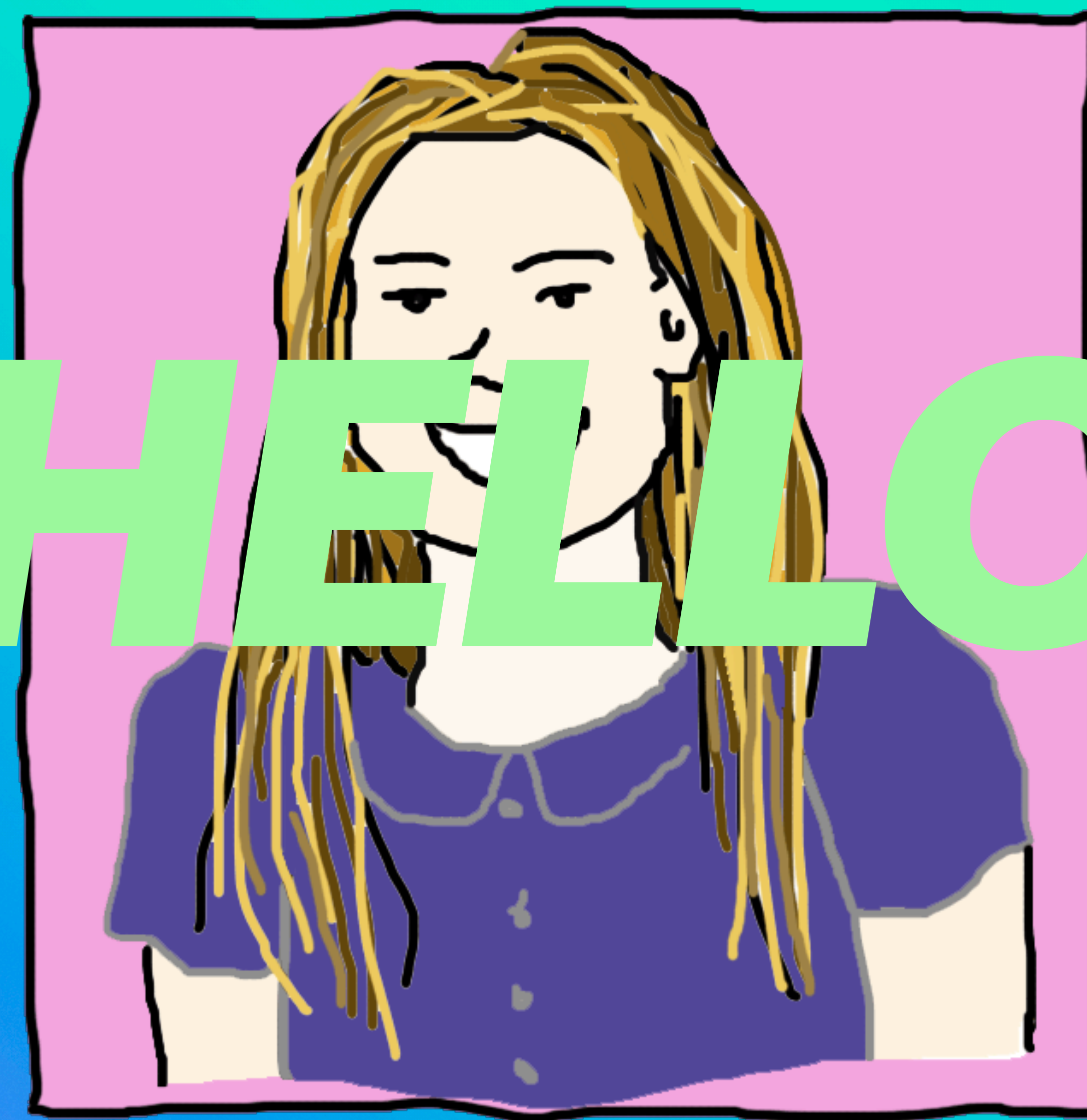


MAKING MUSIC WITH MARKOV (MMM...)

