



# Xuanyi (Steven) Zhu

A current Amazon full-time SDE.

A forward-thinking and versatile computer scientist.

Earned both master's and bachelor's degrees with honors in Computer Science from the University of Illinois at Urbana-Champaign.

Proficient in end-to-end software development, algorithms, ML, and full-stack applications.

(217) 419-6173 zhuxuanyi127@gmail.com

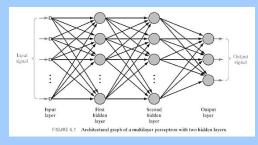
www.linkedin.com/in/xuanyi-zhu

https://github.com/XuanyiZ





- •Used Python and Tensorflow to build neural network(MLP, CNN, LSTM) models that classify noise and human sound to help the audio team achieve noise reduction/cancellation goal with 97% accuracy.
- Applied normalization and regularization with optimal parameters to overcome overfitting issue.
- Optimized and evaluated performance via feature selection, k-fold cross-validation, recall, and precision.
- Presented analysis results to executives. Wrote an ML/Al concept and resources tutorial book.
- Gave lectures to the engineers to help them quickly grasp ML knowledge.



10ms as a frame, 20ms analysis window with 50% overlapping, 512 point FFT, fs: 16,000 Hz

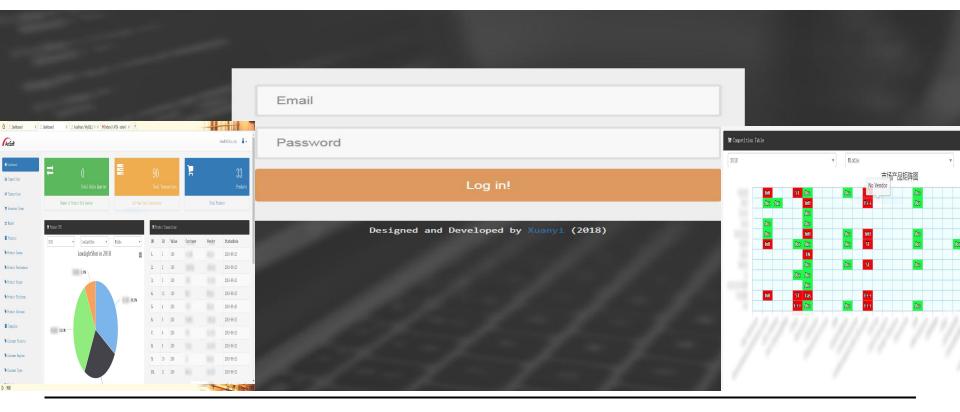


Meet Happy.

Zoom Video Communications

## ArcSoft market inventory management system

A full stack web app that helps track and analyze current market data and reduce labor/time costs.

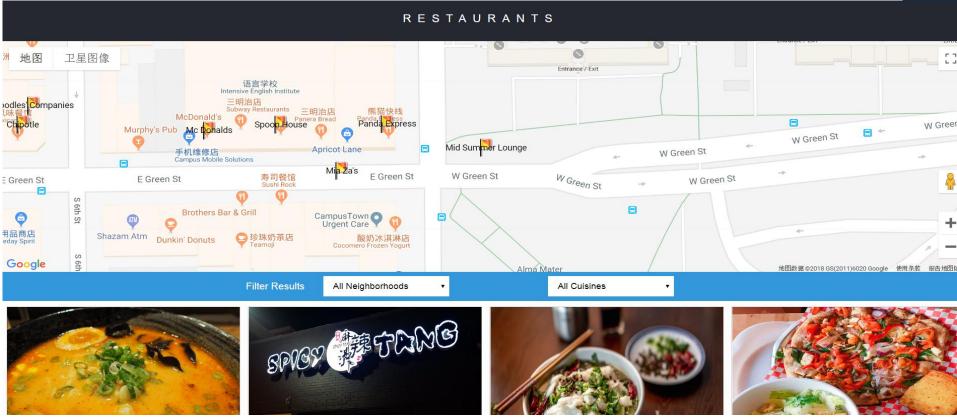


Skills: PHP, JavaScript, JQuery, AJAX, HTML, SQL, Bootstrap, CodeIgniter, HighCharts

## Eatogether

A web app that leveraged busy students' lunchtime to facilitate information exchange and social circle expansion.

