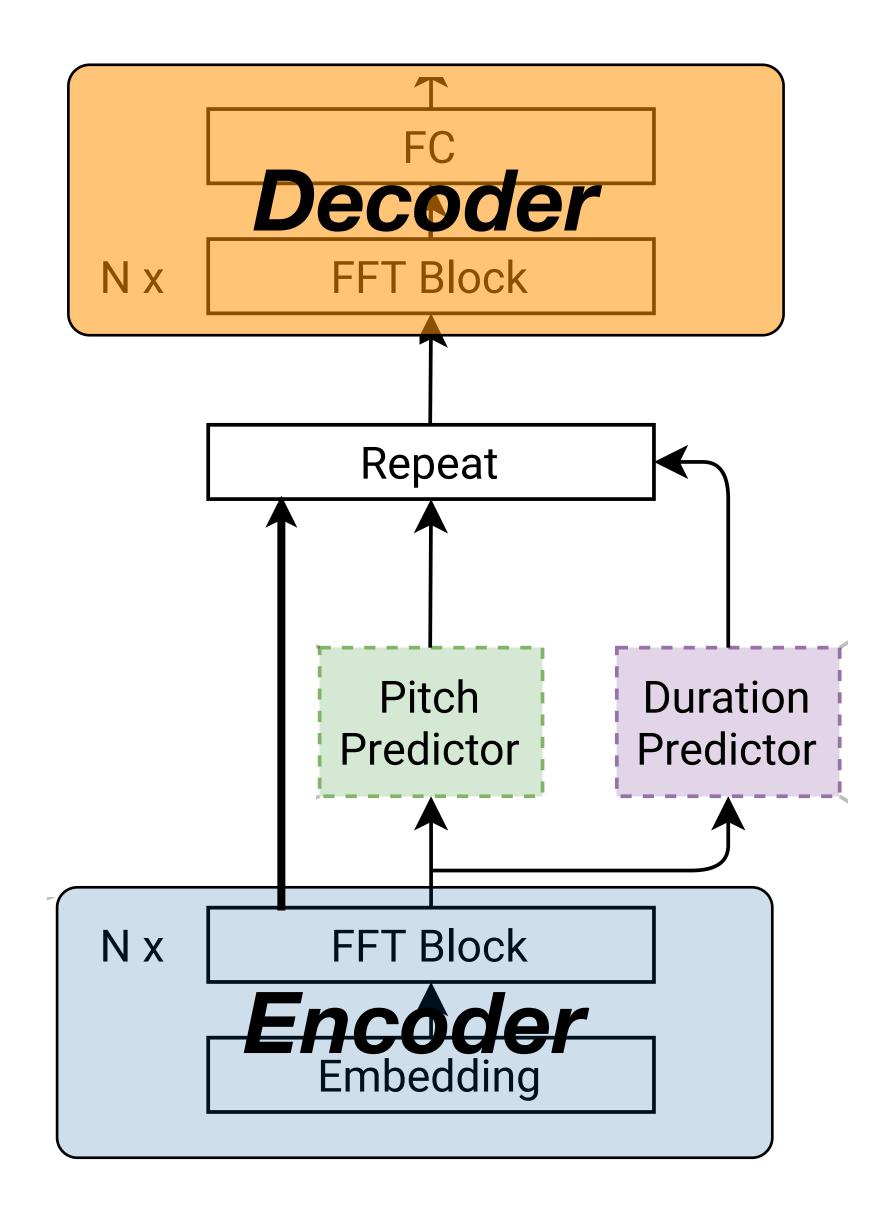


input sequence

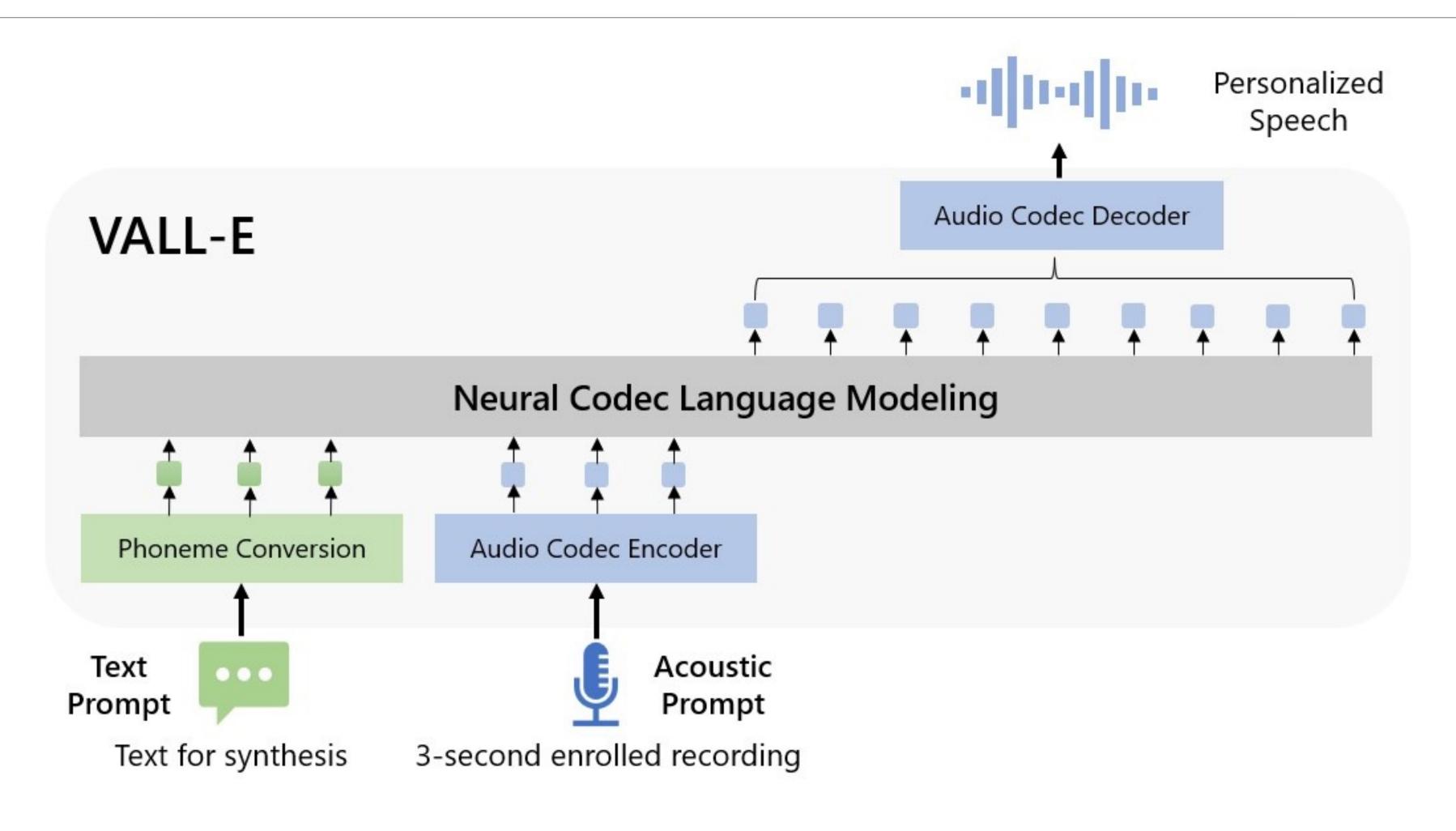
Module 9 - sequence-to-sequence models Class



Context is everything : encoder-decoder (e.g., FastPitch)

