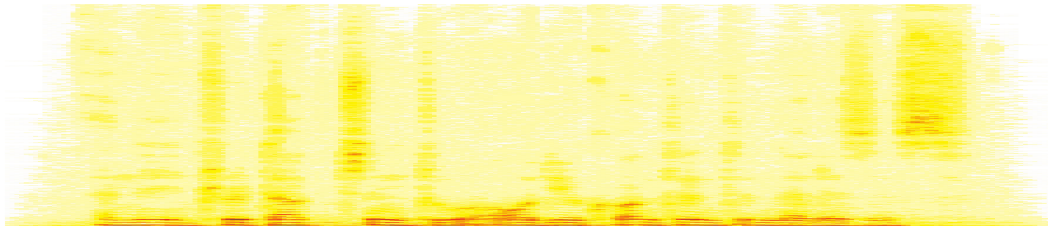


Introduction to Audio Content Analysis

Module 5.8: Chord Detection

alexander lerch



introduction

overview

corresponding textbook section

Chapter 5 — Tonal Analysis: pp. 125–127

- **lecture content**

- musical chords and harmony
- baseline chord detection
- Hidden Markov Models (HMMs) and the Viterbi algorithm

- **learning objectives**

- name basic chords and describe the concept of chord inversions
- discuss commonalities and differences between chord & key detection
- discuss the usefulness of HMMs for chord detection
- explain the Viterbi algorithm with an example



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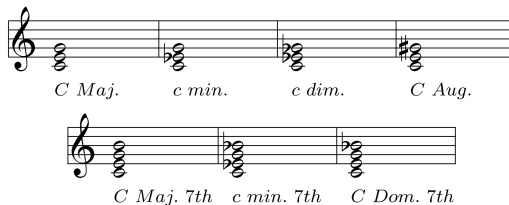
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musical pitch

chords

- simultaneous use of several pitches \Rightarrow **chords**
- usually constructed of (major/minor) thirds

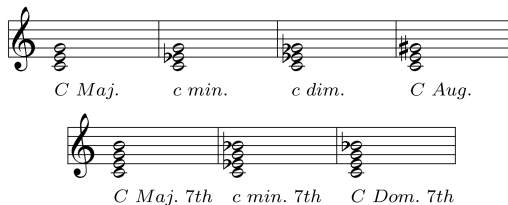


- note:
 - chord type independent of pitch doubling, pitch order
 - same label for keys and chords

musical pitch

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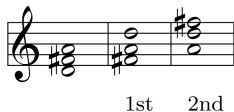


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musical pitch

chord inversion

- most common: root note is lowest note
- otherwise: chord inversion



musical pitch

harmony

- key and tonal context define chord's *harmonic function*
- examples:
 - **tonic:**
chord on 1st scale degree (tonal center)
 - **dominant:**
chord on 5th scale degree (often moves to tonic)
 - **subdominant:**
chord on 4th scale degree
 - ...

chord detection

introduction: key vs. chord detection

- **commonalities**

- chords are octave independent \Rightarrow pitch chroma sufficient
- process flow: pitch chroma extraction + classification

- **differences**

- time frame for pitch chroma calculation
- templates
- number of templates/chords
- many results per song (time series)

chord detection

introduction: key vs. chord detection

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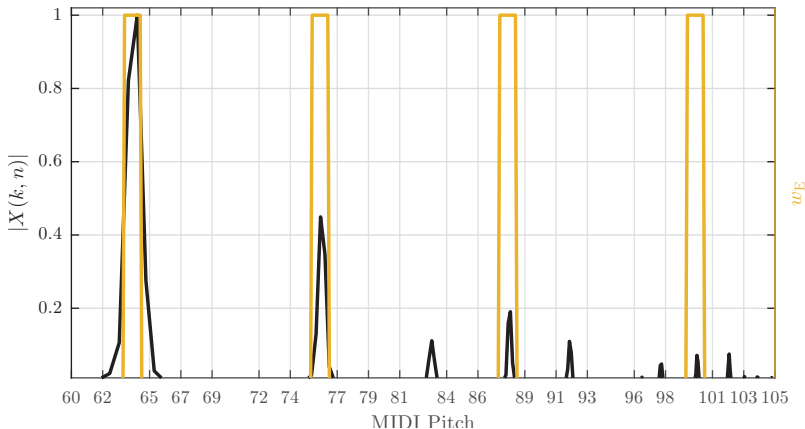
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pitch chroma

