

DANSA

A compact
statistical
model of
coarticulatory
dynamics

introduction

correlations

detection
algorithm

trajectories

summary

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A compact statistical model of coarticulatory dynamics

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11 January 2007

Overview

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summary

- Dealing with context-dependency:
 - speech isn't like "beads on a string", it's a dance!
 - coarticulation applicable in synthesis & robust ASR
 - compact models for efficient exploitation of data
- Building a model of coarticulation:
 - Introduction
 - Critical, dependent & redundant articulation
 - Modelling dependencies dynamically
 - Preliminary evaluation of generated trajectories
 - Summary

Related contributions

- Statistical trajectory-modelling techniques:
 - Tokuda (trajectory HMM), Frankel & King (LDM)
 - Ostendorf, Russell, Holmes (segment models)
 - Deng et al. (dynamic Bayesian networks)
- Articulatory kinematics:
 - Öhman, Mermelstein (“knots and beads on a string”)
 - Lindblom (targets & smoothing)
 - Kaburagi, Honda, Dang, Löfqvist (phone distributions, context-dependent weights)
 - Blackburn & Young (statistics of articulator inertia)
- Biomechanical dynamics:
 - Coker (articulatory priority)
 - Saltzman, Tremblay & Ostry (task dynamics)

Articulatory correlations & rôles

Three main kinds of correlation in articulatory data:

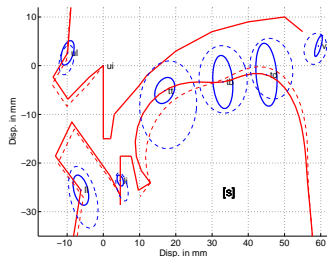
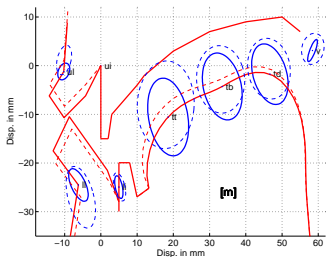
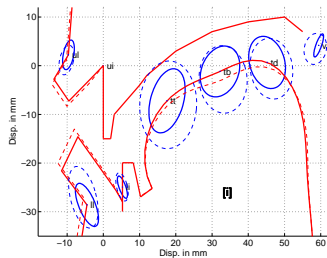
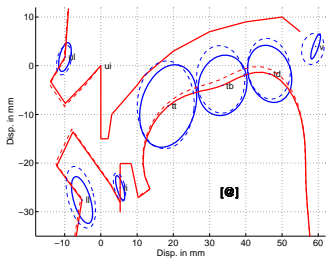
- 1 correlation for each articulator in **space**
- 2 correlations **between** articulators
- 3 correlations of trajectories over **time**

While speaking, articulators continually change their rôle:

- **critical** articulator
 - specified position (or gesture) is essential to phone
- **dependent** articulator
 - movement follows that of correlated critical articulator(s)
- **redundant** articulator
 - irrelevant, only constrained by surrounding context

Study relations 1 & 2 statistically to reveal coarticulation model's structure, based on Mocha data.

Phone distributions



Critical-artic. detection algorithm

- 1 Initialise model to grand distributions (Gaussian)
- 2 Compute phone distributions
- 3 Calculate intra- and inter-articulator correlations:
 - calculate grand correlations and test for significance
 - set low and insignificant values to zero

Univariate correlation matrix

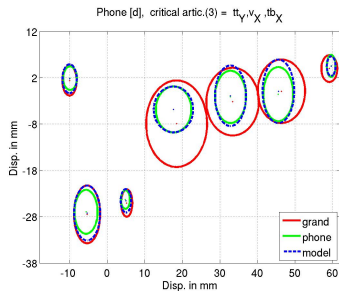
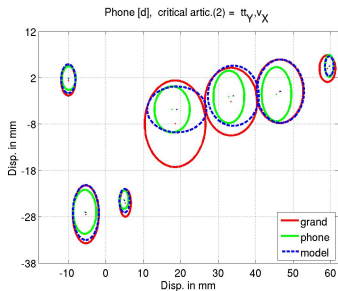
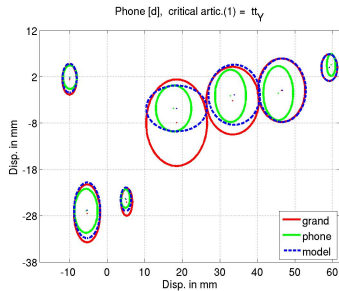
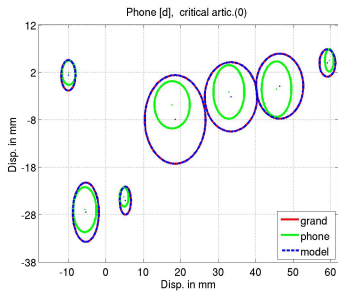
1D grand correlation matrix