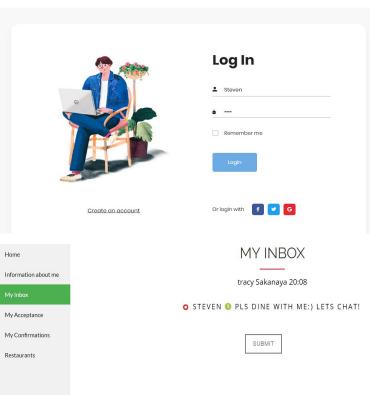


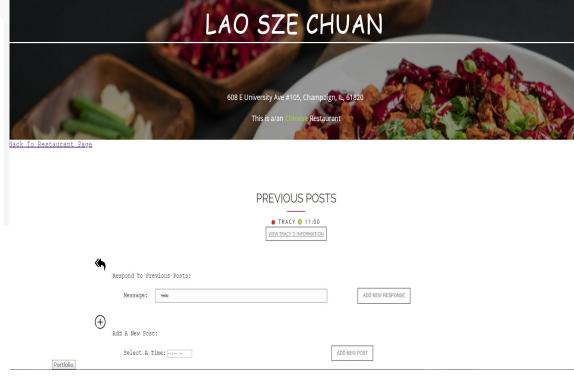


## Eatogether

A web app that leveraged busy students' lunchtime to facilitate information exchange and social circle expansion.







Skills:

Python, Flask, MongoDB, Google map API, cloud, RESTful

Available: http://eatogether.pythonanywhere.com/

# Classic-Game-Collection desktop application



An interactive classic game collection including Snake, Chess, and Sudoku.

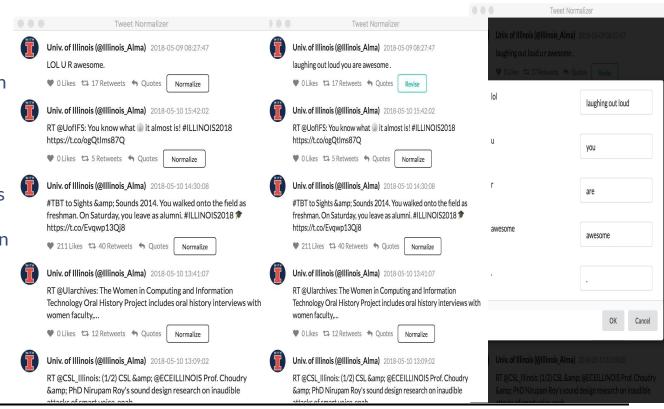


## Twimalizer desktop app

A supervised-machine-learning system to perform lexical normalization for English Twitter text.

### **Key Features:**

- Generated candidates based on past knowledge and a novel string similarity measurement.
- •Used Electron and Vue.js to implement a GUI which interacts real-time with the Twitter API, parsing and displaying the data in the application interface.
- Enhanced accuracy by supplying user-aided revision features that enable normalization engine evolution.



Skills: Python, Electron, JavaScript, Vue.js, Random forest classifier

Available: <a href="https://github.com/XuanyiZ/Text-Normalization">https://github.com/XuanyiZ/Text-Normalization</a>

## NextGreatChef



A recipe sharer web application that helps users to learn and share different recipes worldwide.

My Profile Home All Recipes Logout

## Welcome! nicole111

### \*Your basic information:

Name: nicole

Email address: nicole@gmail.com

Preferred World Cuisine Type: Italian

\*Recipes you added before:

### **Key Features:**

Recipe recommendation: based on users' past view history and desired intake of calories.

Map Of Favorites: a world map shows users' favorite recipes based on the Google map API.



\*Recipes you liked:

Italian Roast Beef

Addictive Sesame Chicken

Mexican Rice

Japanese Onion Soup

Chicken Katsu

## \*We can recommend some recipes

to you based on the maximum calories intake you want

Please enter your maximum intake of calories:

Search





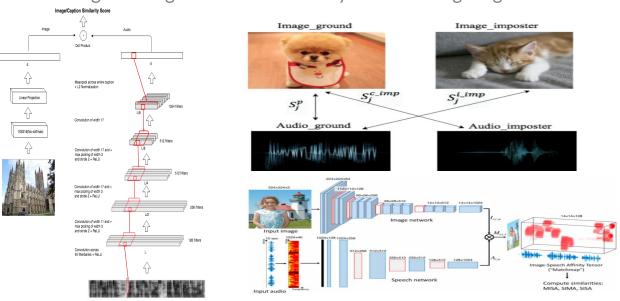


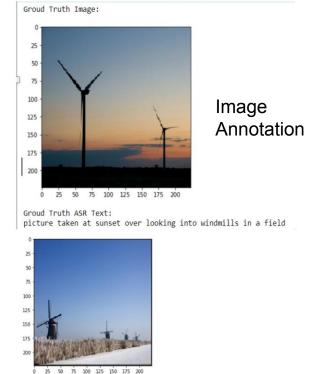
# Optimization Of Multimodal Analysis Of Image And Audio Pairs



Explored the learning ability of neural nets for Image And Audio Pairs Matching.

