

## Information Lattice Learning (ILL)

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#### What makes X X?

(What makes Bach's chorales Bach's chorales?)

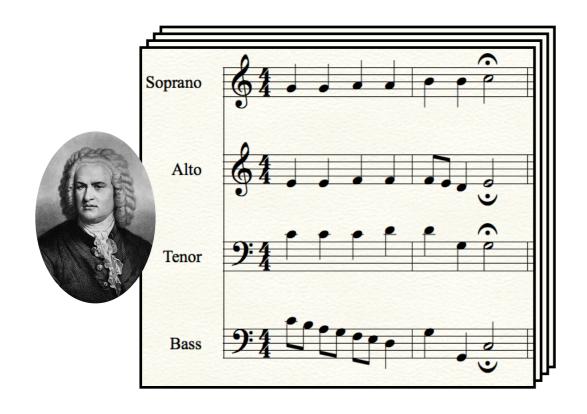
How the above question differs from ...?

# A New Learning Problem

#### **Automatic Concept Learning**

Input

Output



Concepts

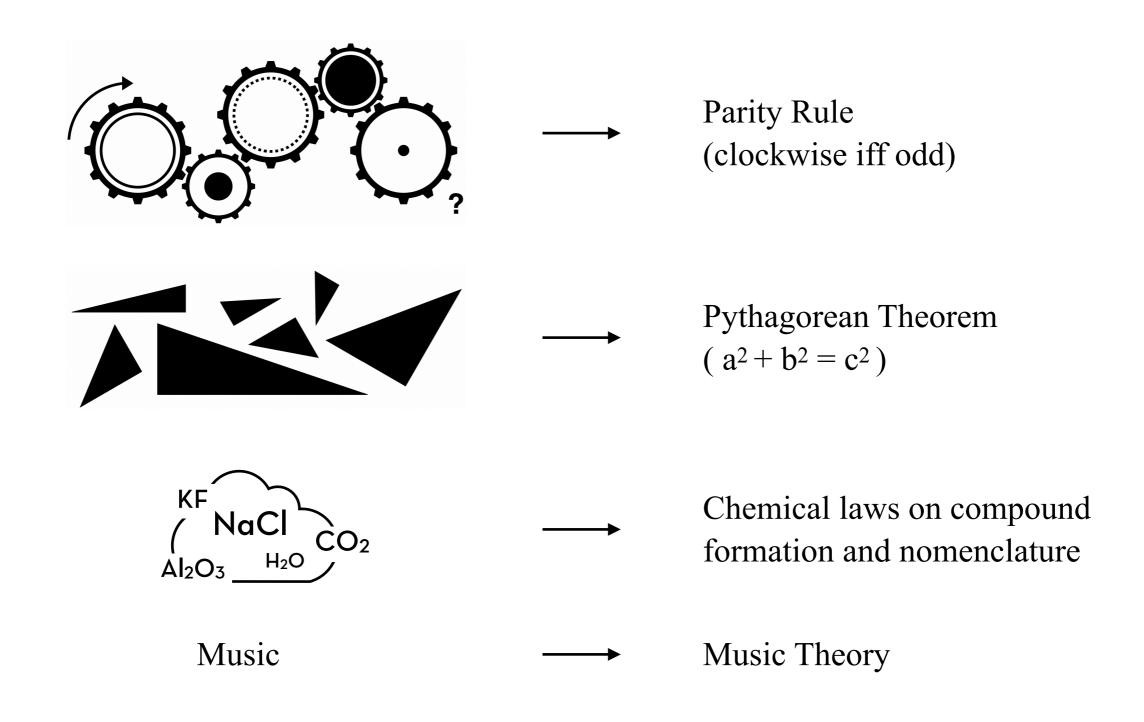
Rules

Laws

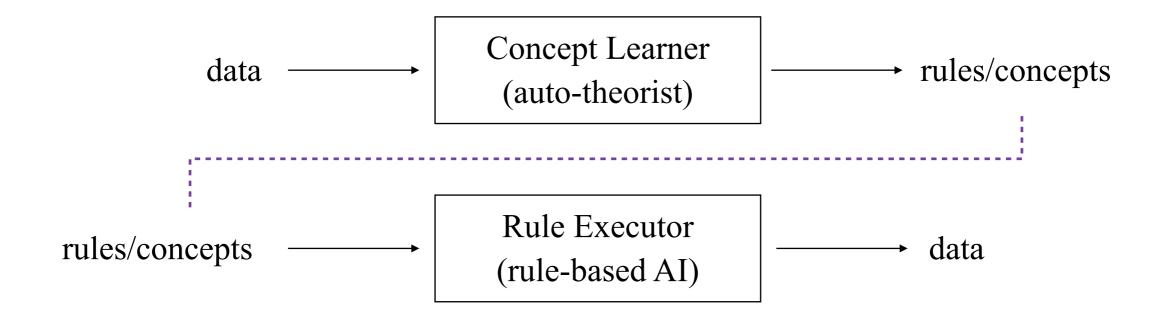
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#### Automatic Concept Learning

From phenomenology to theory



#### Concept Learning in a Nutshell



Information Lattice Learning (ILL)

data rules and concepts data' augment human intelligence augment human creativity

How this differs from an auto-encoder?

# A Learning Paradigm

- Representation: Information Lattice (IL)
- Algorithm: Information Lattice Learning (ILL)

Representation: IL

An abstraction  $\mathcal{A}$  is a partition of the data space X.

$$X = \{x_1, x_2, x_3, x_4, x_5, x_6\}$$

$$\mathcal{A} = \{\{x_1, x_6\}, \{x_3\}, \{x_2, x_4, x_5\}\}$$
cells (or less formally, clusters)
$$\vdots$$
concepts

A concept is a partition cell.

```
X = \{\text{every vertebrate}\}\

\mathcal{A} = \{\text{mammals, birds, fish, amphibians, reptiles}\}\
```

An abstraction A is a partition of the data space X.

$$X = \{x_1, x_2, x_3, x_4, x_5, x_6\}$$

$$\mathcal{A} = \{\{x_1, x_6\}, \{x_3\}, \{x_2, x_4, x_5\}\}$$
cells (or less formally, clusters)
$$\vdots$$
concepts

A concept is a partition cell.