GINA COLLECCHIA /// 08.12.23 /// THEORY CLUB NO. 1

HELLO

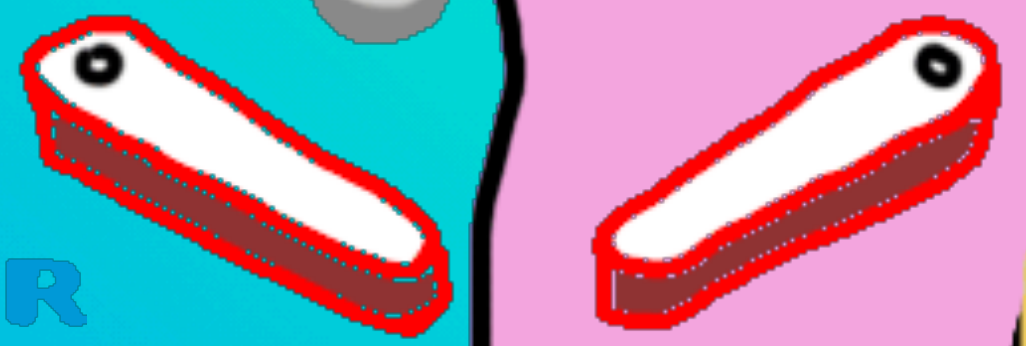


Nudge



JAUNT

SENNHEISER

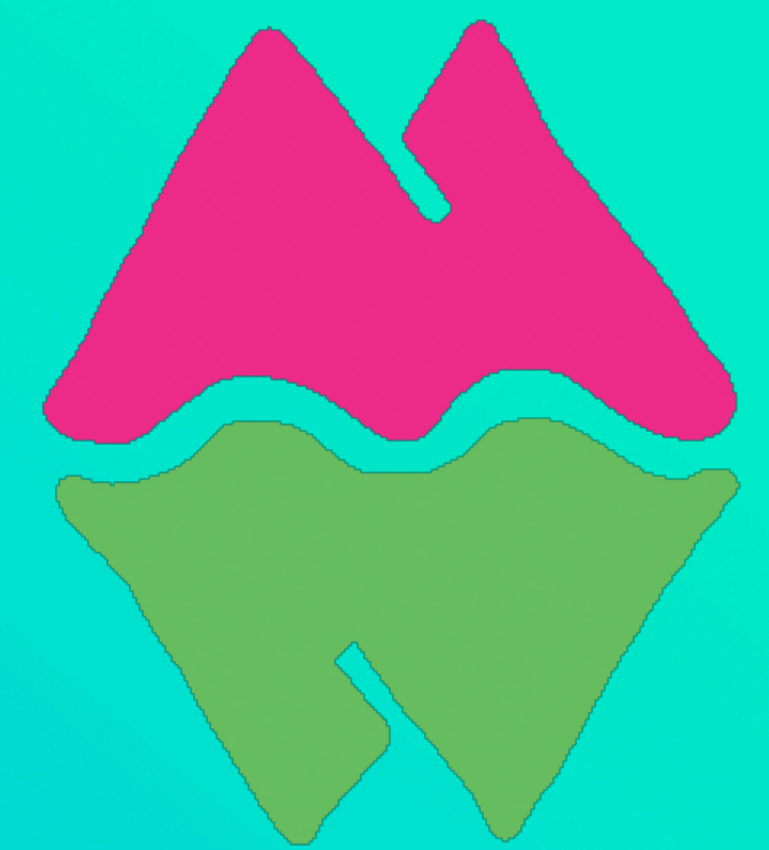


Heuser Hearing Institute

UNIVERSITY OF LOUISVILLE



djz



REED COLLEGE

.SPATIAL

"THE ENTROPY OF MUSIC CLASSIFICATION"

OR, MY REED COLLEGE UNDERGRADUATE THESIS IN MATHEMATICS, 2009

transmission rate to determine how much information is being transmitted, which is $1000 - 1000 = 0$ bits/symbol as we expect for a scheme B , also with 2 outcomes B_1 and B_2 . We might guess that the transmission rate is 81 bits/second, but this does not take into account the entropy. We find that the entropy of the source is $-(0.99 \log_2 0.99 + 0.01 \log_2 0.01) = 0.081$ bits per second, and see that the transmission rate is $81 - 0.081 = 80.919$ bits/second of information. It is rare that two events in the world occur at the same time. Therefore, it is even more rare for the first through the $(t-1)$ st symbols to be the same. The words of a sentence are chosen independently of each other, but together, their meaning is not independent. The words of a sentence are chosen independently of each other, but together, their meaning is not independent. The words of a sentence are chosen independently of each other, but together, their meaning is not independent.

As alluded to earlier, the function of entropy is a function defined for any finite probability scheme A such that $p_k \geq 0$ and $\sum_{k=1}^n p_k = 1$. It is denoted by $H(A)$.

Theorem. Let $H(p_1, p_2, \dots, p_n)$ be a function defined for any finite probability scheme A such that $p_k \geq 0$ and $\sum_{k=1}^n p_k = 1$. It is denoted by $H(A)$.

1. H is maximized when $p_k = \frac{1}{n} \forall k$ (the characteristic we just showed).

2. For the product scheme AB , $H(AB) = H(A) + H(B|A)$, and

3. $H(p_1, p_2, \dots, p_n, 0) = H(p_1, p_2, \dots, p_n)$, i.e., adding an impossible event to a scheme does not change H .

then

$$H(A) = H(p_1, p_2, \dots, p_n) = -\lambda \sum_{k=1}^n p_k \log p_k,$$

where λ is a positive constant.

Proof. Let $H(\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}) = \phi(n)$. We will show that $\phi(n) = \lambda \log(n)$, where $\lambda > 0$. Since H is maximized when $p_k = \frac{1}{n} \forall k$ by the first property, we have

$$\begin{aligned} \phi(n) &= H\left(\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}\right) \\ &= H\left(\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}, 0\right) \\ &\leq H\left(\frac{1}{n+1}, \frac{1}{n+1}, \dots, \frac{1}{n+1}, \frac{1}{n+1}\right) \\ &= \phi(n+1). \end{aligned}$$

For $\lambda = \lambda$ is which case the finite probability scheme features a_1 through a_n is assumed.



The Beatles (whole discography + solo work)



Beethoven, Piano Sonata No. 8, "Pathétique"



The Beatles (whole discography + solo work)



Wire, Pink Flag (1977)