introduction

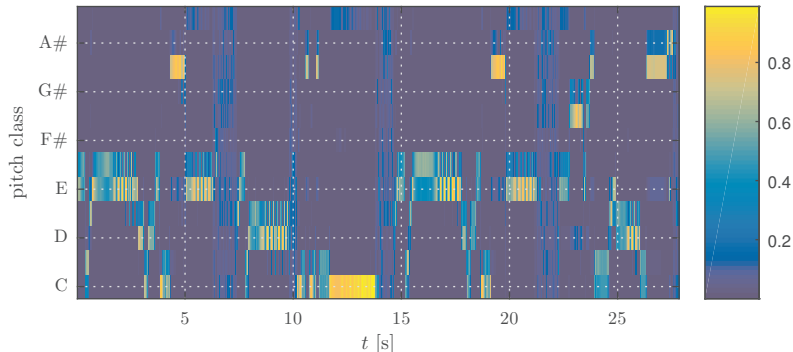
- pitch class distribution: 12-dimensional vector
- map all pitch class bands in all octaves to one



pitch chroma

introduction

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pitch chroma

introduction

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pitch chroma properties

- **no** octave information
 - no differentiation between prime and octave
 - no info on inversion
- robust, timbre-independent representation



chord detection

chord template

- similar to key detection we can simply compare an extracted pitch chroma with a template
 - simplest possible template and distance: linear transformation
example — C major: $\Gamma(0, j) = [1/3, 0, 0, 0, 1/3, 0, 0, 1/3, 0, 0, 0, 0]$
- ⇒ instantaneous chord likelihood:

$$\psi(c, n) = \sum_{j=0}^{11} \Gamma(c, j) \cdot \nu(j, n)$$

chord detection

chord progression 1/2

apply **musical knowledge** to increase the result's robustness and accuracy:

- different probabilities for different chord progressions (similar to key modulations), e.g.
 - cadences: I-IV-V-I
 - sequences: circle progression

⇒ model for *chord progression probabilities*

- ① *analytical model* based on music theory
 - circle of fifths (?!)
 - key profile correlation (?!)
- ② *empirical model* based on data
 - annotate audio
 - symbolic score

chord detection

chord progression 1/2

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chord detection

chord progression 1/2

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chord detection

chord progression 2/2

what properties do chord progression probabilities depend on



chord detection

chord progression 2/2

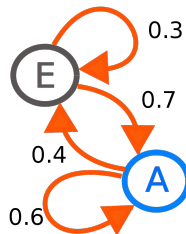
what properties do chord progression probabilities depend on



- musical key
- larger musical context (model order)
- style
- tempo/length??

chord detection

markov chain



- two possible states E, A
- transition probabilities to other state(s) and to self
- sum of transition probabilities equals 1

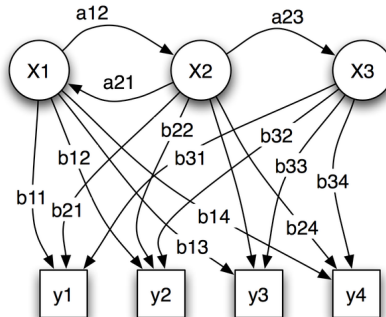
chord detection

hidden markov model: variables

- **states:**
unknown/hidden
- **transition probability:**
probability of transitioning from one state to the other
- **observations:**
measureable time series
- **emission probability:**
probability of an observation given a state
- **start probability:**
probability of the initial state