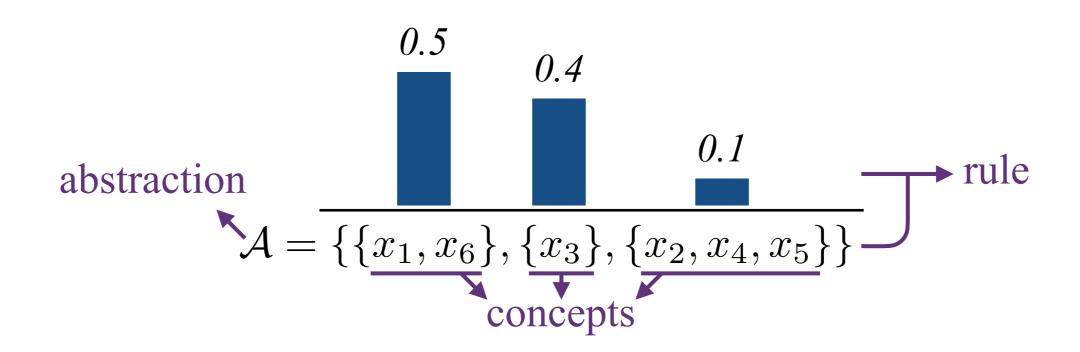
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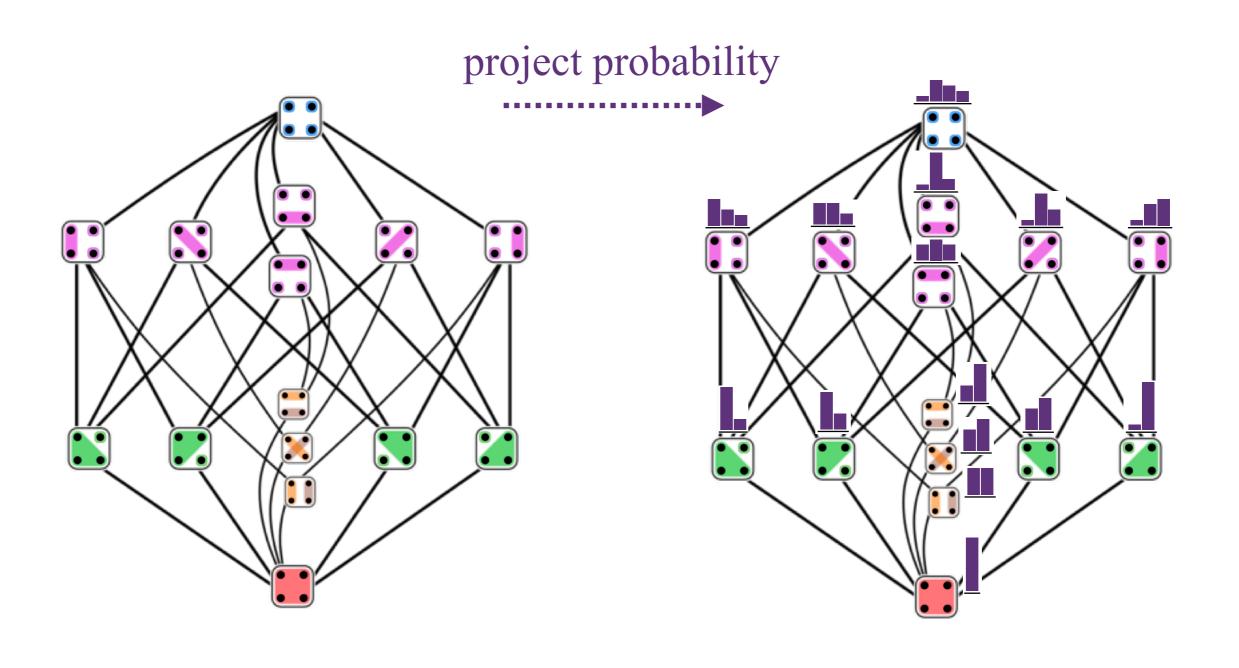
A concept is a partition cell.