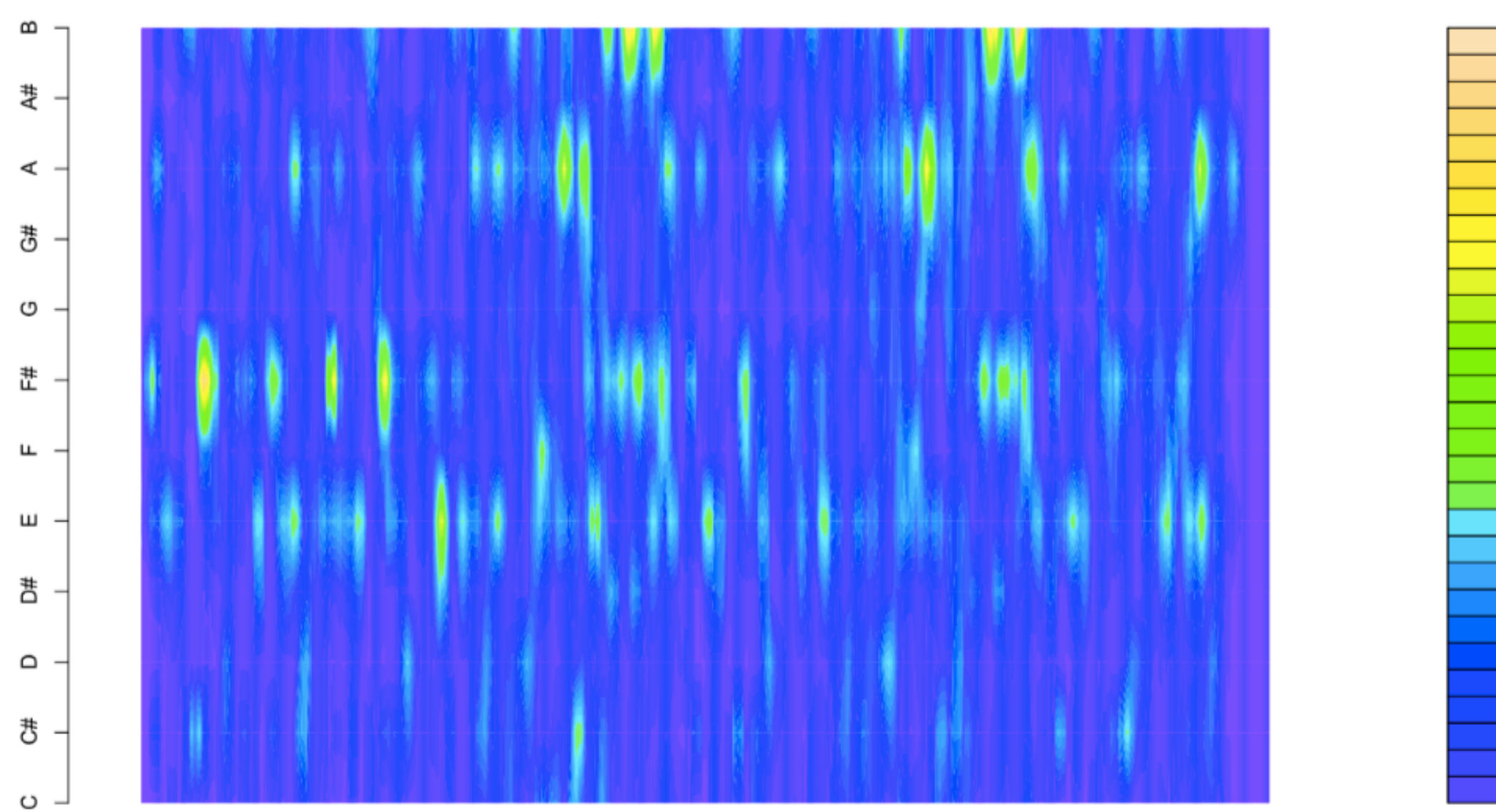


Figure 1.12: Smoothed contour plot of “Oh! Darling”.

Reference Annotations: The Beatles

Note: Please be sure to read [the page describing these annotations](#) before use. In particular, the level of confidence we have in the individual annotations is described there, as well as the original CD issue numbers from which we worked.

Chris Harte's PhD thesis (2010) which describes the chord syntax, transcription process and verification process for the Beatles collection can be downloaded [here](#).

The chord transcription files currently available on this page are version 1.2. of the collection.

[All Beatles annotations in a single tar.gz file](#)

Please Please Me

1. I Saw Her Standing There

- Structural segmentation: [csv](#) [rdf](#)
- Key changes: [csv](#) [rdf](#)
- Chords: [csv](#) [rdf](#) [svl](#)
- Beats: [csv](#) [rdf](#)
- All of the above: [rdf](#)

2. Misery

- Structural segmentation: [csv](#) [rdf](#)
- Key changes: [csv](#) [rdf](#)
- Chords: [csv](#) [rdf](#) [svl](#)
- Beats: [csv](#) [rdf](#)
- All of the above: [rdf](#)

3. Anna (Go To Him)

Working on my thesis with a manually acquired dataset (2009)

Classification	True Entropy Rate
“Tell Me Why” from <i>A Hard Day's Night</i>	0.2437552
“You're Going Lose That Girl” from <i>Help!</i>	0.1828974
“When I'm 64” from <i>Sgt. Pepper's</i>	0.2822246
“Oh! Darling” from <i>Abbey Road</i>	0.3168579
“Two of Us” from <i>Let It Be</i>	0.2249230
<i>A Hard Day's Night</i>	0.3943230
<i>Help!</i>	0.4842243
<i>Sgt. Pepper's</i>	0.4727886
<i>Abbey Road</i>	0.4709648
<i>Let It Be</i>	0.3357670
First mvmt. of Beethoven	0.3933457
Second mvmt. of Beethoven	0.2513897
Third mvmt. of Beethoven	0.3841676
All of Beethoven	0.4578376

A quick introduction to

MARKOV CHAINS

WHAT IS A MARKOV PROCESS?

A.K.A., A MARKOV CHAIN





















- A system in which the probability of each "state" depends *only* on the previous state
- When a state changes, it's said to "transition"
- Characterized by a square NxN matrix of "transition probabilities" where N is the size of the state space:

