

# Discovering the line of fifths in a large historical corpus

Future Directions of Music Cognition Virtual Conference

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# Corpus Studies – Music Theory – Music Analysis

A sketch of the corpus studies cycle:

1. analyze a piece or some salient aspect
2. compare it to other pieces and wonder whether observation holds more generally
3. gather/create suitable corpus to study question on a large scale
4. operationalize question for quantitative analysis
5. test against data
6. usually: new (general) observations
7. do they apply to individual pieces?
8. go back to 1.

# Different realizations of hexachords

Comparing tonal material



Figure 1: Josquin Desprez, *Missa sine nomine*, *Agnus Dei II* (1514).

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## Comparing tonal material



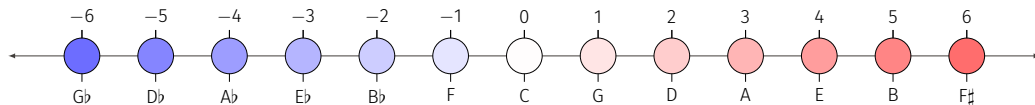
Figure 1: Josquin Desprez, *Missa sine nomine*, *Agnus Dei II* (1514).



Figure 2: Chick Corea, *Children's Song No. 2* (1984), mm. 1–4.

# The Line of Fifths

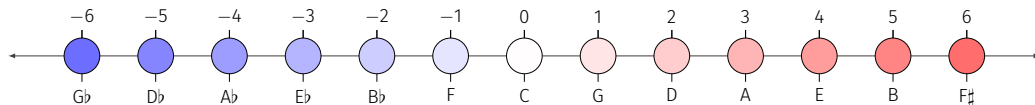
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**Figure 3:** The *Line of Fifths*. Color mapping emphasizes flat- and sharpwards directions.

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**Figure 3:** The *Line of Fifths*. Color mapping emphasizes flat- and sharpwards directions.

The LOF contains many common scales as **contiguous subsegments**, e.g.

- diatonic
- pentatonic
- chromatic
- hexachord (Desprez, Corea)
- ...

## Research question

If the line of fifths (and its subsegments) are central for tonal organization in many pieces, we should be able to see its relevance in a corpus study.

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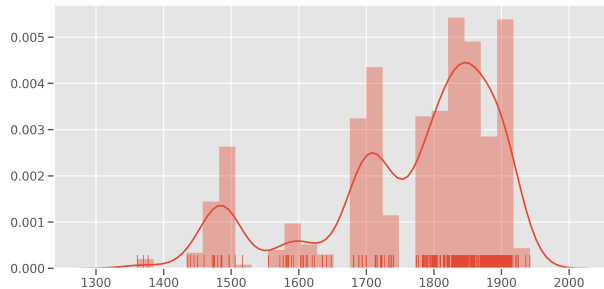
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### Operationalization

- vector-space model: a piece is represented by the distribution of its pitch classes
- representation of notes as tonal pitch-classes (e.g. F $\sharp$ , G $\flat$ , A $\sharp\sharp$ , ...)
- restriction from F $\flat\flat$  to B $\sharp\sharp$  (35 dimensions)
- important: no inherent ordering of pitch classes



# The corpus



**Figure 4:** Convenience sample of 2012 pieces by 75 composers, 1361–1942 (Moss, Neuwirth, and Rohrmeier, 2020).

## Principal Components Analysis (PCA)

- We use PCA for dimensionality reduction.
- Principal components are dimensions that maximize variance in the data.
- Inspecting principle components thus allows to draw conclusions about which factors affect the data distribution most.

# Method

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- Principal components are dimensions that **maximize variance** in the data.
- Inspecting principle components thus allows to draw conclusions about which **factors** affect the data distribution most.

## Procedure

- corpus 'lives' in 35-D space, all dimensions are assumed to be independent
- we inspect the first two principal components, explaining 64% of data variance
- each data point is colored according to its most frequent pitch class

# Structure Discovery with Dimensionality Reduction

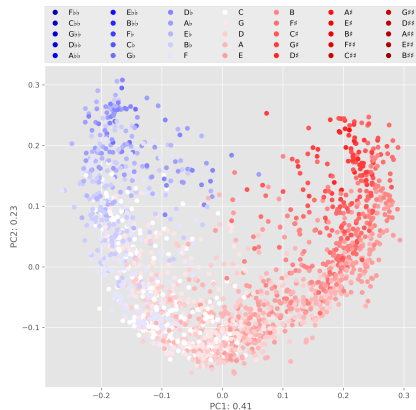


Figure 5: PCA, global view.