A (probabilistic) rule is a probability distribution projected onto a partition

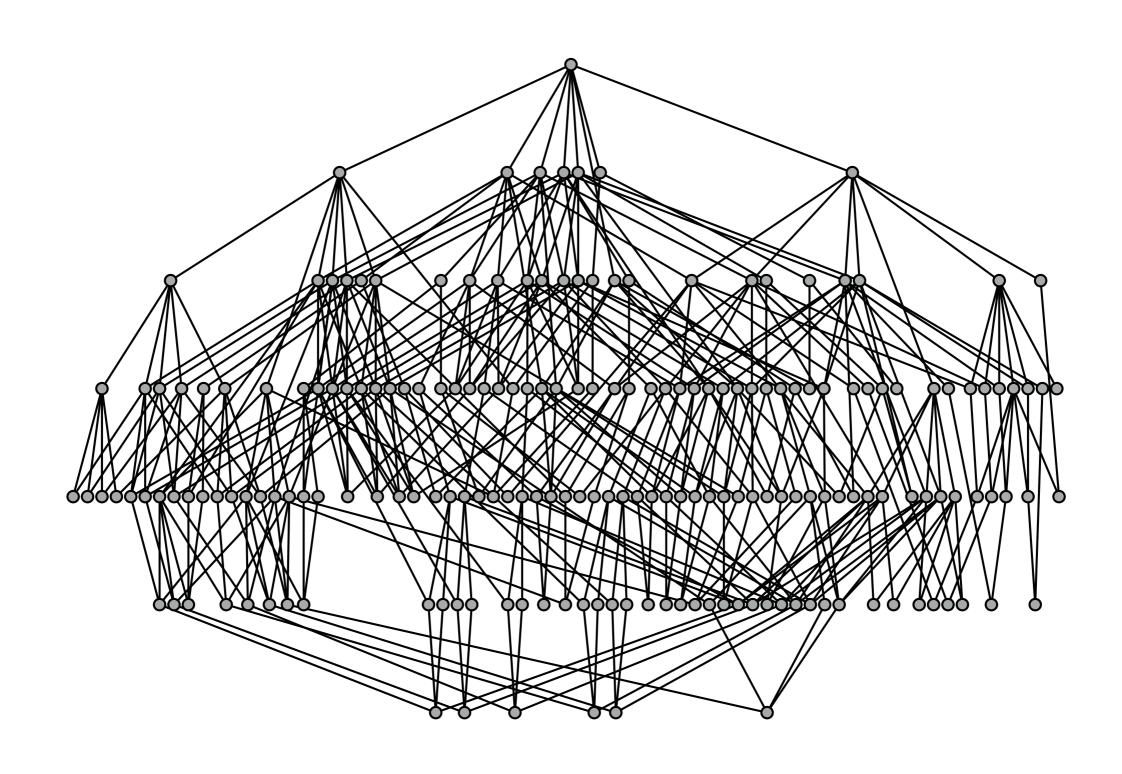


#### Partition Lattice

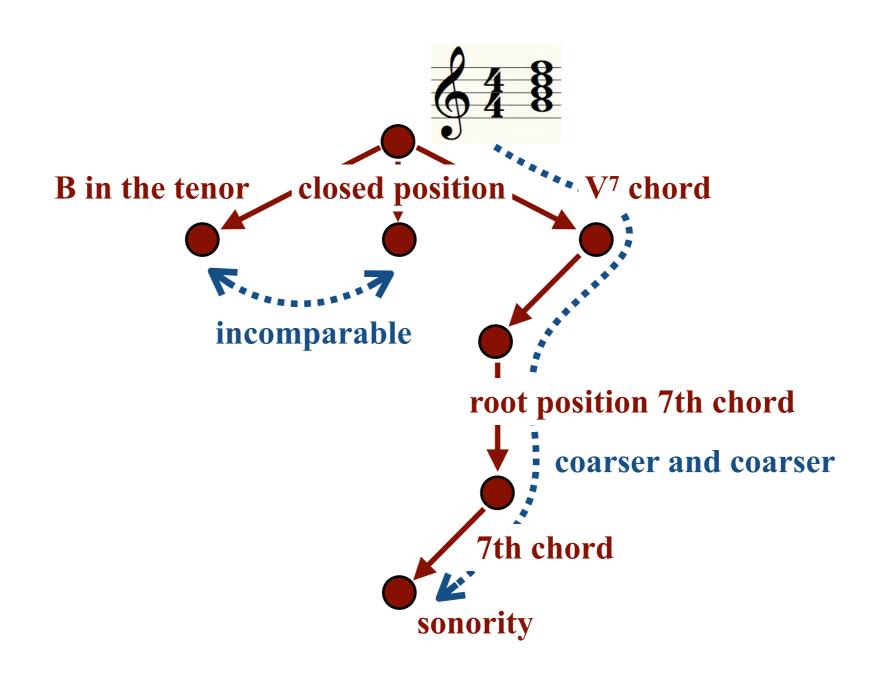
#### Information Lattice



### Bach's Information Lattice



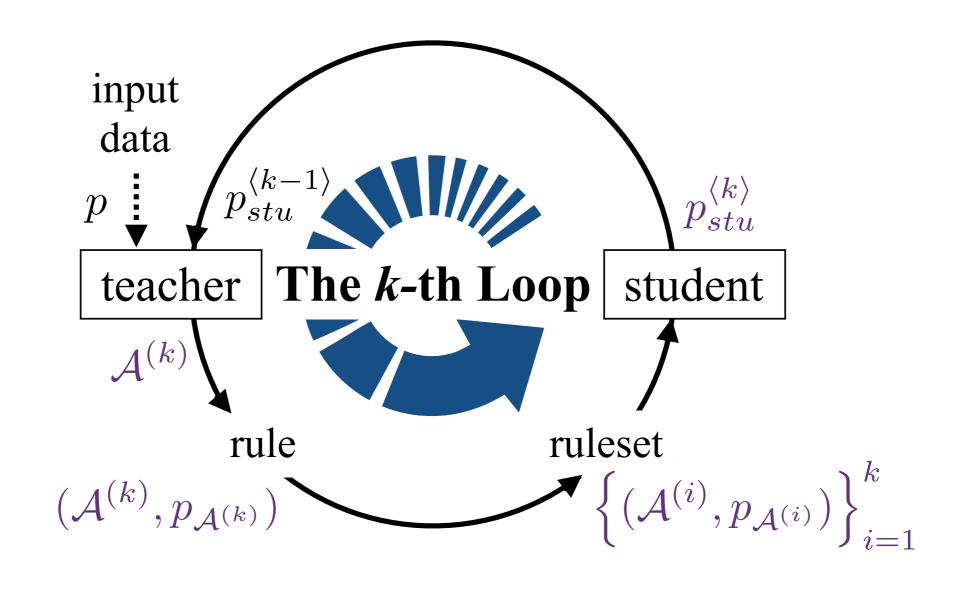
### Bach's Concept Lattice



Algorithm: ILL

### The Self-Learning Loop

A Teacher-Student Architecture: Learning by Comparison



#### Teacher: a Discriminative Model

The teacher solves an optimization problem:

maximize 
$$D_{KL}\left(p_{\mathcal{A},stu}^{\langle k-1\rangle} \parallel p_{\mathcal{A}}\right)$$
  