$$p_{ij} = \Pr(X_{n+1} = j \mid X_n = i)$$

New York

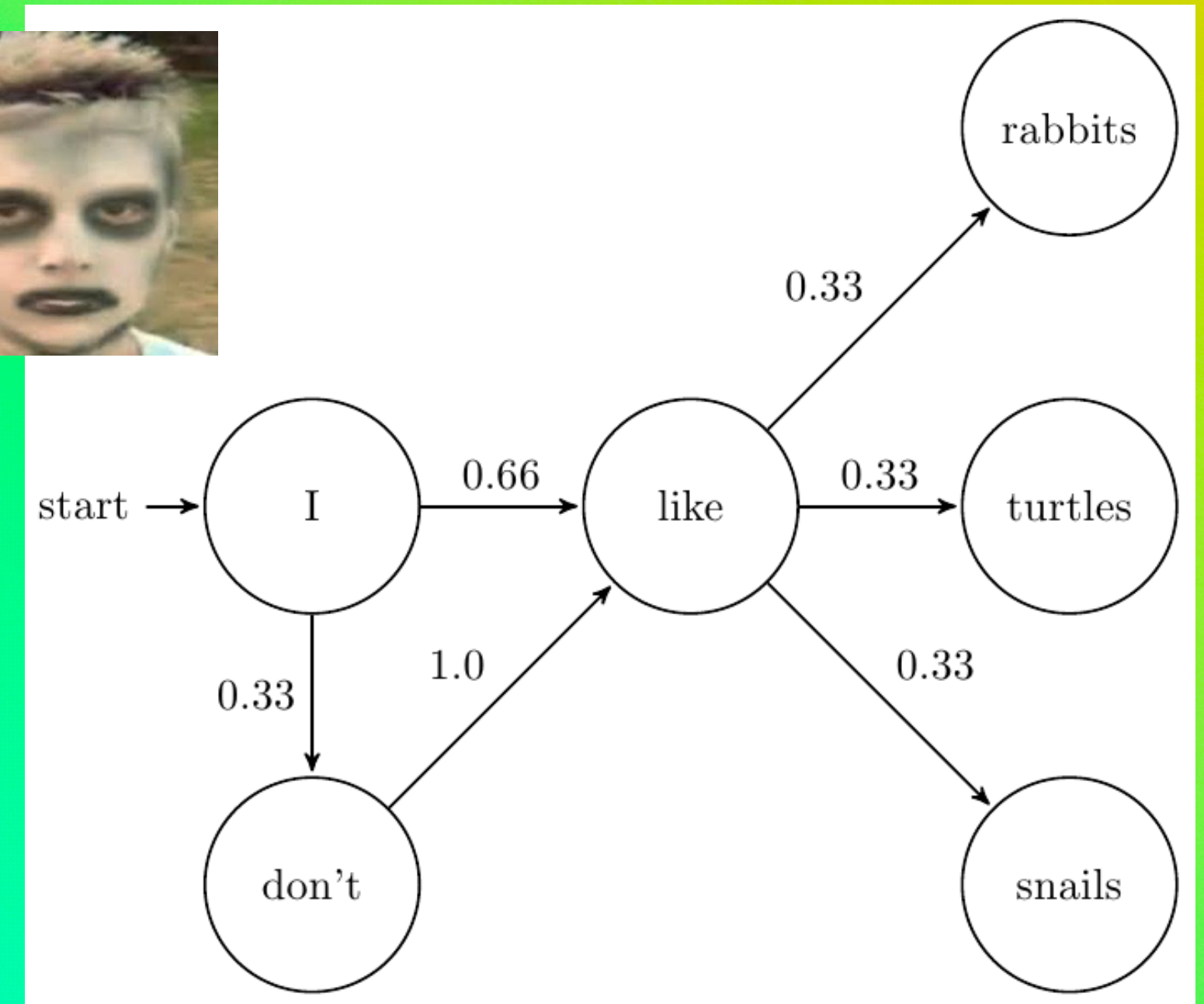
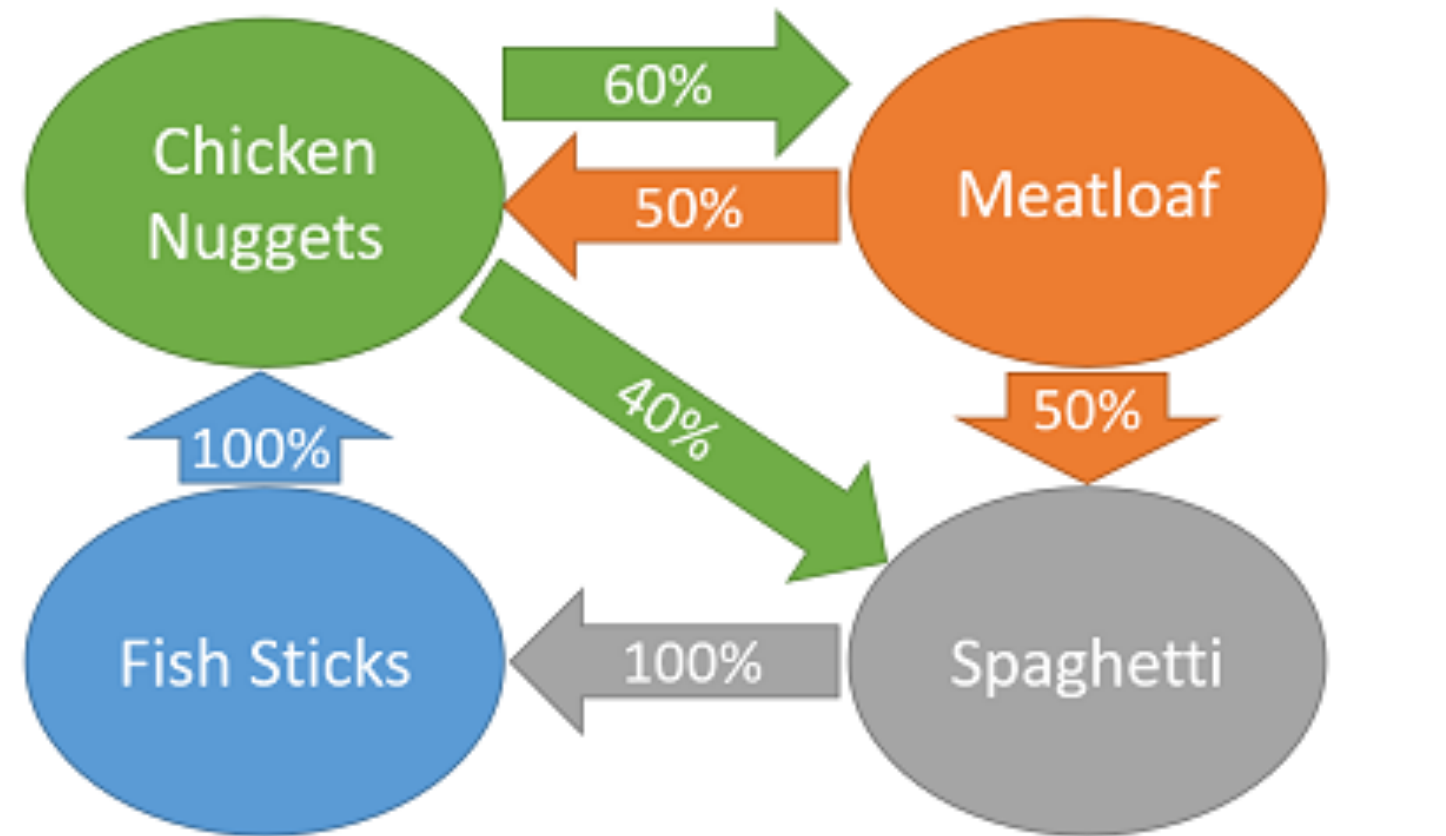
