Songs Generator App



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Difficulties & Used Technologies





Future Vision



Music has always been tied to technology.



Song = Music + Lyrics +Speech

The creation of songs, music, and speech is a significant aspect of human creativity.

Can we teach machines to convey their emotions and ideas through Al songs



Soul Composer: is an app that generate lyrics and music according to the user's selected mood (happy, sad, etc.), then generate voice based on the lyrics and music that our app developed.

Comparison

| | Boomy | My Lyrics Maker | Melobytes | Sound raw | Soul Composer |
|----------------------|----------------------|--------------------|-----------------------------|-----------|------------------|
| Lyrics Generator | Х | From Topic | According To Your Lyrics | X | ✓ |
| Music Generator | √ | ✓ | ✓ | ✓ | ✓ |
| Voice Synthesis | Record Your Voice | X | ✓ | X | ✓ |
| Mood Choice | √ | Х | X | ✓ | ✓ |
| Save As Play List | X | ✓ | X | → | |

Project Pipeline

User Authentication User selects mood or emotion Lyrics will be inspired by emotions

Music will be produced according to emotions

The song can be added to the user's playlist.

Voice Synthesis will be created from the lyrics and music



Lyrics Generator

Structure



Model Objective: generate lyrics based on emotion

Dataset

Preprocessing

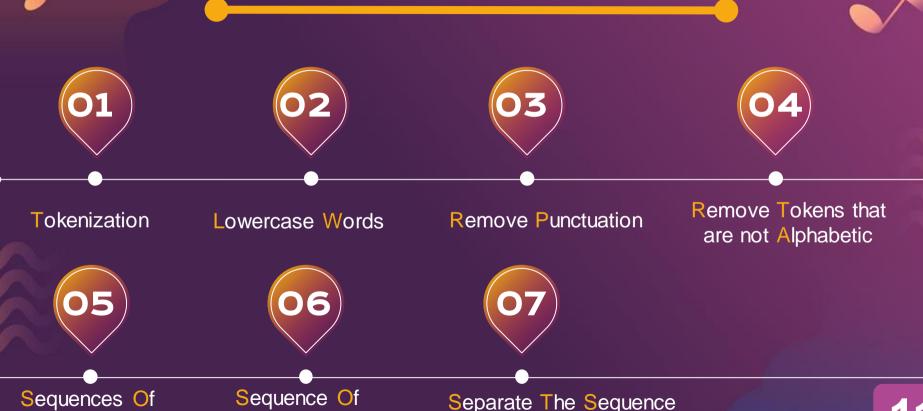
Architecture

Results

Data Set

- We have collected dataset manually
- Based on five emotions, five datasets
- has 200 songs, one for each emotion.

Preprocessing Data



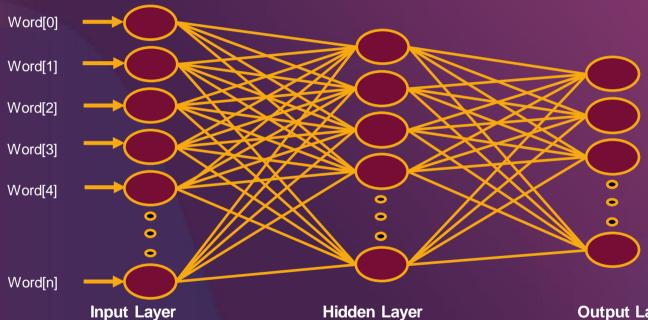
to input and output

Integers

Token

11

Architecture



One Embedding Layer + two LSTM layers + One Dense layer with Relu

Output Layer

Dense layer SoftMax

HOW WILL THE MODEL DO IT?

—(4) Upon adding hyper parameters, lyrics is obtained.

3 The output of the LSTM was sent to two dense layers.

The output of the embedding layer is then taken by the two LSTM layers.

Vocabulary Size, Output Size, and Input length are entered into Embedding layer.