subject to  $\mathcal{A} \notin \mathfrak{P}^{\langle k-1\rangle}$   
 $H(p_{\mathcal{A}}) \leq \delta_k$ 

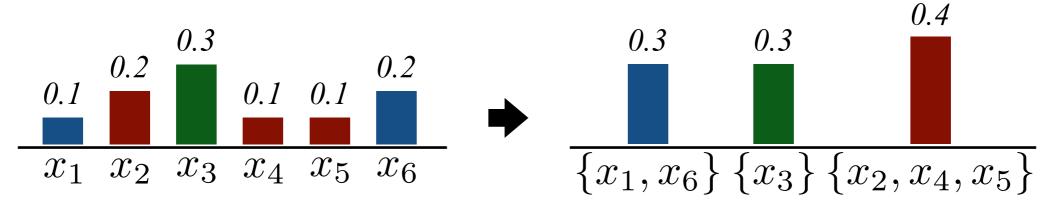
Want to find the abstraction which reveals the largest statistical difference between the student and the input data.

#### Teacher: a Discriminative Model

The teacher solves an optimization problem:

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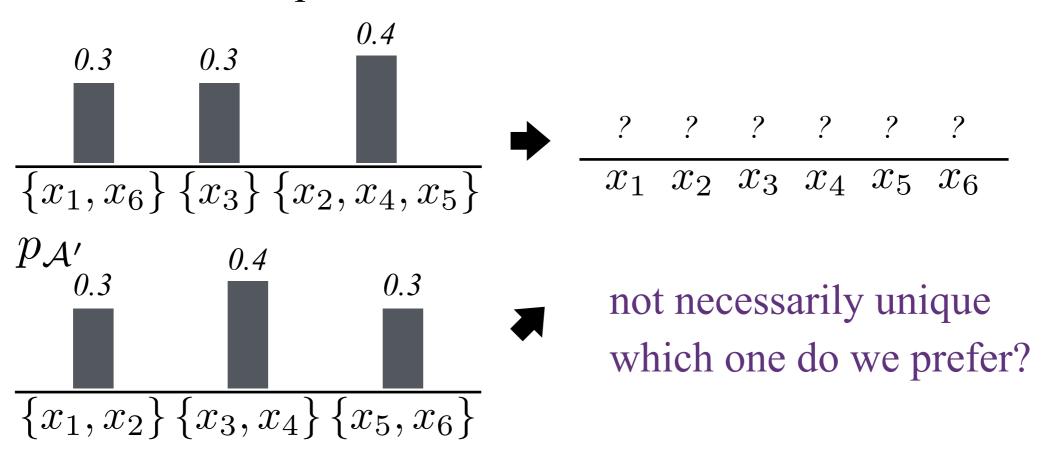
Projection: given P, computing PA is easy:



#### Student: a Generative Model

Apply probabilistic rules, which is known as the rule realization problem.