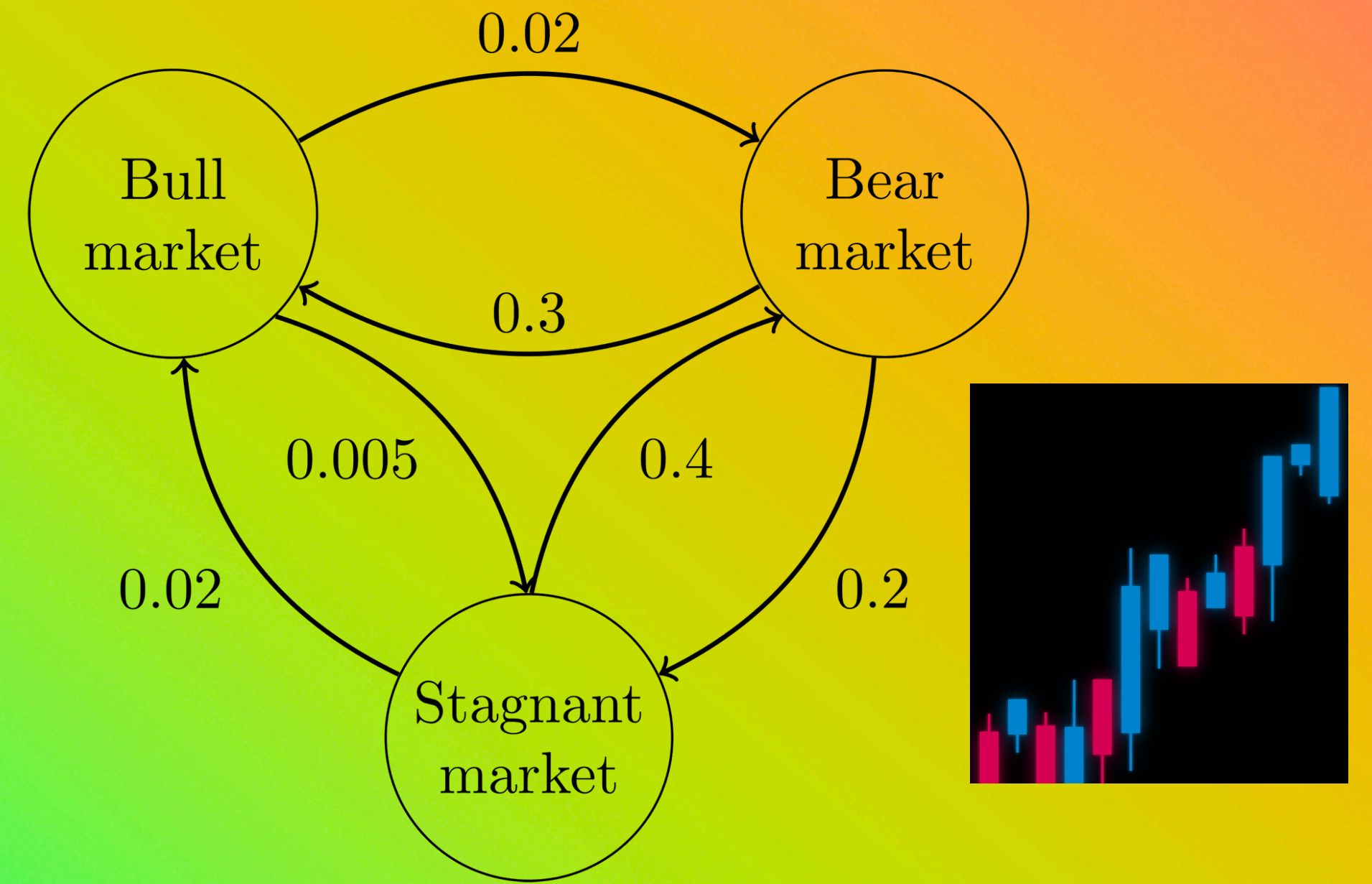
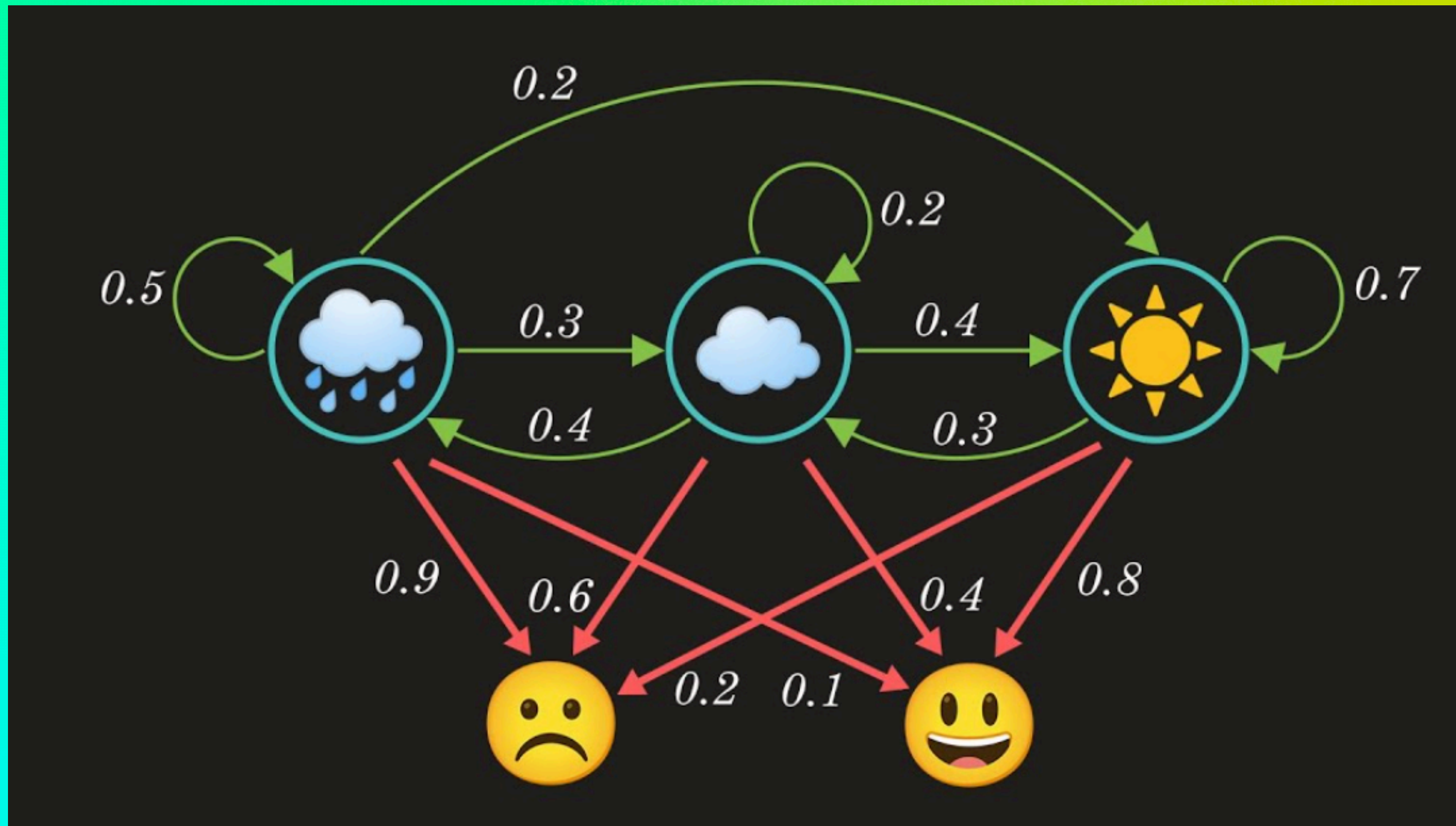
69° | Clear