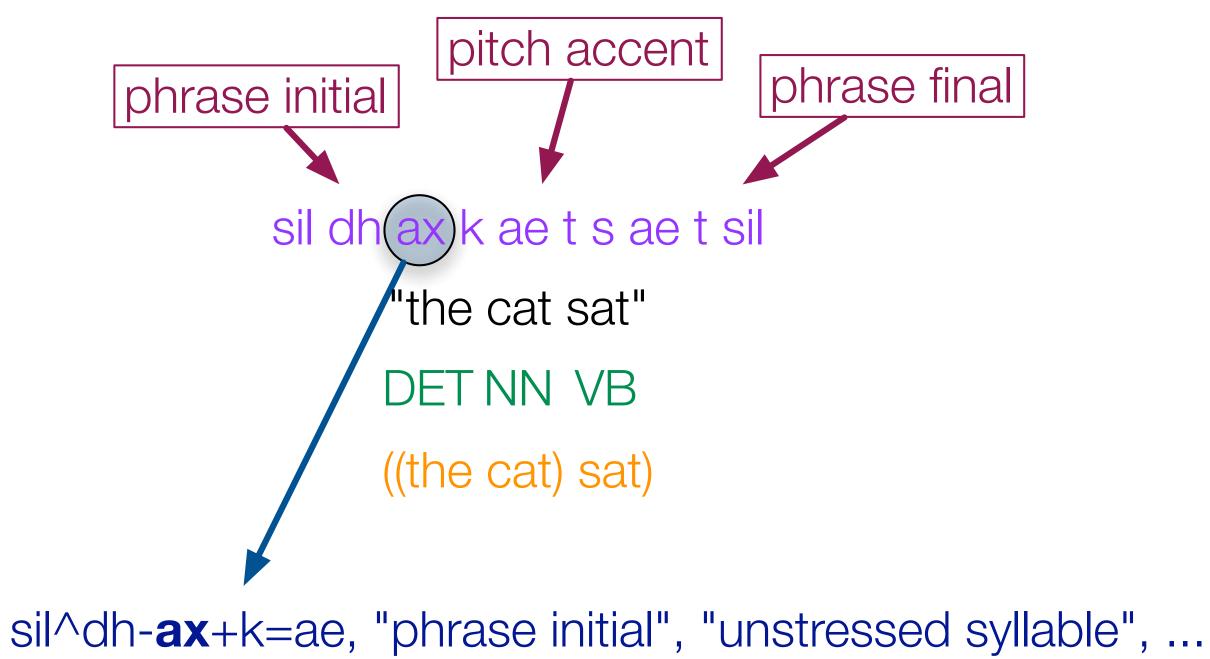


Context is everything : (large) language model approaches

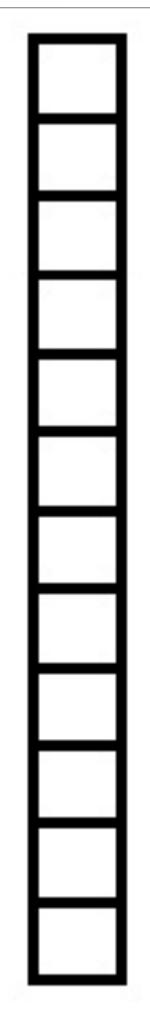


https://www.microsoft.com/en-us/research/project/vall-e-x

