

Speech Recognition - EQ2340

HMM (Hidden Markov Models)

Alessio Russo



Control Engineering
(Politecnico di Milano)

Lars Lindemann



Systems, Control and
Robotics (KTH)

Royal Institute of Technology (KTH, Stockholm)
3rd November, 2015

1. Introduction

- Problem Formulation
- System architecture

2. System Design, Training and Testing

- Feature Extraction
- HMM
- Training data and validation set
- Testing and tweaking

3. Results

- System Performance
- Conclusion
- Live Demonstration

Speech Recognition of a limited speech corpus

Speech Recognition of a limited speech corpus

- ▶ High demand in industry
- ▶ Usage in current systems (e.g. Siri)
- ▶ Easy to understand general principle

Speech Recognition of a limited speech corpus

- ▶ High demand in industry
- ▶ Usage in current systems (e.g. Siri)
- ▶ Easy to understand general principle

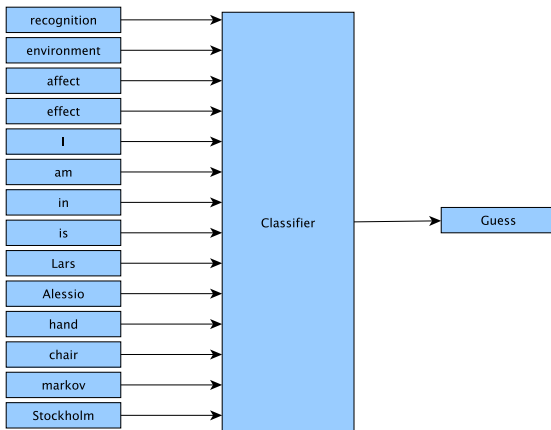
Speech Corpus

- ▶ General speech corpus to form sentences
- ▶ Multisyllabic, similar and short words to challenge the system

Introduction

Problem Formulation

Speech Recognition of a limited speech corpus



Speech Recognition of a limited speech corpus

- ▶ Two examples...

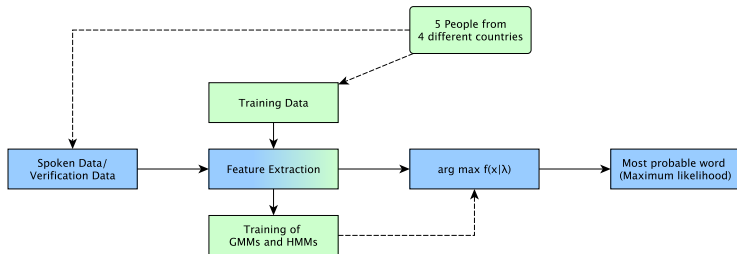
Lars: affect

Natalie: recognition

Introduction

System architecture

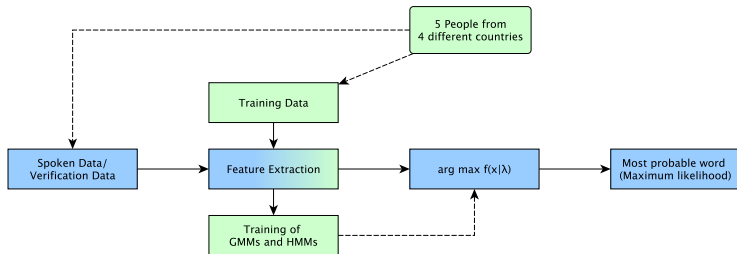
Overview of the Implementation



Introduction

System architecture

Overview of the Implementation



- Distinguish between training and validation/live demonstration

1. Introduction

- Problem Formulation
- System architecture

2. System Design, Training and Testing

- Feature Extraction
- HMM
- Training data and validation set
- Testing and tweaking

3. Results

- System Performance
- Conclusion
- Live Demonstration

Possible problems

- ▶ Pitch
- ▶ Different speakers (absolut output value)
- ▶ Noise

Possible problems

- ▶ Pitch
- ▶ Different speakers (absolut output value)
- ▶ Noise

Possible problems

- ▶ Pitch
- ▶ Different speakers (absolut output value)
- ▶ Noise

Possible problems

- ▶ Pitch
- ▶ Different speakers (absolut output value)
- ▶ Noise

Continuous feature vectors

- ▶ 13 MFCC (Mel-frequency cepstrum coefficients)
- ▶ 26 dynamical features (independent of absolute value)
- ▶ 30 ms time frame