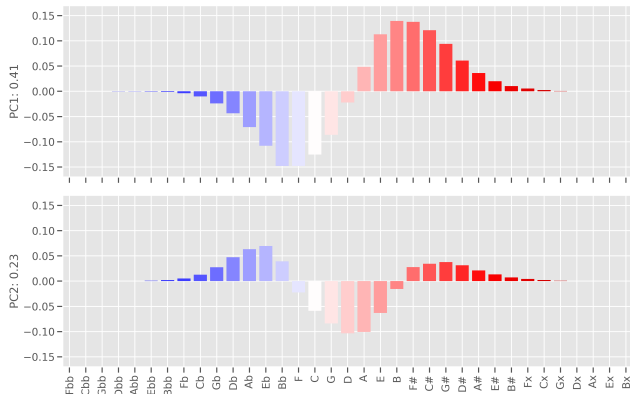
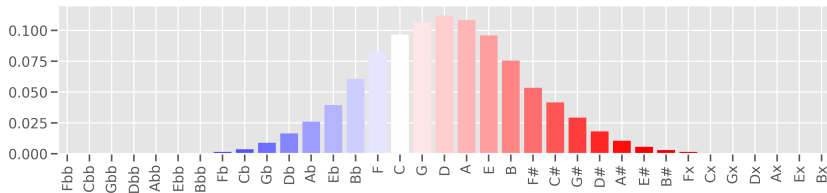


Figure 6: PCA, first two components: distance to center D (top) and 'accidental regions' (bottom).

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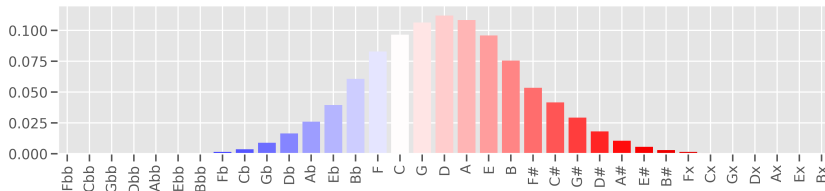


**Figure 7:** Average tonal pitch-class distribution in the corpus.

Observations:

- almost normally distributed
- centered on D (mid-point on line of fifths)

# Structure Discovery with Dimensionality Reduction



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This is a global statistic. What does it look like under a **historical** perspective?

# Pitch-Class Co-Evolution

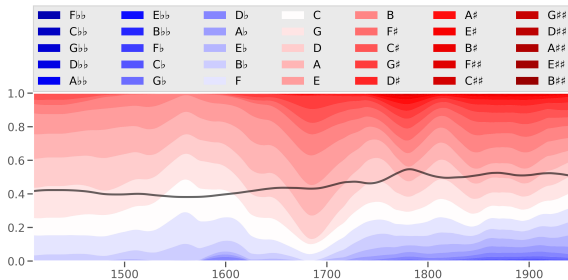


Figure 8: Relative pitch-class frequencies over time.

- **local variations** possibly due to specific sample (corpus)
- amount pitch-classes with one or several **accidentals** increases
- **entropy** remains relatively stable
- relative frequencies appear to be **correlated**

# Pitch-Class Co-Evolution

- high correlations around the central axis

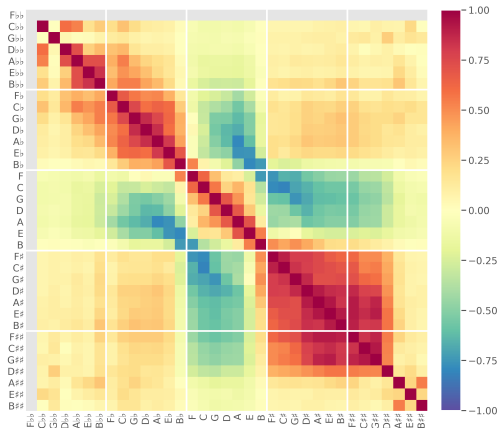


Figure 9: PC co-evolution across historical periods. 10/12



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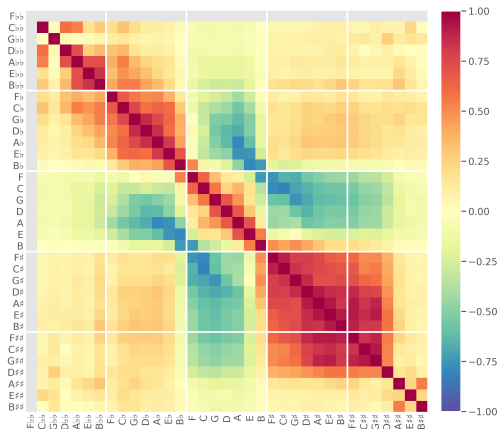