chord detection

hidden markov model: variables



- X: states
- y: possible observations
- a: state transition probabilities
- b: emission probabilities

chord detection

hidden markov model: example (WP) 1/2

- **scenario**

- doctor diagnoses fever by how patients feel
- patient may feel normal, dizzy, or cold
- patient visits multiple days in a row

what are the states and observations in this case



chord detection

hidden markov model: example (WP) 1/2

- **scenario**

- doctor diagnoses fever by how patients feel
- patient may feel normal, dizzy, or cold
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what are the states and observations in this case



- **states**

- *healthy*
- *fever*

- **observations:**

- *normal*
- *cold*
- *dizzy*

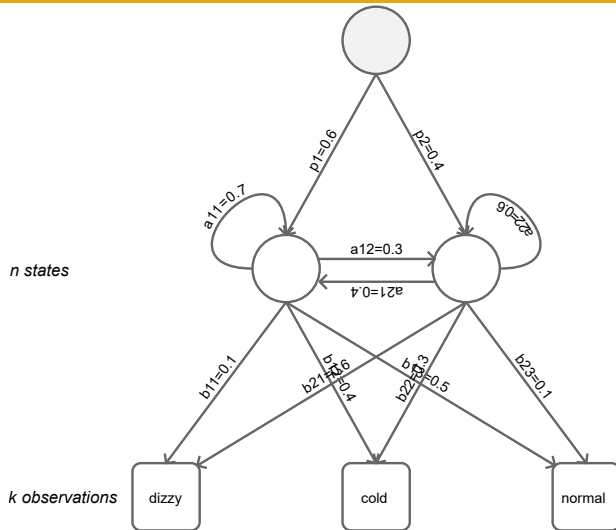
chord detection

hidden markov model: example (WP) 2/2

- **start probabilities** (initial state assumption)
 - *healthy*: 0.6
 - *fever*: 0.4
- **emission probabilities** (prob of obs given state)
 - *healthy*: normal 0.5, cold 0.4, dizzy 0.1
 - *fever*: : normal 0.1, cold 0.3, dizzy 0.6
- **transition probabilities**
 - *healthy*: healthy 0.7, fever 0.3
 - *fever*: : healthy 0.4, fever 0.6

chord detection

hidden markov model: example (WP) 2/2



start probabilities $\sum_n p_n = 1$

transition probabilities $\sum_n a_{1,n} = 1$

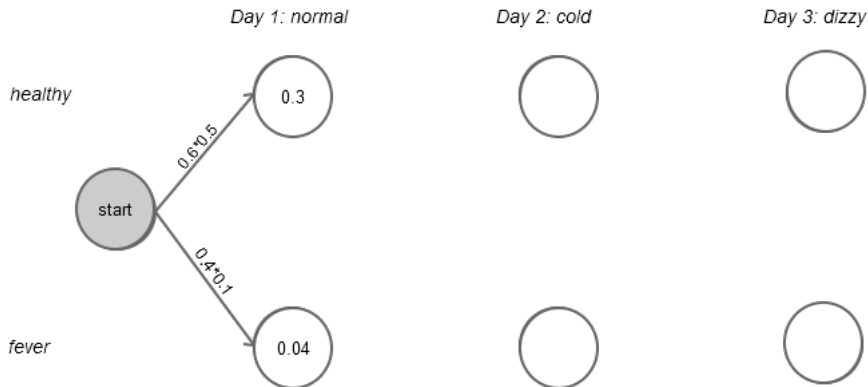
emission probabilities $\sum_k b_{1,k} = 1$

chord detection

hidden markov model: example (WP) 2/2

three observations:

Day 1 *normal* → Day 2 *cold* → Day 3 *dizzy*

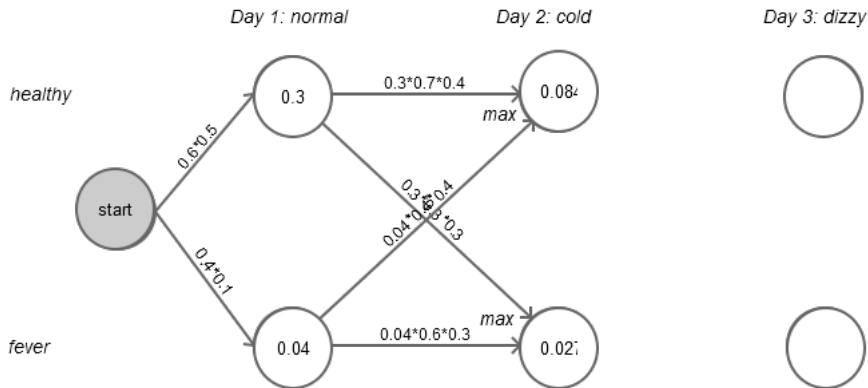


chord detection

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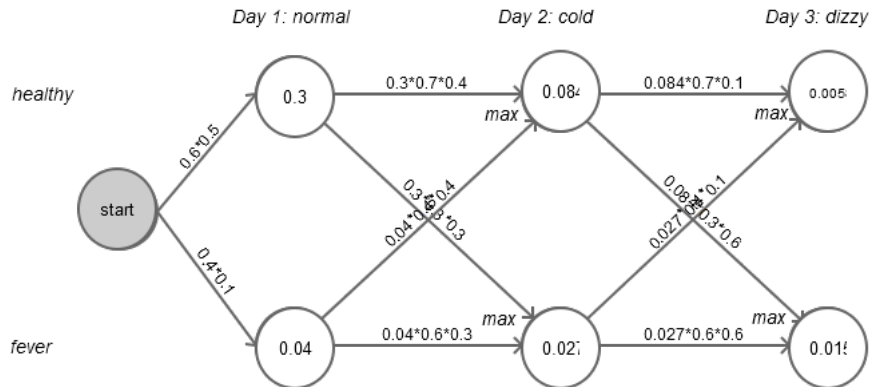


chord detection

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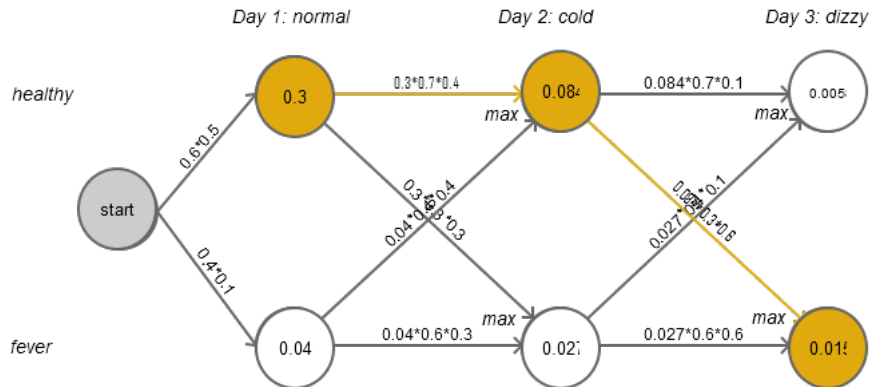


chord detection

hidden markov model: example (WP) 2/2

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chord detection

HMMs for chord detection

- states → chords
- observations → pitch chroma
- emission probability → trained with pitch chroma
- transition probability → trained from dataset
- start probability → chord statistics (style dependent?)

summary

lecture content

- **chords**
 - combination of three or more pitches
 - usually stacked thirds
 - can be inverted
- **chord detection**
 - processing steps
 - pitch chroma extraction
 - template matching
 - chord transition model
- **Viterbi algorithm**
 - find globally optimal path through state space
 - estimate state sequence with
 - emission probabilities
 - transition probabilities

