### . I AN I t pp. .t. t. r pot yp Ip they y T G Story I at ony F F 1 pp p-1. 1-1. r r " shy fr r r f r r r r r r. Mode Classification and Natural Units in Plainchant Cornelissen, Zuidema & Burgoyne

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um er gloriam plebis tue il ra hel Hunc. &

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We study 3 approaches to mode classification in medieval **plainchant** using 20.000 melodies from the Cantus database.

# 1. Modes

To determine the mode of a chant:



Plainchant uses 8 modes, grouped in 4 pairs.

story

vine on us

1

11

A

Modes in a pair use the same **final** note, but have a different range.

Authentic modes move mostly above the final, Plagal modes descend further and move around the final.

# 2. Overview

#### **A Melodic transcriptions in Cantus**



The Cantus database contains over **60,000** transcribed **melodies**, with mode annotations.

We use **two genres**: 7,000 responsories and 13,000 antiphons.

We study 3 approaches to mode classification.

The distributional approach represents chants as vectors of weighted motif frequencies.

We compare different motifs or units: natural units group notes that form neumes, syllables or words. **Baselines** include *n*-grams.

We compare pitch, intervalic and contour representations.

We use the **tf-idf** weighting from textual information retrieval.

It corrects overly common terms (units) in favour of more specific terms, frequent in fewer documents (chants).

# **3. Features**

#### **A** Classical features





#### C tf-idf vectors



#### Modes are clearly distinguished by both the classical features and the pitch profiles.

The distributional approach aims to capture the melodic aspect of mode, even without pitches.

And as we move to interval or contour representations, the modes are less clearly distinguished.

Shown are PCA embeddings of the chant vectors, coloured by mode.

# 4. Results

#### A Classical approach

#### **C** Distributional approach

| final                | 40        |
|----------------------|-----------|
| range                | 56        |
| initial              | 37        |
| final & range        | 89        |
| final, range & init. | <u>90</u> |

### **B** Profile approach

| pitch class profile | 85        |
|---------------------|-----------|
| pitch profile       | <u>88</u> |
| repetition profile  | 81        |

| 0% | Weighted F1-score | 100% |
|----|-------------------|------|

| units    | pitch     | interval  | contour   |
|----------|-----------|-----------|-----------|
| neume    | 92        | 79        | 52        |
| syllable | <u>93</u> | 86        | 76        |
| word     | 90        | <u>86</u> | <u>81</u> |
| 1-gram   | 87        | 7         | 7         |
| 2-gram   | 91        | 38        | 17        |
| 3-gram   | 92        | 65        | 23        |
| 4-gram   | 91        | 75        | 34        |
| 5-gram   | 91        | 81        | 43        |
| 6-gram   | 88        | 82        | 51        |
| 8-gram   | 82        | 78        | 60        |
| 10-gram  | 76        | 74        | 66        |
| 12-gram  | 71        | 69        | 65        |

representation ----->

segmentation

### **Results**

- All three approaches work well
- Natural units perform better than *n*-gram baselines, and perform well even without pitch information

### What does this show?

- Contours alone are very informative.
- Perhaps: is (this) music built from units, just like a sentence is composed of words?

**Code:** github.com/bacor/ISMIR2020 **CantusCorpus:** github.com/bacor/CantusCorpus Chant21 Python package: github.com/bacor/chant21 *Volpiano & gabc in music21; see our DLfM paper.*