

# Automatic interpretation of music structure analyses

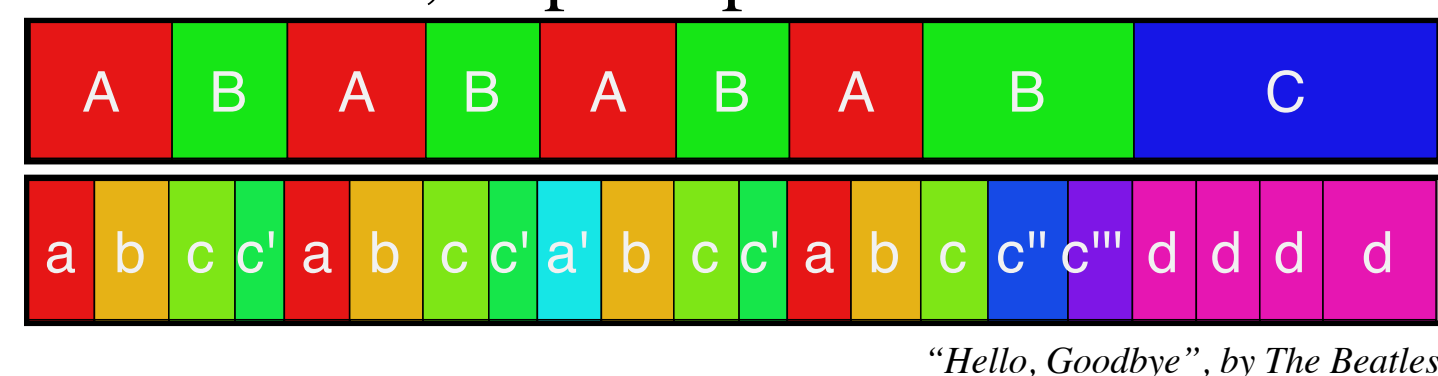
A validated technique for post-hoc estimation of the rationale for an annotation

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# 1. Motivation

Structural descriptions are usually single-dimensional, or perhaps hierarchical:



This annotation tells us that sections **A** and **B** are different—but what makes them different? Do listeners think **B** is defined by a harmonic or melodic progression, or by a timbre? What was the listener's **rationale**?

Collecting this information from listeners is onerous, and the introspection required is difficult. Instead, we aim to **automatically interpret existing annotations** by comparing them to the audio.

If successful, we could visualize structure to see which musical attributes characterize each section:



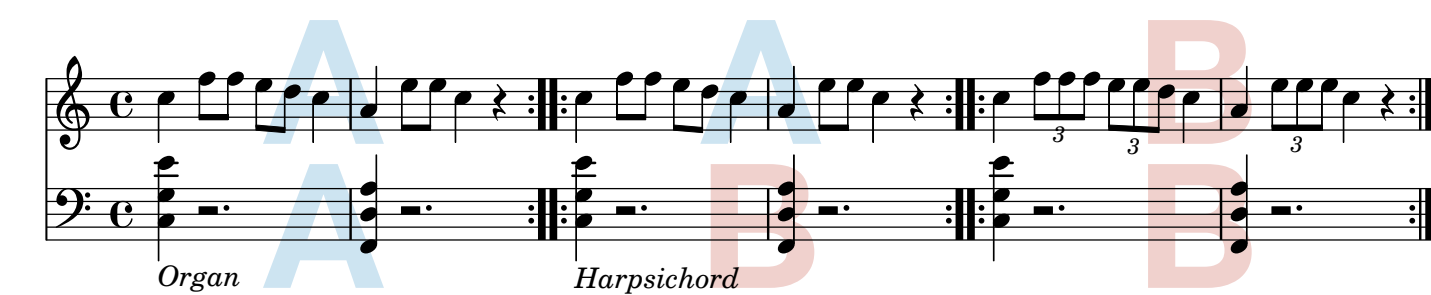
**How to read:** cells are brightest when a feature is:

1. homogenous throughout that section;
2. similar in other sections with the same label;
3. different in other sections with different labels.

## 2. Data

Finding appropriate data is not trivial! To validate the algorithm, we need structural annotations **paired** with listener rationales.