10-DAY FORECAST

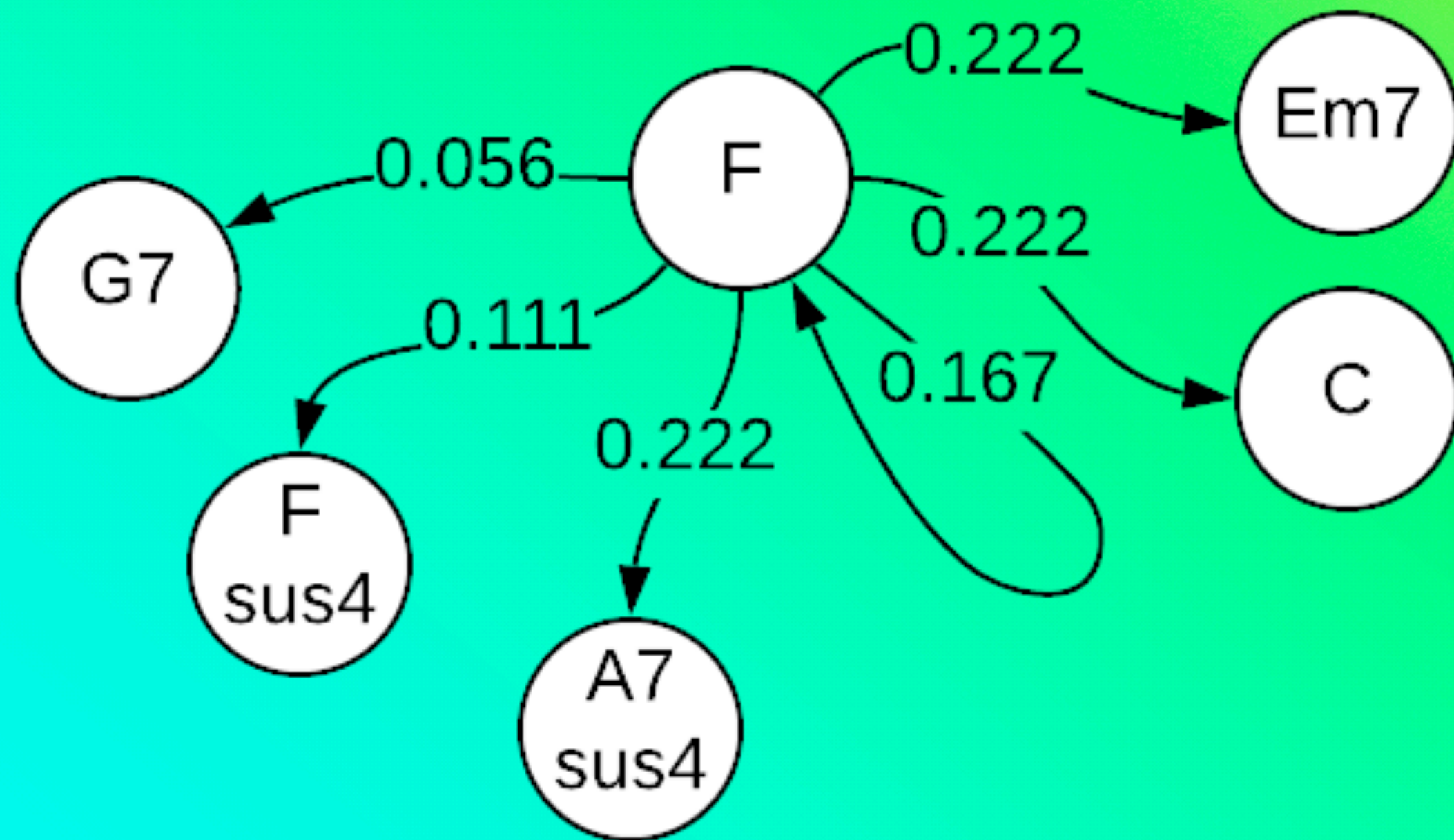
Today		67°		85°
Sat		69°		86°
Sun		73°		90°
Mon		69°		81°
Tue	 50%	71°		85°
Wed		69°		85°
Thu		70°		87°
Fri		70°		86°
Sat		73°		88°
Sun		71°		90°