Results

Hyperparameters:

- * Activation : Relu and softmax
- Different epochs: for each five models
- ❖ Batch size: 128
- optimization algorithm : Adam
- loss function: categorical crossentropy

After applying hyper parameters to our five models, we obtain varying degrees of accuracy (92–96%).



Music Generator

Structure



Model Objective: generate music depending on emotion from midi files

Dataset

Preprocessing

Architecture

Results

Data Set

- Our dataset POP909 includes piano tunes stored in the MIDI
 (Musical Instrument Digital Interface) format
- Using MuseScore3, we separate that dataset to five emotions.
- Each dataset contains between 50 to 85 midi file
- music21 (a toolkit for computer-aided musicology, MIT) These MIDI files data were extracted using a Python toolbox.

Preprocessing Data

06



Extract chords from objects scores

Obtain notes and their duration.

Choose the C major key

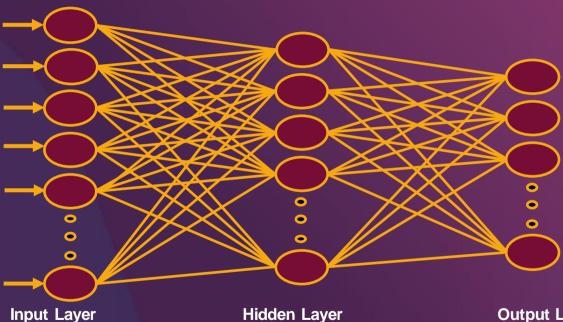
05

Mapping chords and durations to integers

Create Sequence Of Integers for chords and durations

Construct train and target sequences for chords and durations

Architecture



Two Embedding Layer + two LSTM layers + one Dense layers with tanh Function

Output Layer

SoftMax Classification

HOW WILL THE MODEL DO IT?

- Upon adding hyper parameters, notes and durations are obtained.
 - Dense layer changing the output's dimension to 1
 - Then, the two LSTM layers take the concatenated versions of both inputs.
 - Notes and durations are entered into two embedding layers.

Results

Hyperparameters:

- Activation : Tanh and softmax
- Different epochs: for each five models
- ❖ Batch size: 128
- optimization algorithm: RMSprop
- Learning rate: 0.01
- loss function: sparse categorical crossentropy

After applying hyper parameters to our five models, we obtain varying degrees of accuracy (94–96%).



Voice Synthesis

Structure



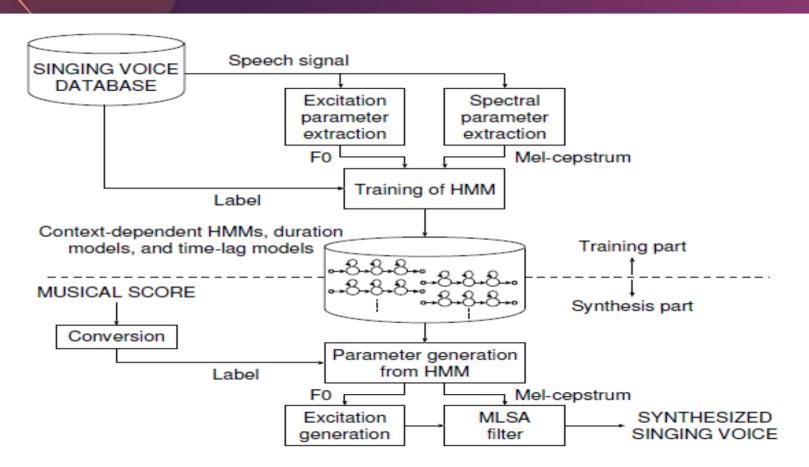
Model's objective: is to generate voice from midifiles and lyrics.

Hidden Markov Models (HMMs) were used to create a singing voice synthesizer.

Data Set

- Children's Song is the dataset we use.
- This collection includes 50 English songs.
- Each audio file is accompanied with lyrics and a MIDI transcription.
- Songs last a total of 72 minutes.