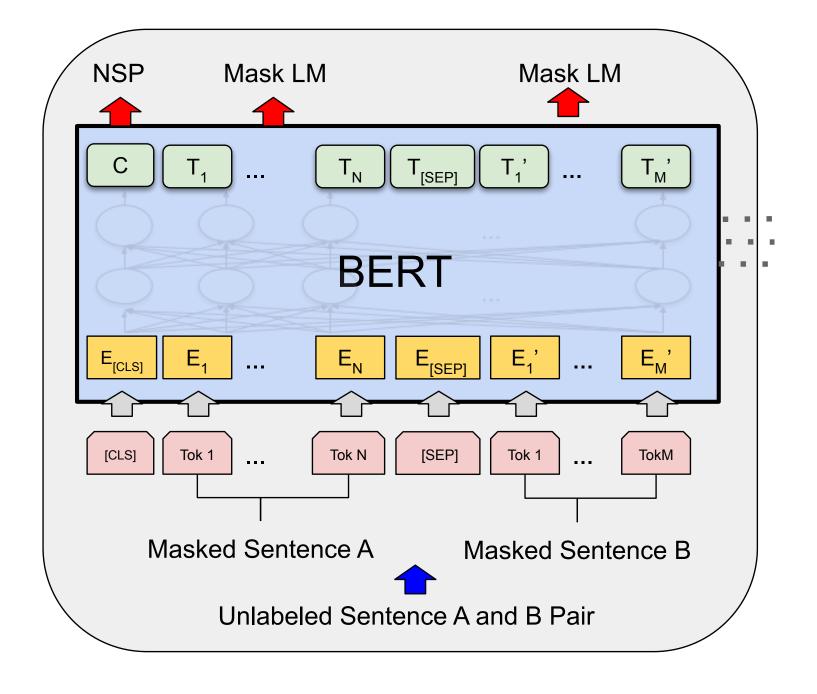




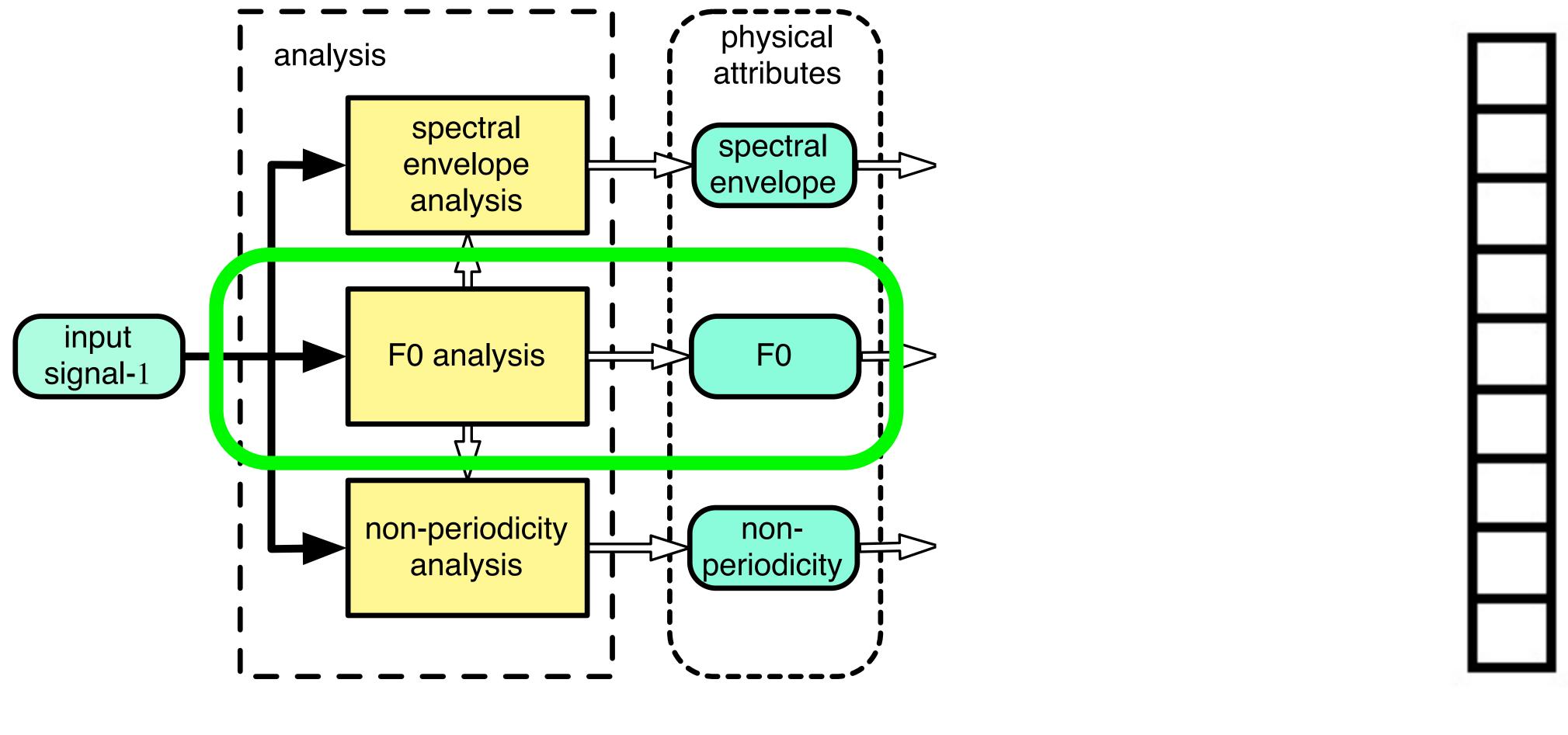
Module 7 - statistical parametric speech synthesis Video I - Text-to-Speech as a regression problem phrase final