NO DICE



THE TRANSITION MATRIX

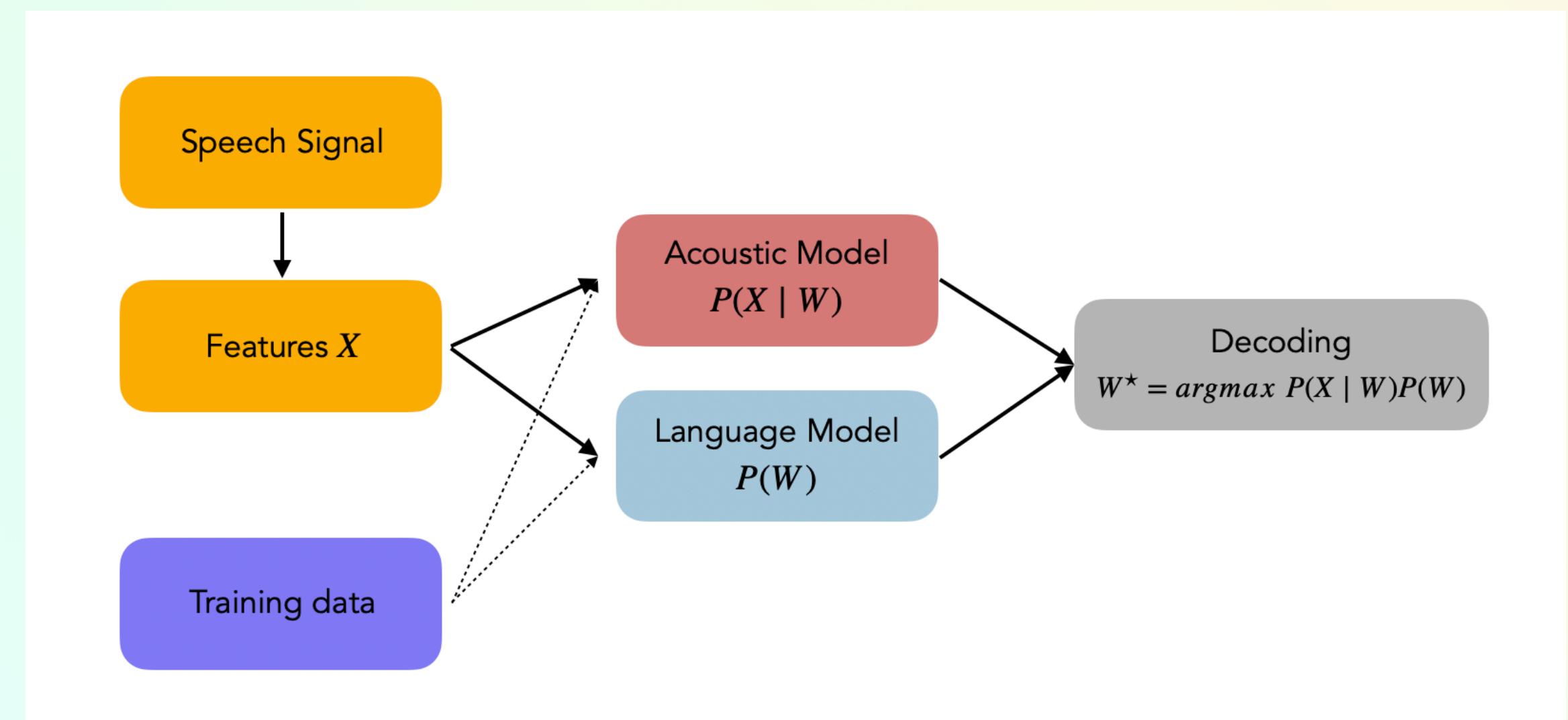
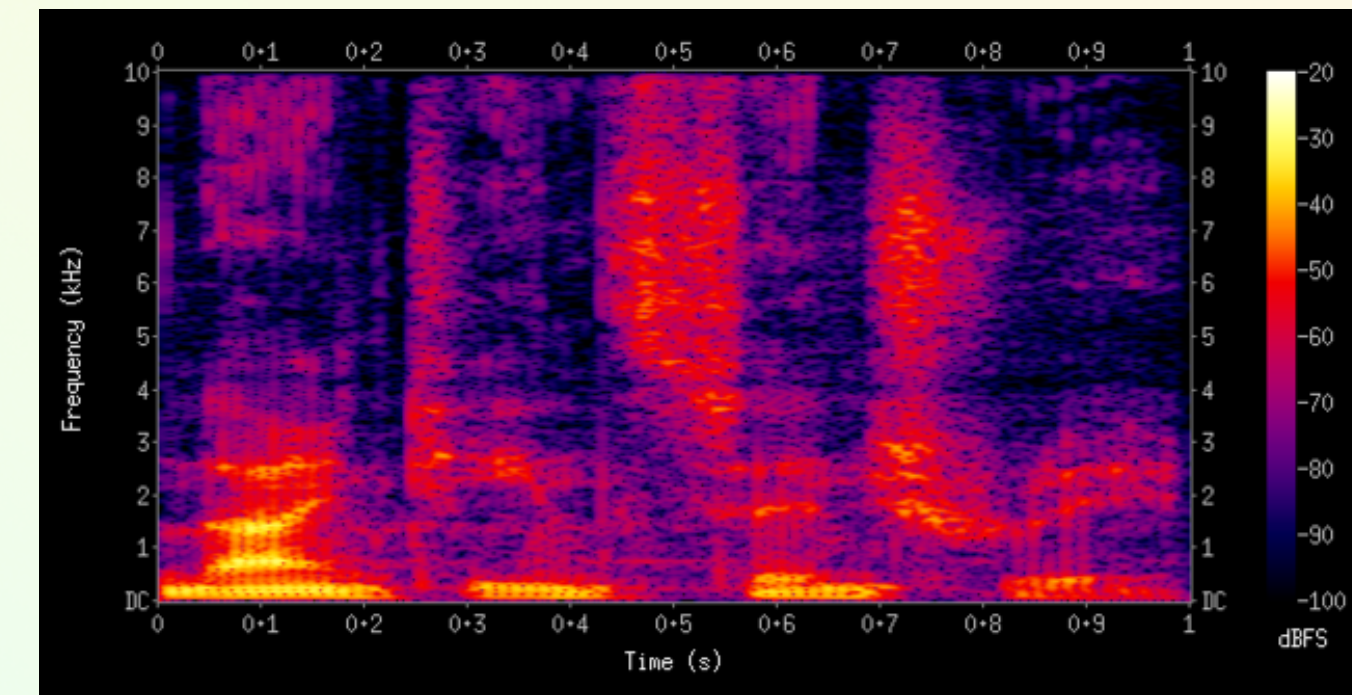


	F	Em7	C	A7sus 4	Fsus4	G7
F	0.167	0.222	0.222	0.222	0.111	0.056
Em7	0.0	0.0	0.5	0.5	0.0	0.0
C	0.3	0.1	0.0	0.1	0.1	0.4
A7su s4	0.0	0.667	0.0	0.0	0.0	0.333
Fsus4	0.1	0.0	0.4	0.0	0.0	0.5
G7	0.3	0.0	0.4	0.0	0.2	0.0