Possible problems

- ▶ Pitch
- ▶ Different speakers (absolut output value)
- ▶ Noise

Continuous feature vectors

- ▶ 13 MFCC (Mel-frequency cepstrum coefficients)
- ▶ 26 dynamical features (independent of absolute value)
- ▶ 30 ms time frame

Possible problems

- ▶ Pitch
- ▶ Different speakers (absolut output value)
- ▶ Noise

Continuous feature vectors

- ▶ 13 MFCC (Mel-frequency cepstrum coefficients)
- ▶ 26 dynamical features (independent of absolute value)
- ▶ 30 ms time frame

Number of States for left-right HMM

- ▶ Trade off: too few parameters vs. amount of training data
- ▶ Also: Limited training data
- ▶ State assignment due to syllables + start/end state

Number of States for left-right HMM

- ▶ Trade off: too few parameters vs. amount of training data
- ▶ Also: Limited training data
- ▶ State assignment due to syllables + start/end state

Number of States for left-right HMM

- ▶ Trade off: too few parameters vs. amount of training data
- ▶ Also: Limited training data
- ▶ State assignment due to syllables + start/end state

Number of States for left-right HMM

- ▶ Trade off: too few parameters vs. amount of training data
- ▶ Also: Limited training data
- ▶ State assignment due to syllables + start/end state

environment: 6 states

chair: 4 states

Number of States for left-right HMM

- ▶ Trade off: too few parameters vs. amount of training data
- ▶ Also: Limited training data
- ▶ State assignment due to syllables + start/end state

environment: 6 states

chair: 4 states

Output distributions

- ▶ GMM (Gaussian Mixture Models)

System Design, Training and Testing

Training data and validation set

Recorded data

- ▶ 5 people: 15 recordings per word
- ▶ One person has been disregarded
- ▶ 840 recordings in total, 60 per word

System Design, Training and Testing

Training data and validation set

Recorded data

- ▶ 5 people: 15 recordings per word
- ▶ One person has been disregarded
- ▶ 840 recordings in total, 60 per word

k-fold approach

- ▶ $k=5$ sets
- ▶ 48 training and 12 validation samples

Final HMM

- ▶ 5 sets have been tweaked on their validation set and compared on the whole set.
- ▶ Recognition rate in table below

Final HMM

- ▶ 5 sets have been tweaked on their validation set and compared on the whole set.
- ▶ Recognition rate in table below

System Design, Training and Testing

Testing and tweaking

Final HMM

- ▶ 5 sets have been tweaked on their validation set and compared on the whole set.
- ▶ Recognition rate in table below

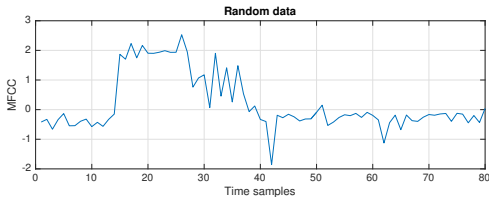
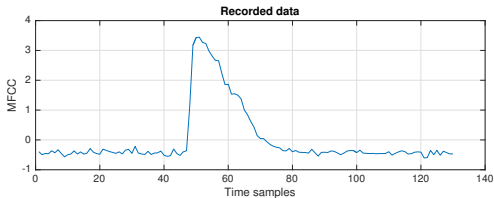
word	1	2	3	4	5	6	7
validation set	1.00	1.00	1.00	1.00	1.00	1.00	1.00
overall	0.76	0.95	1.00	1.00	1.00	1.00	1.00
word	8	9	10	11	12	13	14
validation set	1.00	1.00	1.00	0.833	1.00	1.00	1.00
overall	1.00	0.95	1.00	0.7	1.00	0.95	1.00