input feature vector



,,



speech parameters

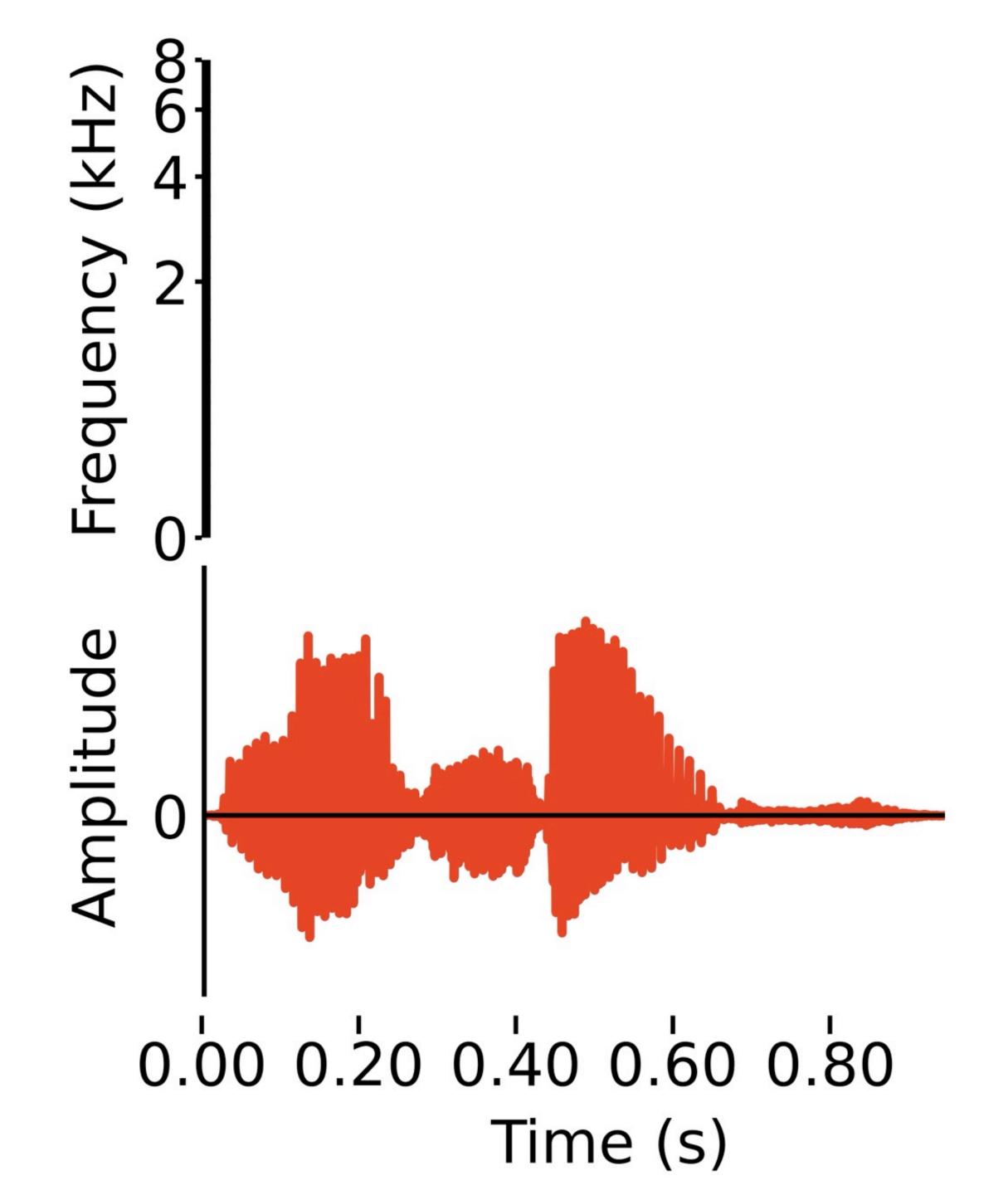
Module 7 - statistical parametric speech synthesis Video I - Text-to-Speech as a regression problem output feature vector

,,

Amplitude						
0						
Ċ)	5	10	15	20	25
			Tim	e (m	5)	