Source: "Markov Chain for music generation", Alexander Osipenko:
<https://towardsdatascience.com/markov-chain-for-music-generation-932ea8a88305>

SPEECH RECOGNITION WITH MARKOV MODELS

- Until about 2010, the acoustic model might use a Hidden Markov Model (HMM) to predict the sequence of phonemes in speech
- Recently these were replaced with other machine learning models such as a Long Short Term Memory (LSTM) model



MUSIC IS A LANGUAGE

I AM PRETTY SURE

GRAPHEMES \Leftrightarrow ***NOTES***

PHONEMES \Leftrightarrow ***TIMBRE, PITCH***

MORPHEMES \Leftrightarrow ***HARMONY***

MEANING \Leftrightarrow ***MEANING***

Keep most of the harmony simple by starting from stage I and only make harmonic changes that complement the bassline

Ib and VI are the best approach chords to IIb7 in the run up to a cadence

All other things being equal, it is a good idea to be in root position I on the first strong beat of a chorale or its upbeat

Root progressions using falling thirds (e.g. VI to IV and I to VI) are much better than those using rising thirds

If you use the same chord twice in a row, make sure that one is in root position and the next is in first inversion

Don't use second inversions other than in IC - V - I

Avoid progressions using II other than as an approach chord to V. Chord IIb can also be used as an approach to IC at a cadence.

Don't use II in root position in minor keys

Don't use iii other than as an approach to vi and avoid in minor keys altogether

Avoid progressions between IV and V unless the melody is going in the opposite direction to the root progression of the two chords

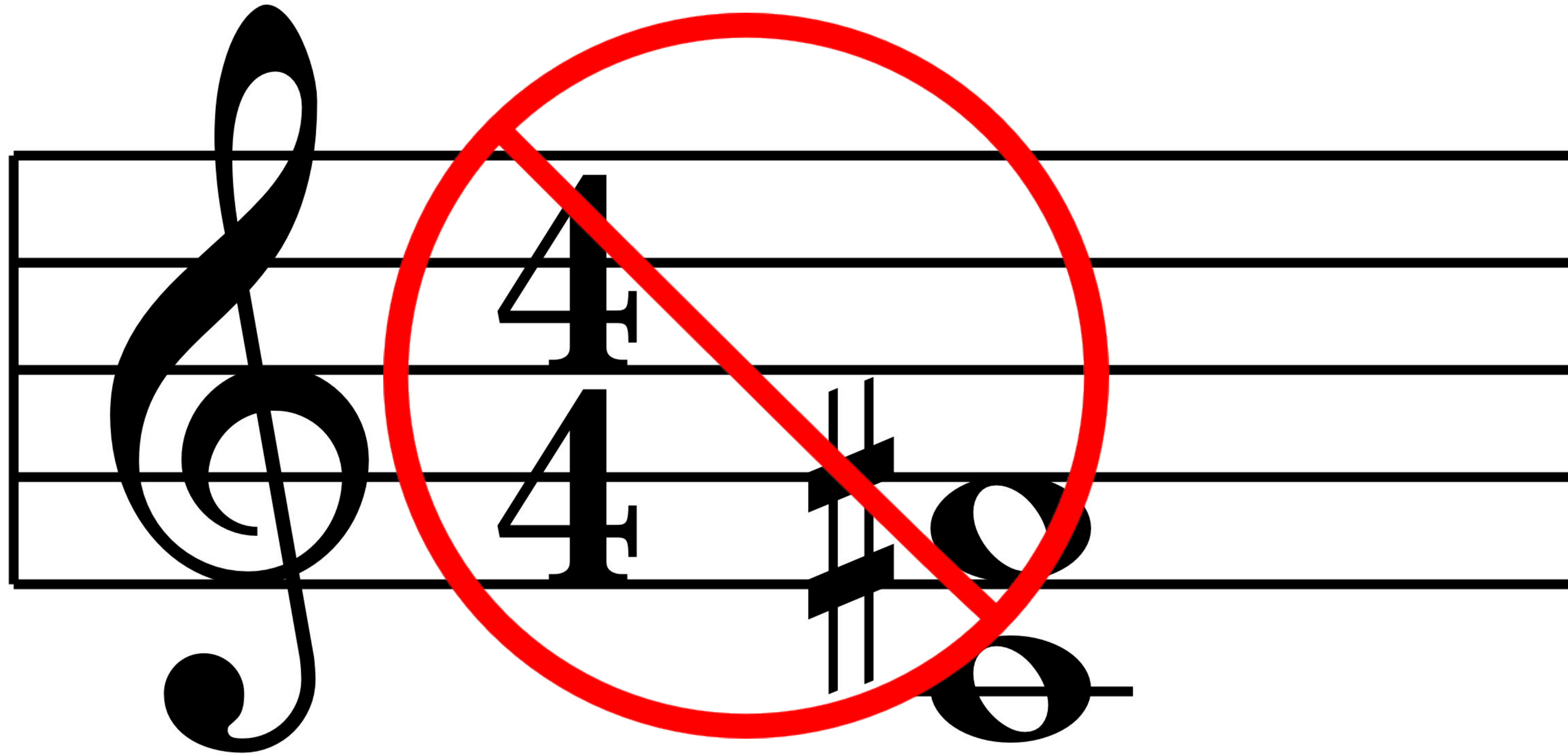
Only use VI in root position

Don't use viio other than as viibo resolving either to I or Ib

Don't repeat the bass note (except from the upbeat to the first beat of a phrase)

BACH DOS AND DON'TS

TRITONE ALERT!!



DEVIL MUSIC

DEMONSTRATION

PLEASE STAND BY

FUTURE WORK

COOL IDEAS I DIDN'T HAVE TIME TO DO YET SO ACTUALLY UNSURE HOW COOL THEY ARE BUT WHATEVER

- MIDI data, Guitar tabs (chords) -> Markov model
- Increase or decrease randomness of output
- Introduce rules about repetition or specific beats
- Whole album models
- Songwriter models

THANK U :)