Voice Synthesizer



Voice Synthesizer

- In the training part, first we extract spectral and excitation parameters
- Duration models and time-lag models are also estimated.
- In the synthesis part, first musical score to be synthesized is converted to a label sequence with lyrics.

Voice Synthesizer

- according to the label sequence, spectral and excitation parameters are generated by HMM
- Finally, a speech waveform is synthesized directly from the generated spectral and excitation parameters

How We Dolt?



Voice Synthesizer Example

We would want to express our gratitude to the doctors and listeners, in particular Dr. Nada and Dr. Masa, for guiding us during this journey.



Project Application

How We Do That?

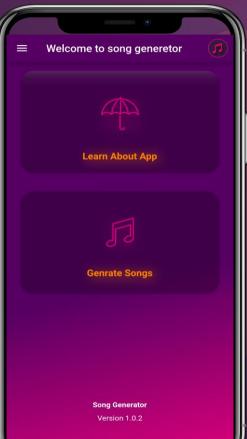
User selects lyrics and music emotion, and API is called.

When an API detects an emotion, it sends it to a model, which then creates music and lyrics.

Generated music and lyrics are sent to the voice model. to generate a voice

Send music, lyrics and voice to the application, the user can now play it 1 2 3

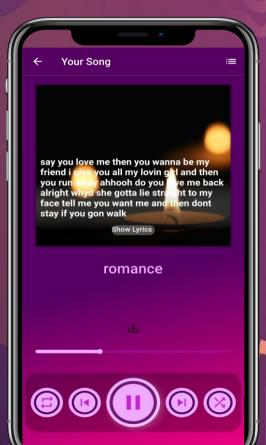


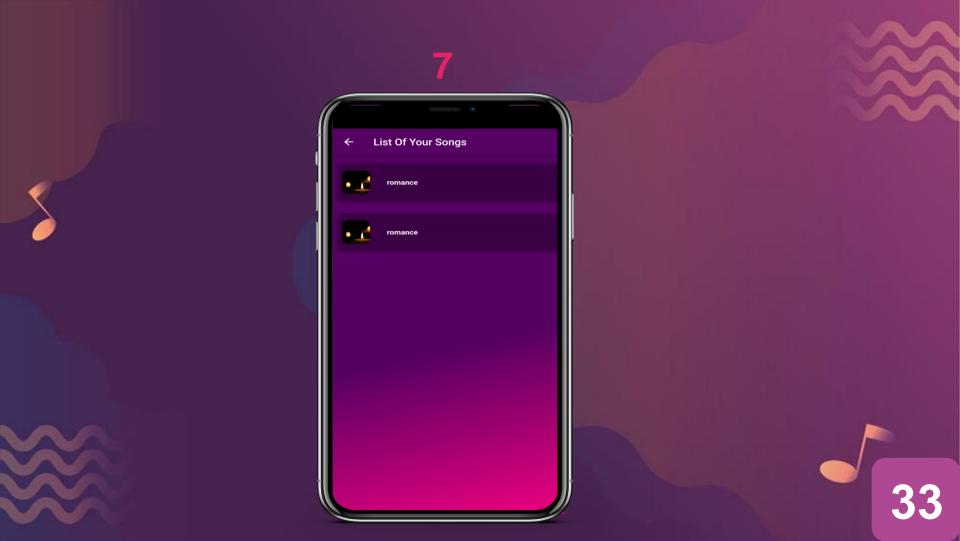












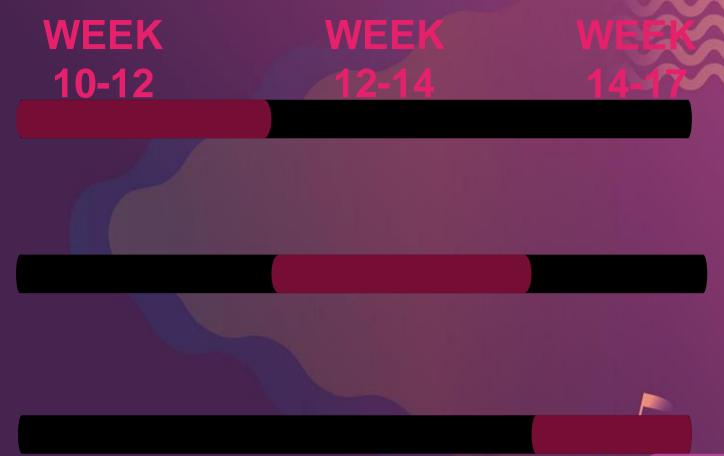
Project Timeline

WEEK 1-3 Building Lyrics **Building** Music Voice **Synthesis**

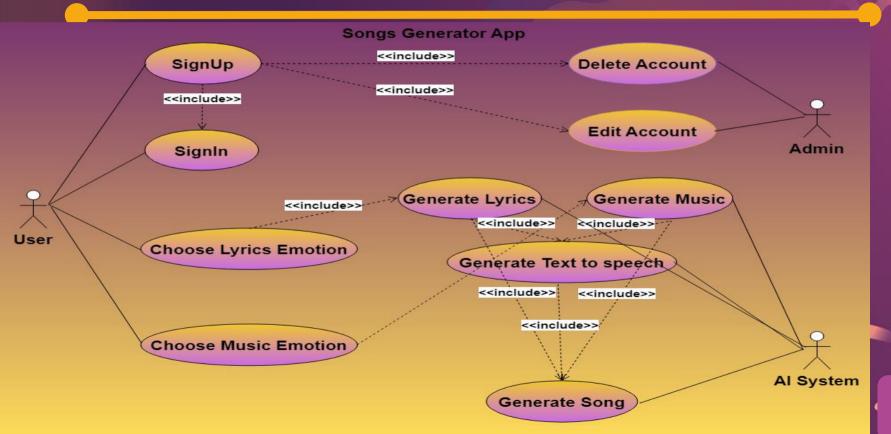
System Analysis

Impleme ntation

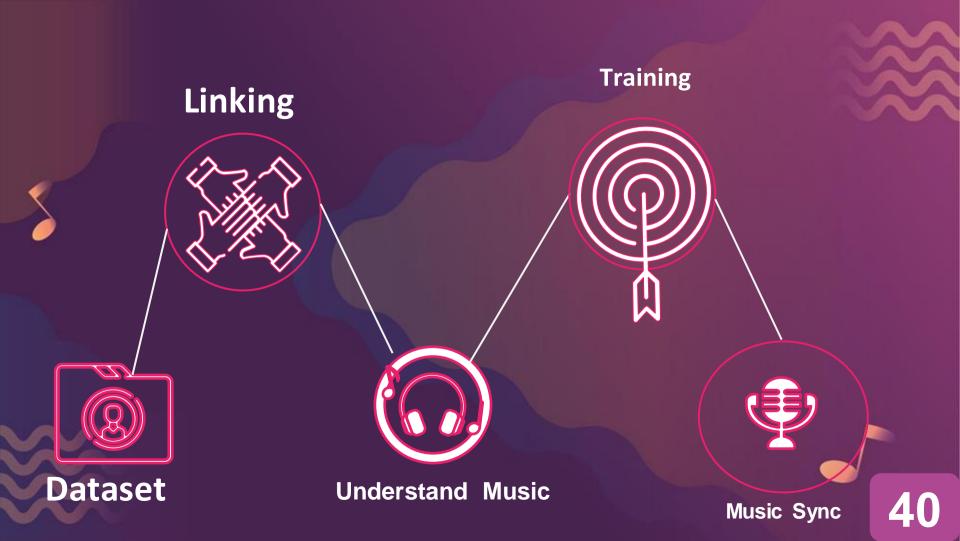
Testing and Report



Use case Diagram













Application Building







Model Developing









THANKS!

For your time, attention, everything!