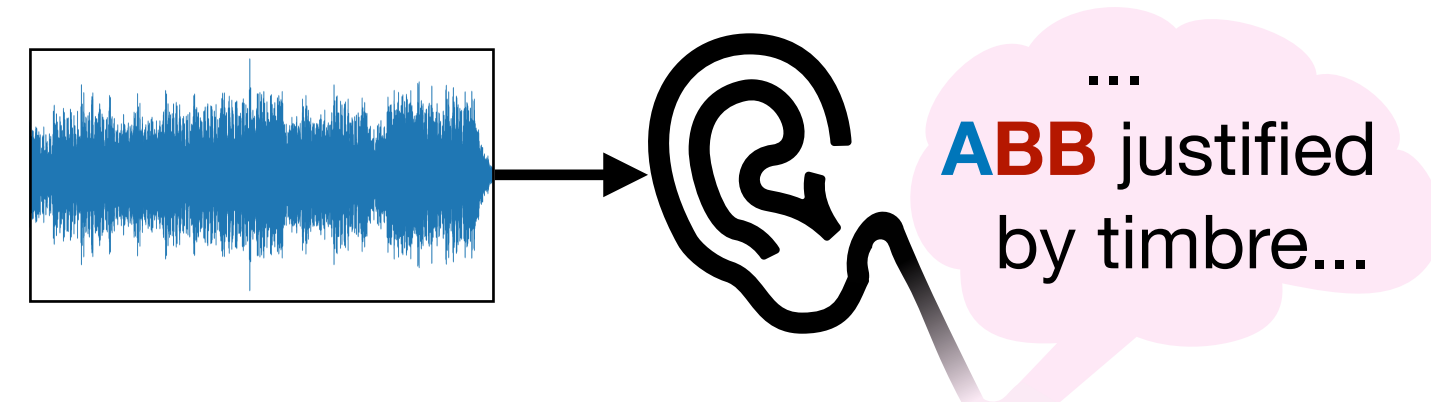
We obtained the data in a music perception study: we composed stimuli with *intended* forms, each suited to *intended* rationales:



**AAB** justified by rhythm

**ABB** justified by timbre

We also confirmed that listeners perceived these structure with the same rationales:



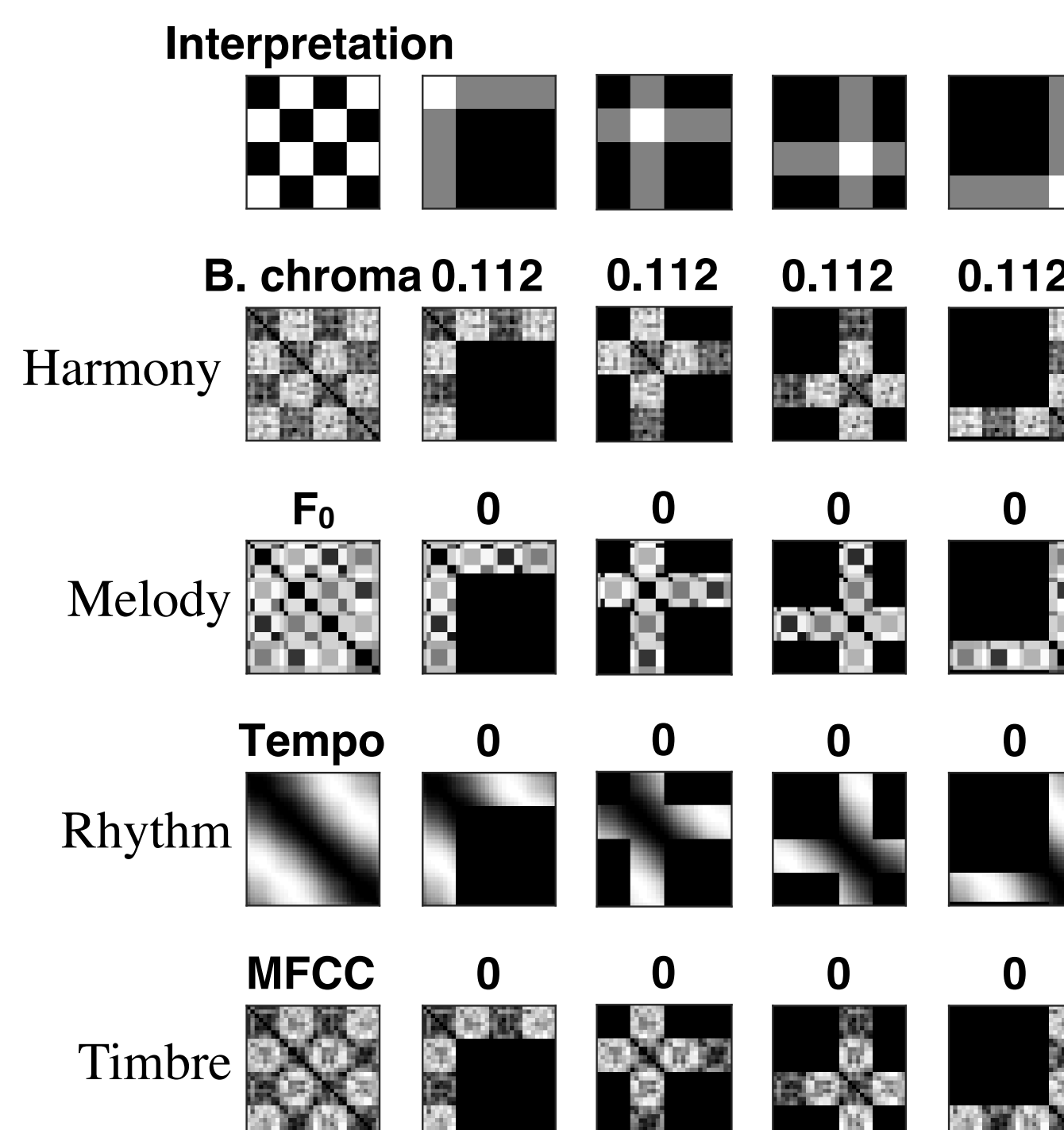
We have a large number of stimuli, in three styles, with either 3 parts (**AAB** vs. **ABB**) or 4 parts (**AABB** vs. **ABAB** vs. **ABBA**).