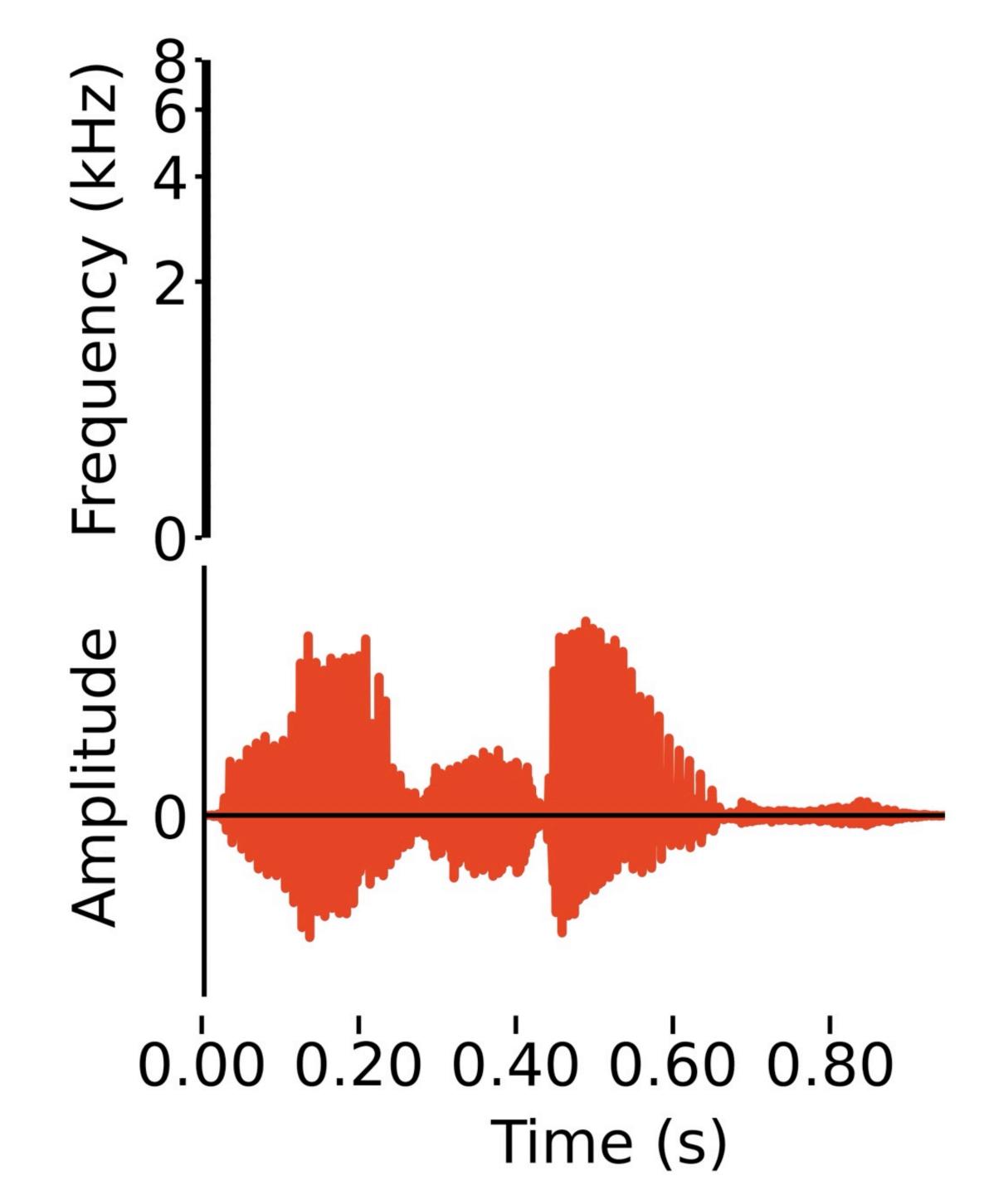
-3-Magnitude (dB) -5 -6 7 -8 -9ł 2 3 4 5 6 7 8 1 Frequency (kHz)