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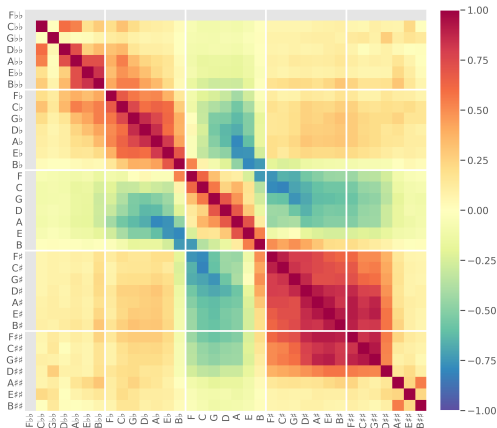


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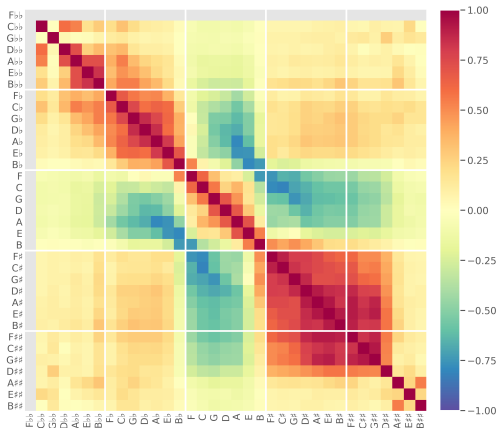


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→ pieces further away from the LOF center are generally more chromatic

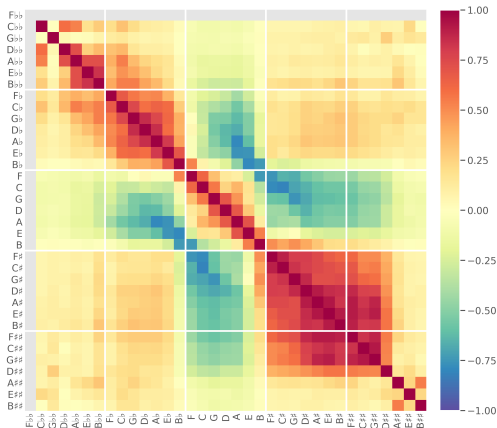


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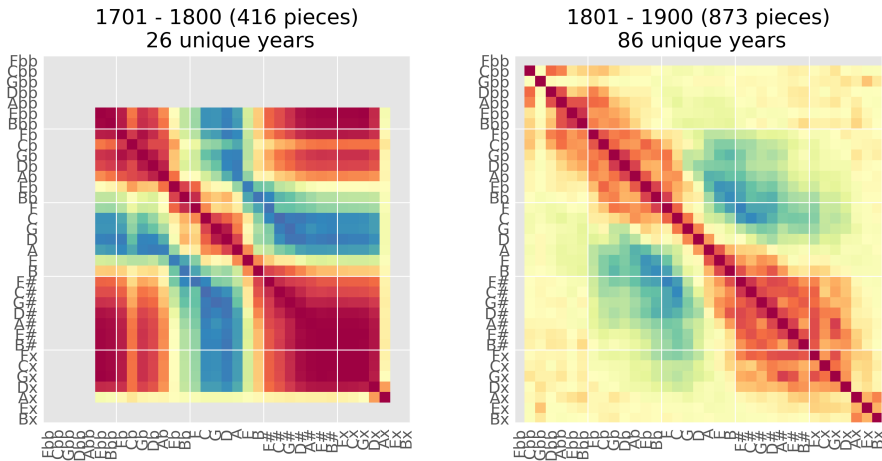


Figure 10: PC co-evolution in the 18th (left) and 19th (right) centuries.

# Conclusions

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## Future work

- expanding the corpus (also beyond canon)
- taking other parameters into account (harmony, meter, form...)
- zooming on from the global view

Thank you for your attention!

Slides available at [fabian-moss.de](http://fabian-moss.de)

## References

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- Moss, F. C., Neuwirth, M., & Rohrmeier, M. (2020). Tonal Pitch-Class Counts Corpus (TP3C) (Version v1.0.0). Zenodo. <https://doi.org/10.5281/zenodo.3600088>
- Temperley, D. (2000). The line of fifths. *Music Analysis*, 19(3), 289–319.  
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