Given  $\mathcal{P}\mathcal{A}$ , compute  $\mathcal{P}$ :



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#### Student: a Generative Model

The student solves a MaxEnt problem:

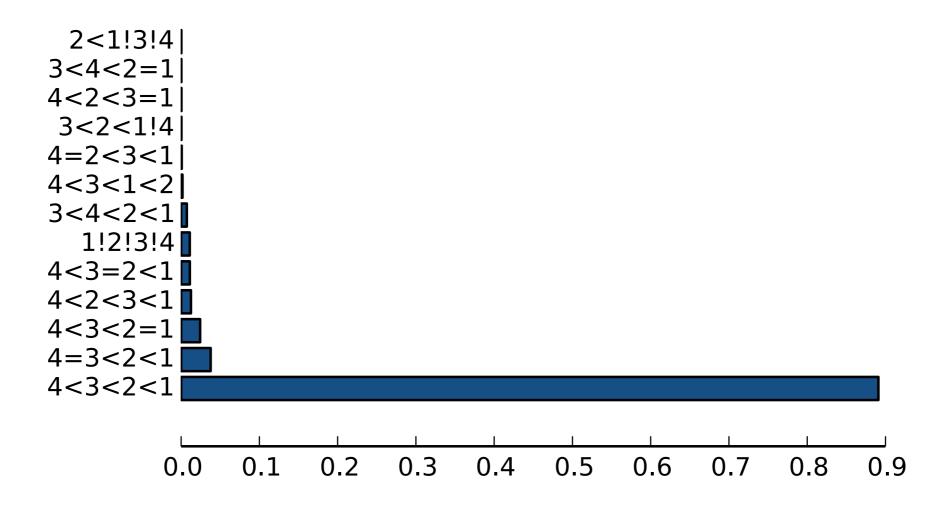
$$\begin{array}{ll}
\text{maximize} & S_q(p_{stu}^{\langle k \rangle}) := (q-1)^{-1} \left( 1 - \| p_{stu}^{\langle k \rangle} \|_q^q \right) \\
\text{subject to} & A^{(i)} p_{stu}^{\langle k \rangle} = p_{\mathcal{A}^{(i)}}, \quad i = 1, \dots, k
\end{array}$$

Want to find the most creative probabilistic model which satisfies all the rules (quality) while at the same time enables novelty