System Design, Training and Testing

Testing and tweaking

Some realizations

- ▶ word / and 1st MFCC coefficient

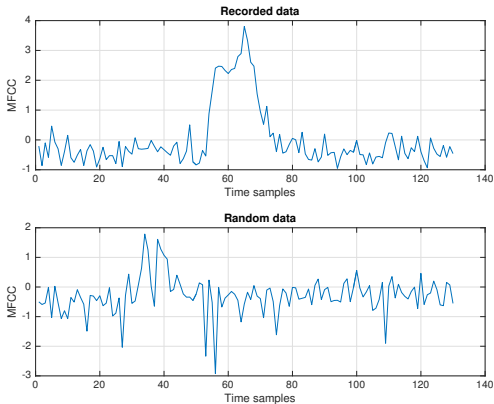


System Design, Training and Testing

Testing and tweaking

Some realizations

- ▶ word / and 2nd MFCC coefficient



1. Introduction

- Problem Formulation
- System architecture

2. System Design, Training and Testing

- Feature Extraction
- HMM
- Training data and validation set
- Testing and tweaking

3. Results

- System Performance
- Conclusion
- Live Demonstration

Classification Errors

- ▶ Average Classification Error: 1.2 %(validation) and 4.9 % (overall)
- ▶ Most commonly missclassified: *hand* with 16.6 % and 30 %

Classification Errors

- ▶ Average Classification Error: 1.2 %(validation) and 4.9 % (overall)
- ▶ Most commonly missclassified: *hand* with 16.6 % and 30 %

Results

System Performance

$$C = \begin{pmatrix} 46 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 14 & 0 & 0 & 0 & 0 \\ 0 & 57 & 0 & 0 & 3 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 60 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 60 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 60 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 60 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 60 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 60 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 57 & 0 & 1 & 0 & 0 & 2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 60 & 0 & 0 & 0 & 0 \\ 0 & 0 & 7 & 0 & 0 & 11 & 0 & 0 & 0 & 0 & 42 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 60 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 57 & 2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 60 \end{pmatrix}$$

Results

System Performance

Lars: recognition

Martin: recognition

Natalie: recognition

Take aways

- ▶ If data is rare, use smaller k in k -fold approach
- ▶ Good training data is important
- ▶ Collect more training data (remember trade-off)
but: then be aware of overfitting!

The system

- ▶ Satisfying overall recognition rate
- ▶ Problems with the words *hand* and *recognition*
- ▶ less problems with *affect* and *effect* or short words

Take aways

- ▶ If data is rare, use smaller k in k -fold approach
- ▶ Good training data is important
- ▶ Collect more training data (remember trade-off)
but: then be aware of overfitting!

The system

- ▶ Satisfying overall recognition rate
- ▶ Problems with the words *hand* and *recognition*
- ▶ less problems with *affect* and *effect* or short words

Take aways

- ▶ If data is rare, use smaller k in k -fold approach
- ▶ Good training data is important
- ▶ Collect more training data (remember trade-off)
but: then be aware of overfitting!

The system

- ▶ Satisfying overall recognition rate
- ▶ Problems with the words *hand* and *recognition*
- ▶ less problems with *affect* and *effect* or short words

Take aways

- ▶ If data is rare, use smaller k in k -fold approach
- ▶ Good training data is important
- ▶ Collect more training data (remember trade-off)
but: then be aware of overfitting!

The system

- ▶ Satisfying overall recognition rate
- ▶ Problems with the words *hand* and *recognition*
- ▶ less problems with *affect* and *effect* or short words

Take aways

- ▶ If data is rare, use smaller k in k -fold approach
- ▶ Good training data is important
- ▶ Collect more training data (remember trade-off)
but: then be aware of overfitting!

The system

- ▶ Satisfying overall recognition rate
- ▶ Problems with the words *hand* and *recognition*
- ▶ less problems with *affect* and *effect* or short words

Take aways

- ▶ If data is rare, use smaller k in k -fold approach
- ▶ Good training data is important
- ▶ Collect more training data (remember trade-off)
but: then be aware of overfitting!

The system

- ▶ Satisfying overall recognition rate
- ▶ Problems with the words *hand* and *recognition*
- ▶ less problems with *affect* and *effect* or short words

Results

Live Demonstration

...