	ul_x	ul_y	ll_x	ll_y	li_x	li_y	tt_x	tt_y	tb_x	tb_y	td_x	td_y	v_x	v_y
ul_x	1.00	.53	.34	-.15	.00	-.17	-.18	.00	-.16	.00	-.22	.00	.00	.00
ul_y	.53	1.00	.27	-.31	.00	.00	.00	.29	.00	.19	-.15	.00	.13	.00
ll_x	.34	.27	1.00	-.70	.61	-.55	.00	-.31	-.19	.00	-.17	.14	.11	.00
ll_y	-.15	-.31	-.70	1.00	-.49	.65	.00	.32	.14	.00	.10	-.10	-.18	.00
li_x	.00	.00	.61	-.49	1.00	-.71	.00	-.43	.00	-.36	.00	.00	.12	.00
li_y	-.17	.00	-.55	.65	-.71	1.00	.00	.60	.12	.42	.00	.00	-.12	.00
tt_x	-.18	.00	.00	.00	.00	.00	1.00	.00	.90	.00	.82	.00	.24	.19
tt_y	.00	.29	-.31	.32	-.43	.60	.00	1.00	.18	.53	.11	.00	.00	.00
tb_x	-.16	.00	-.19	.14	.00	.12	.90	.18	1.00	.00	.92	-.24	.14	.00
tb_y	.00	.19	.00	.00	-.36	.42	.00	.53	.00	1.00	.00	.75	.00	.00
td_x	-.22	-.15	-.17	.10	.00	.00	.82	.11	.92	.00	1.00	-.21	.00	.00
td_y	.00	.00	.14	-.10	.00	.00	.00	.00	-.24	.75	-.21	1.00	.00	.23
v_x	.00	.13	.11	-.18	.12	-.12	.24	.00	.14	.00	.00	.00	1.00	.81
v_y	.00	.00	.00	.00	.00	.00	.19	.00	.00	.00	.00	.23	.81	1.00

Critical-artic. detection algorithm

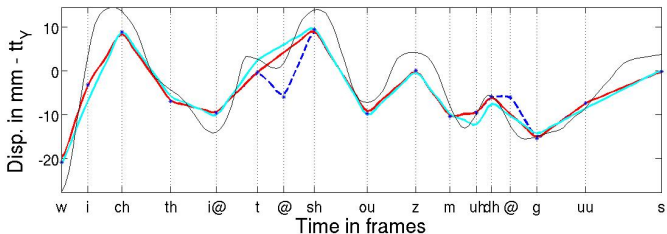
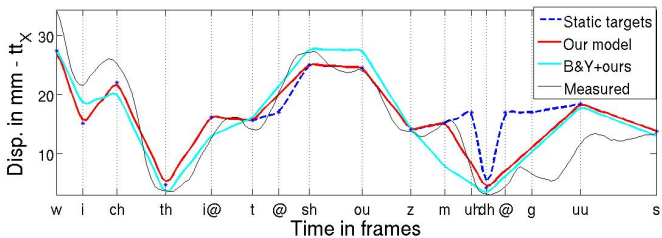
- 1 Initialise model to grand distributions (Gaussian)
- 2 Compute phone distributions
- 3 Calculate intra- and inter-articulator correlations:
 - calculate grand correlations and test for significance
 - set low and insignificant values to zero
- 4 Compute modified Kullback-Leibler divergences:
 - calculate std. error for phone and model pdfs
 - compute KL divergence including std. errors
- 5 Select articulator with max. divergence
 - test whether the value is above divergence threshold
 - set model pdf to phone pdf for critical articulator
- 6 Update model distributions:
 - for each dependent articulator, apply conditional mean and covariance based on critical articulator(s)
- 7 Repeat from step 4.

Univariate model convergence



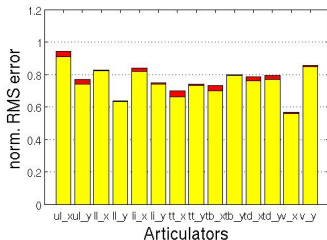
Generation of trajectories

"Which theatre shows Mother Goose?"

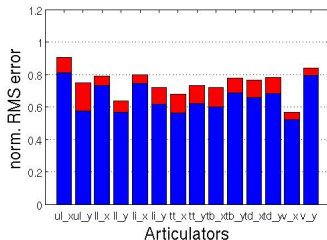


Average model error

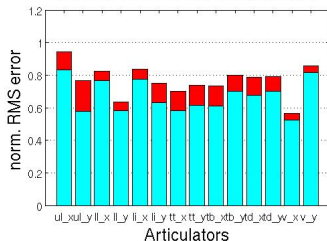
Our model: 2.3% reduction



B&Y model: 12.3% reduction



Combined: 12.6% reduction



Summary

- Coarticulatory model of speech dynamics
 - Tested for significant correlations (1D & 2D)
 - Detected critical articulator(s) for each phone, considering dependencies
 - Modified trajectory interpolation through redundancies
 - Preliminary results of model gave an improvement
- Future investigation
 - Development of trajectory interpolation (e.g., to 2D)
 - Detailed evaluation of the model's performance
 - Model incorporation into multi-level SEGREC system



Questions?