### How MUS-ROVER Self-Evolves?

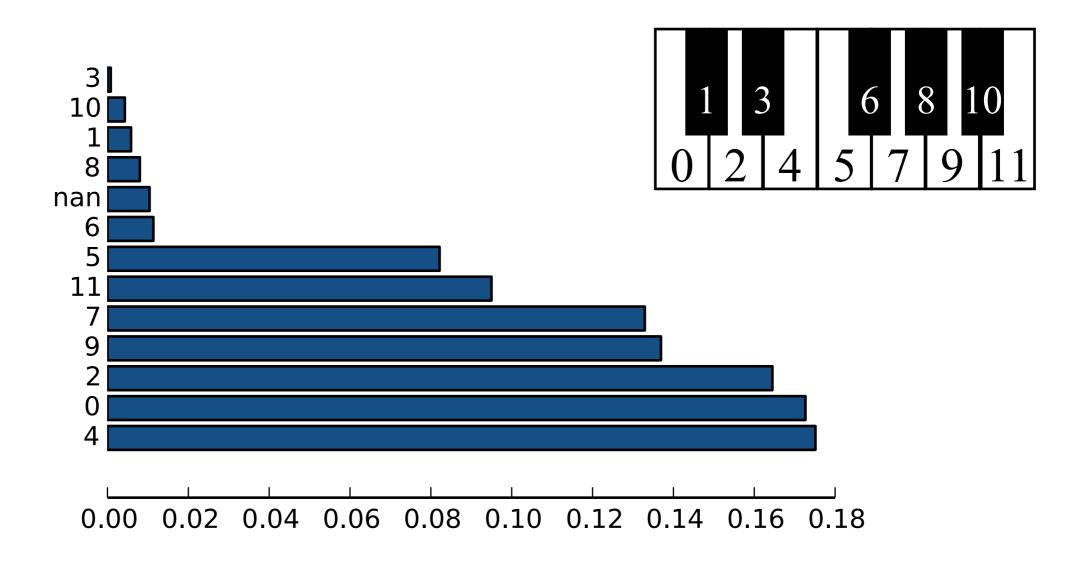


Rule 1: order  $\circ w_{1,2,3,4}$ 





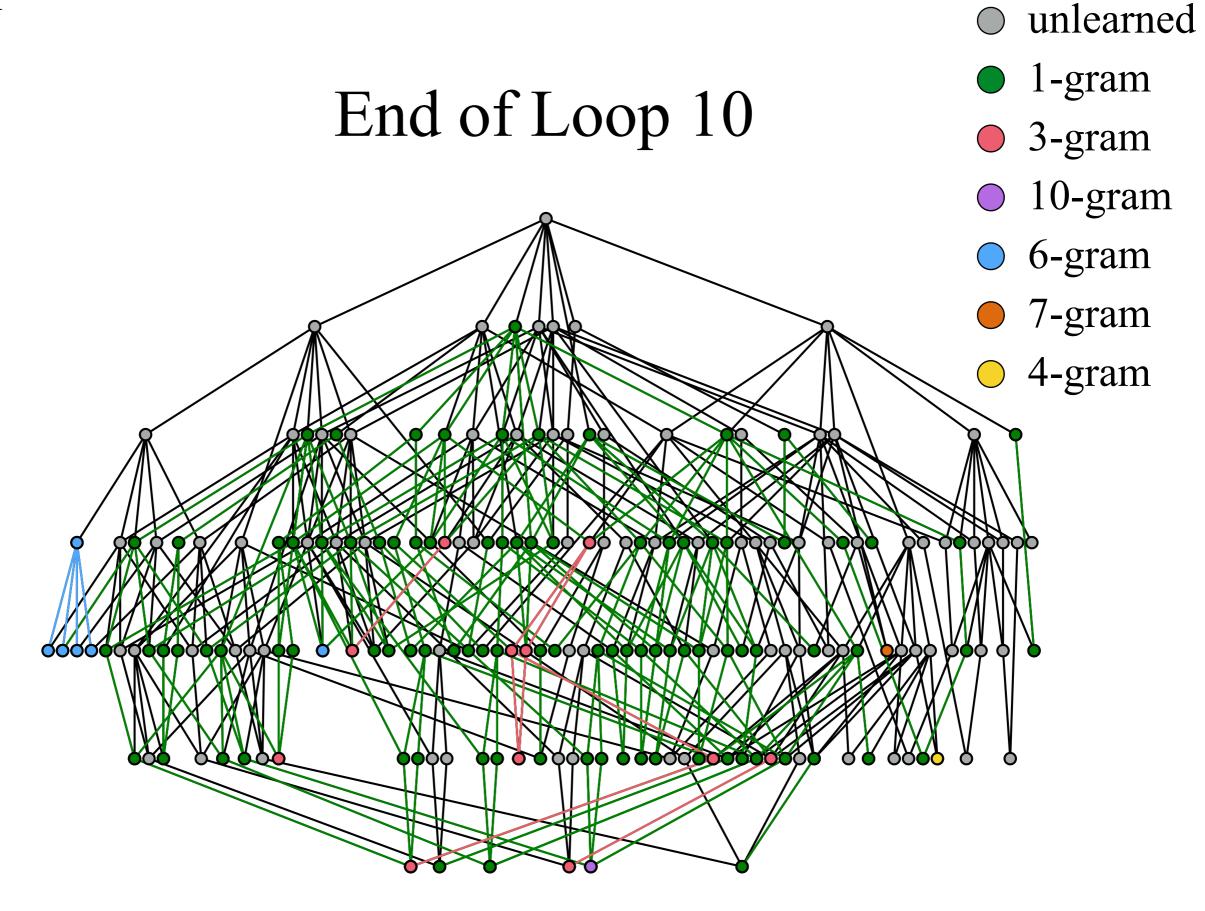
Rule 2:  $mod_{12} \circ w_1$ 







n-gram



# Music Mosaic