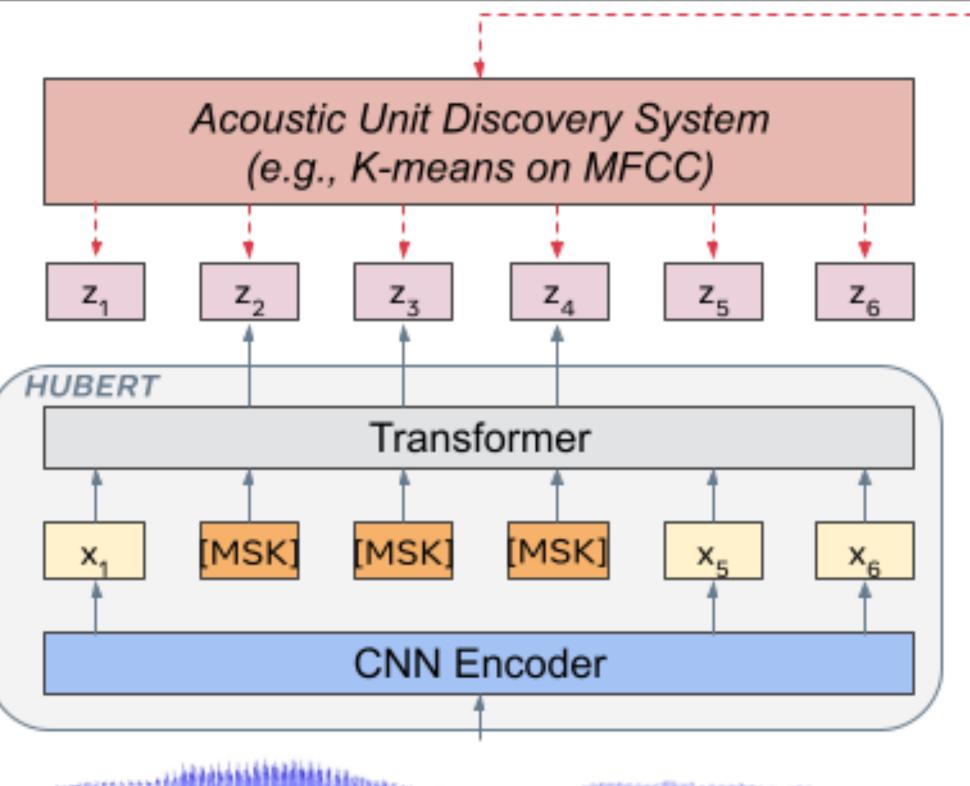
Magnitude (dB) 6 -8 -91 2 3 4 5 6 7 8 Frequency (kHz)





Magnitude (dB) 6 -8 -9+ 2 3 4 5 6 7 8 Frequency (kHz)





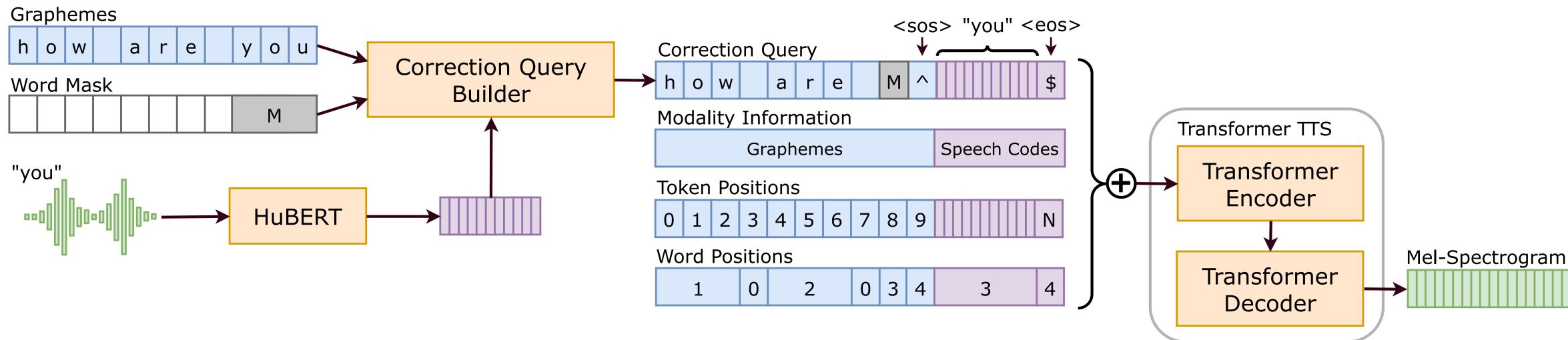
Hsu et al "HuBERT: Self-Supervised Speech Representation Learning by Masked Prediction of Hidden Units," in IEEE/ACM Transactions on Audio, Speech, and Language Processing, vol. 29, pp. 3451-3460, 2021, doi: 10.1109/TASLP.2021.3122291.