### 3. Algorithm

We compute self-similarity matrices (SSMs) from several **audio features**, each of which is *assumed* to correlate with a relevant **musical attribute**.

We generate **masked SSM segments**, each revealing the relationship of a segment to the rest of the piece.

Then, a **quadratic program** (QP) estimates coefficients to recreate the ground truth SSM from the masked segments. E.g.:



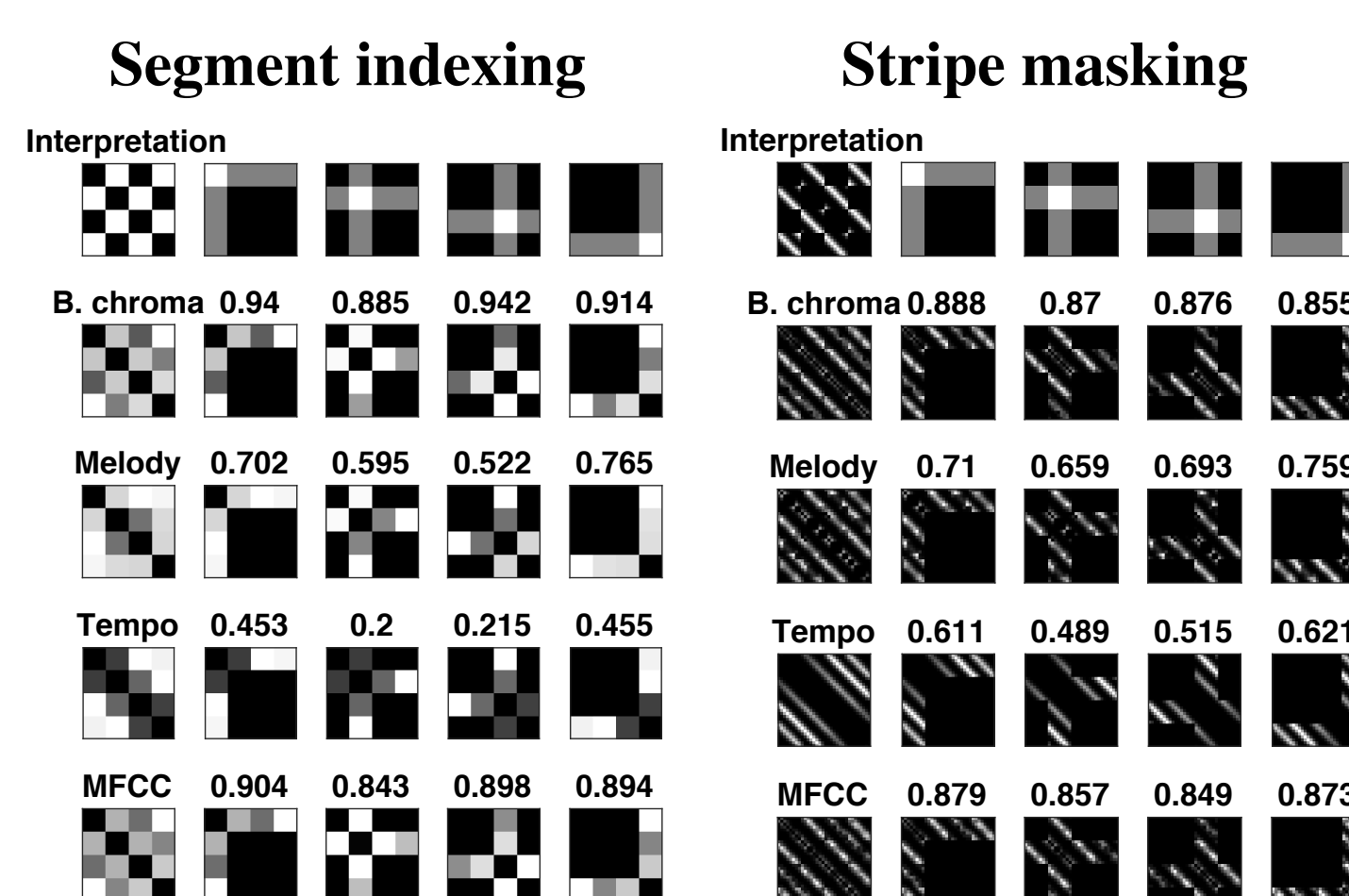
This piece has structure:

- ABBA** justified by timbre
- AABB** justified by rhythm
- ABAB** justified by harmony

The QP reconstructs the **ABAB** interpretation using only bass chroma.

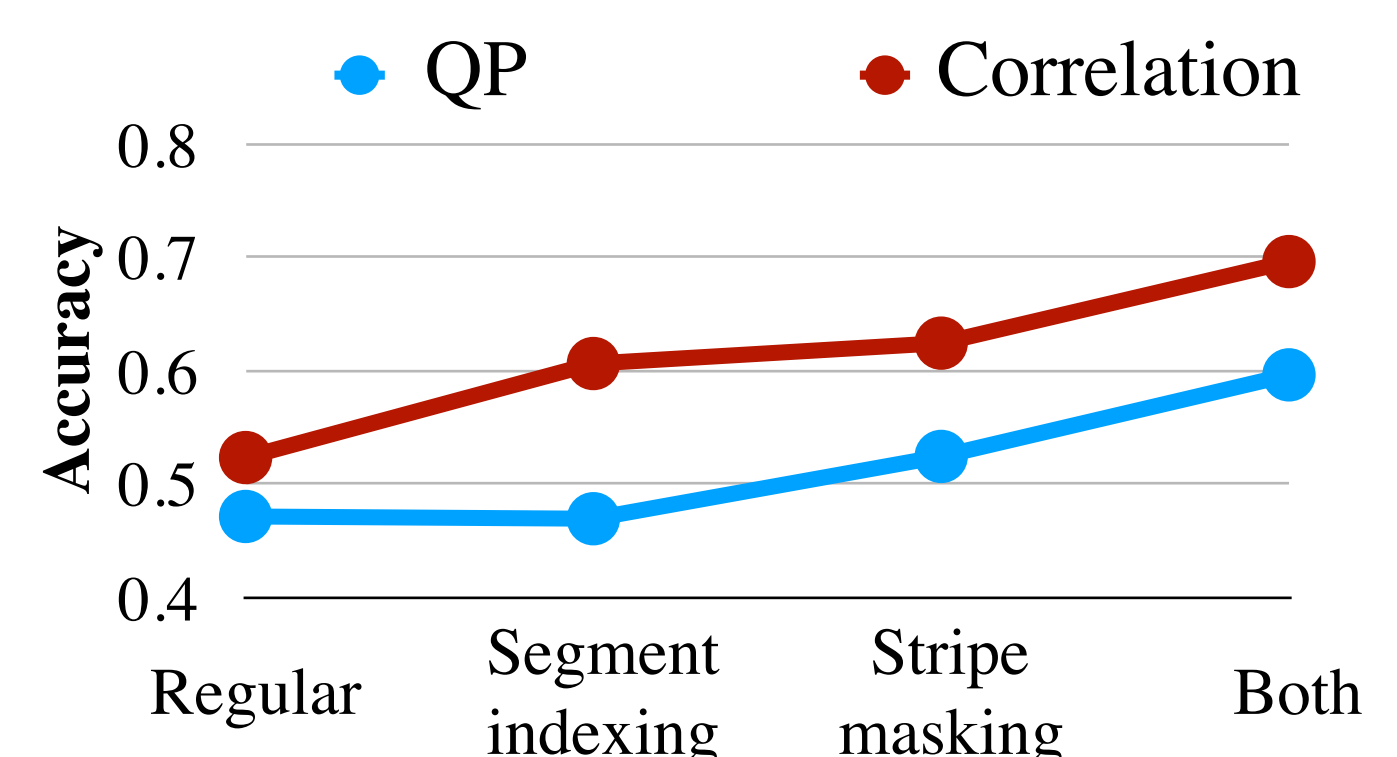
The QP approach has clear limitations:

- If two musical attributes explain a section equally, the QP might only point to one. Instead, we can measure **correlation**.
- Sequences that are repeated but non-homogenous may be overlooked in a point-wise SSM comparison. Instead, we can use **segment-indexed** SSMs, or apply additional **stripe masking**.

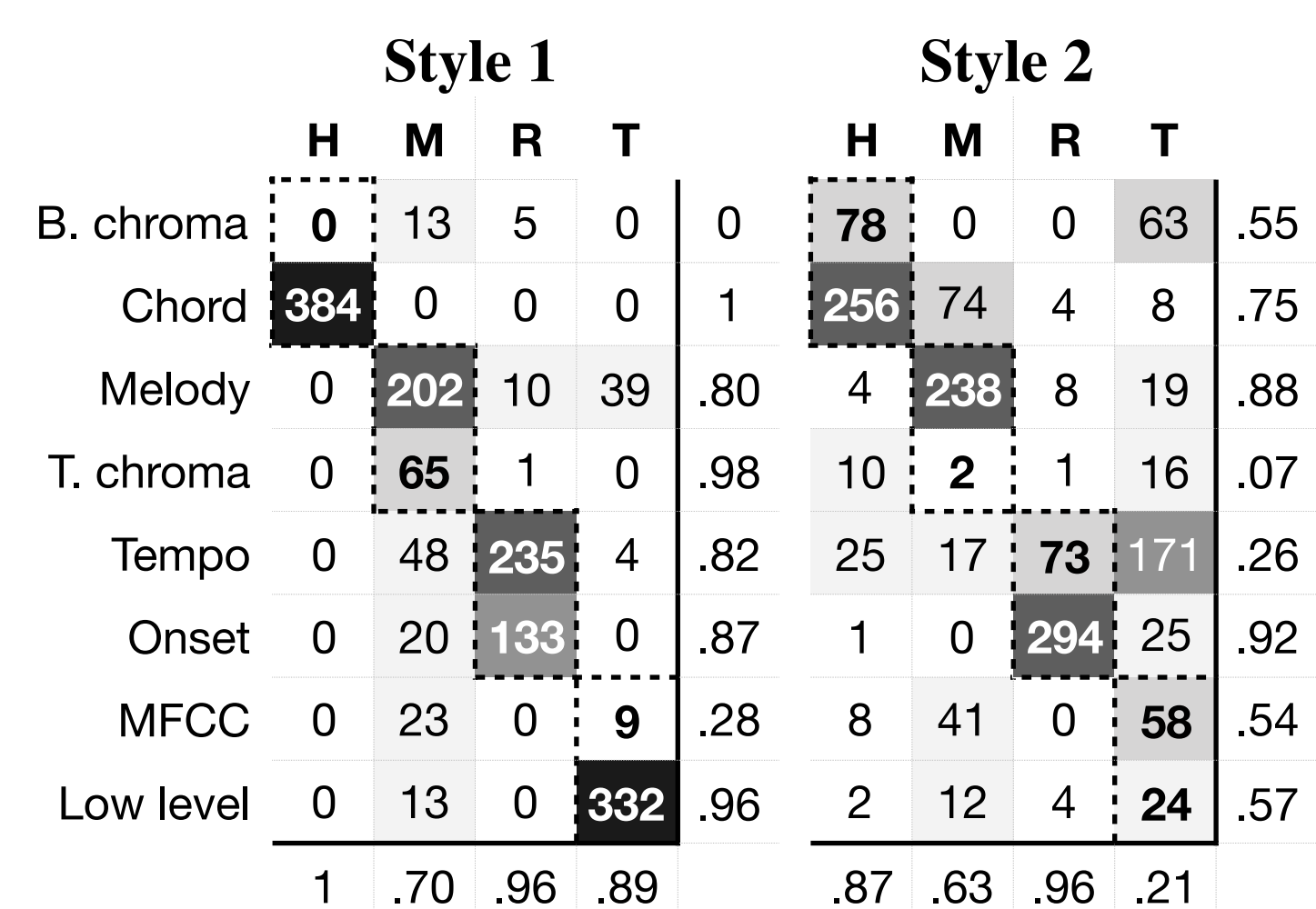


## 4. Validation

The suggested improvements all had a positive impact: the best algorithm used the stripe-masked SSMs, indexing by segment, and correlation instead of the QP output.



But accuracy varied among musical styles and features, as these confusion plots show:



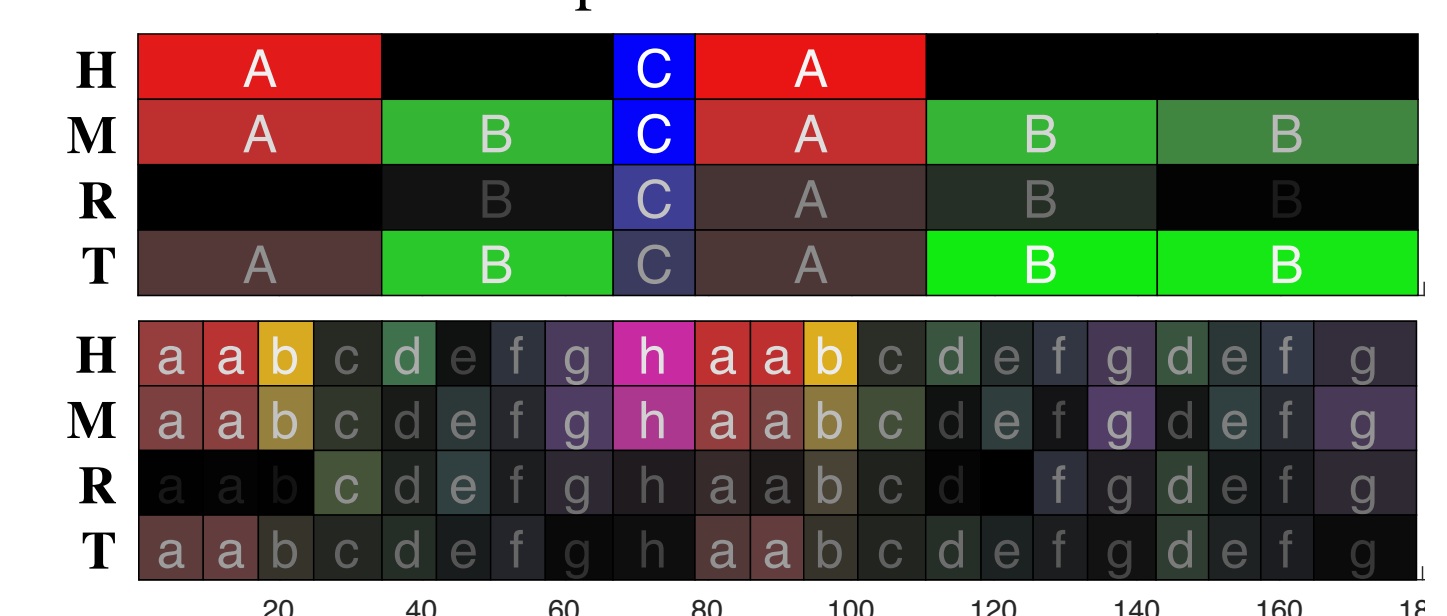
## 5. Application

We can use the validated approach to analyze SALAMI annotations:

*“We Are The Champions”, by Queen*

**A:** Harmonies stable, orchestration builds up;  
→ harmonies in **a** and **b** are unique across the piece.

**B:** Complex chord sequence, stable timbre;  
→ timbre cannot explain individuated subsections.



Some analyses have prime markers. If we consider primed sections to be similar or different changes the interpretation.

*“Another One Bites The Dust”, by Queen*

**d=d'**: Stable, stripped-down harmony throughout.

**d≠d'**: Sections feature odd, varying sound effects.