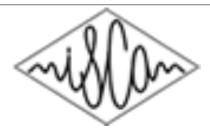


Interspeech 2022 18-22 September 2022, Incheon, Korea

Speech Audio Corrector: using speech from non-target speakers for one-off correction of mispronunciations in grapheme-input text-to-speech

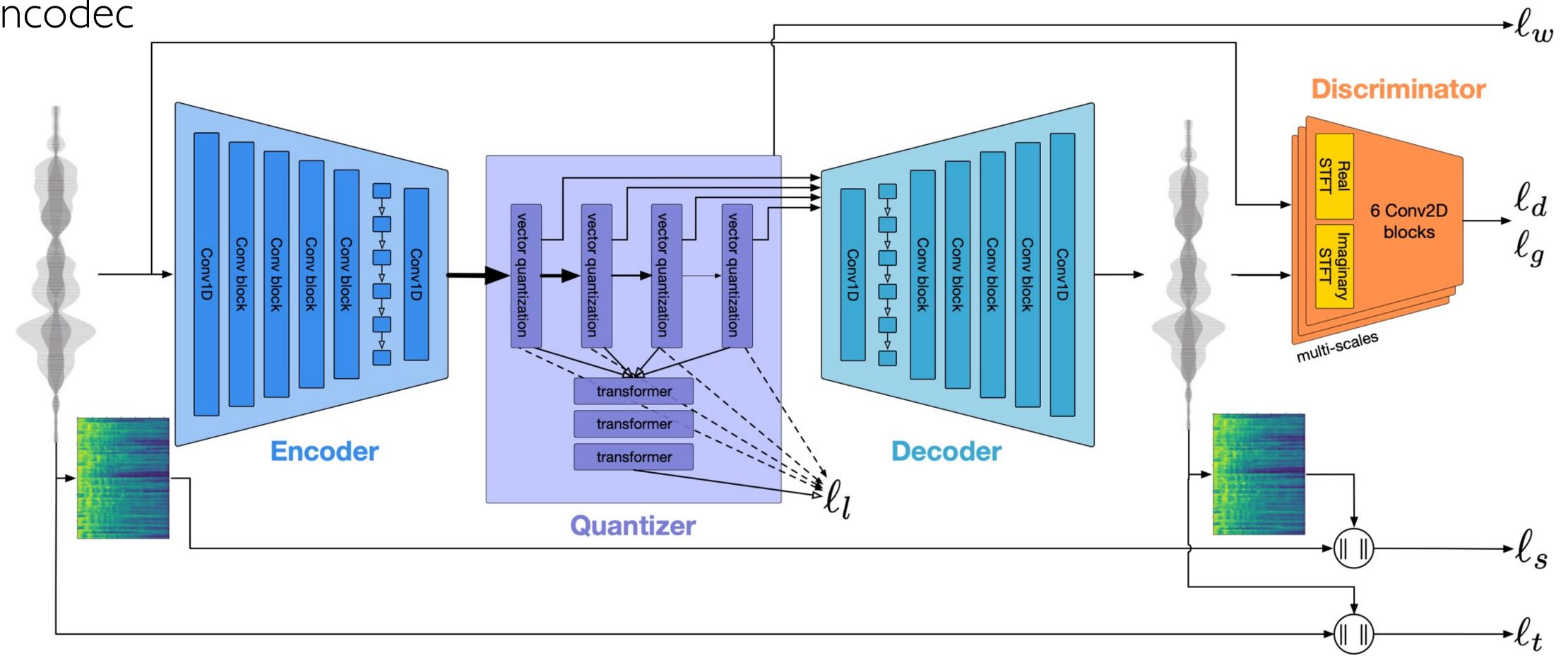
Jason Fong¹, Daniel Lyth¹, Gustav Eje Henter², Hao Tang¹, Simon King¹







Encodec



https://github.com/facebookresearch/encodec



Introduction to the coursework - see speech.zone

Module I - introduction Class