Research about neural nets discovering word-like acoustic units from continuous speech at the waveform level with no additional text transcriptions or conventional speech recognition apparatus and grounding them to semantically relevant image regions.





a wheat field with many windmills lined up in it there is a clear blue sky in the background

Available: <a href="https://github.com/XuanyiZ/Optimization-Of-Multimodal-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-And-Audio-Pairs-Analysis-Of-Image-Analysis-Of-Imag

# Optimization Of Multimodal Analysis Of Image And Audio Pairs

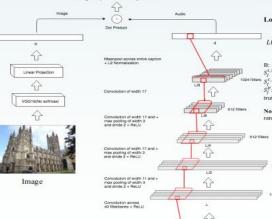


#### Abstract

Given a dataset of images and spoken audio captions, the paper written by Harwath et al. (2016) presents a method for discovering word-like acoustic units from continuous speech at the waveform level with no additional text transcriptions or conventional speech recognition apparatus and grounding them to semantically relevant image regions. Based on their work, we extend and optimize their model to achieve better performance and accuracy to find a better match between image and audios. And our optimized machine learning algorithm can be used to search images based on audio captions or search things in the opposite way.

Image/Caption Similarity Score

#### Model Architecture



#### Baseline Model

#### Image Branch:

Taking VGG 16 layer network(weights fixed) and replacing the softmax classification layer with a linear transform which maps the 4096-dimensional activations of the second fully connected layer into our 1024-dimensional multimodal embedding space.

#### Audio Caption Branch:

Transform the audio segments into speech spectrograms and feed into a customized convolutional neural network as shown below.

#### Connection Part:

The overall multimodal network is formed by tying together the learned visual feature vectors and learned audio feature vectors and map them into a shared embedding space. Then, computes an inno product between them, representing the similarity score between a given image/caption pair.

#### Terms:

Imposter Image/Caption: An image/caption that is randomly selected from the current batch, but is its content is different from the current ground truth image/caption.

#### Loss Function:

$$L(\theta) = \sum_{i=1}^{B} \max(0, S_{j}^{c,imp} - S_{j}^{p} + 1) + \max(0, S_{j}^{i,imp} - S_{j}^{p} + 1)$$

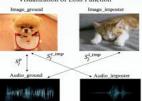
B: Number of Image/Caption pairs.

 $S_j^{\ell,lmp}$ : The Similarity score between  $j^{th}$  caption and its imposter image.

 $S_i^{f,lmp}$ : The Similarity score between  $f^{th}$  image and its imposter caption.  $S_i^{p}$ : The Similarity score between  $f^{th}$  image and  $f^{th}$  caption, which is the ground truth image & audio caption pair.

Note: The index of the imposter image and the index of the imposter caption are randomly selected and they can be different.

Visualization of Loss Function



Audio Caption

#### Optimization

#### Modification of Loss Function

The basic idea is inspired by the original loss function. But we will find the matched image for a selected imposter caption and find the matched caption for a selected image for calculating loss function. And we modified the loss function to be following:

$$L(\theta) = \sum_{i=1}^{B} t1 + t2 + t3$$

B: Number of Image/Caption pairs.

$$1 = \max(0, S_j^{impiCaption.glmage} - S_j^{gCaption.glmage} + 1)$$

$$2 = \max(0, S_j^{gCaption\_imp1Image} - S_j^{imp1Caption\_imp1Image} + 1)$$

$$+ \max(0, S_j^{impzcaption\_grange} - s_j^{impzcaption\_impzlmage} + 1$$
  
=  $\max(0, S_j^{gCaption\_impzlmage} - S_j^{impzCaption\_impzlmage} + 1$ 

$$= \max(0, s_j^{imp2Caption\_glmage} - s_j^{imp2Caption\_imp2Image} + 1)$$

 $S_j^{ACaption\_Bimage}$ : The Similarity score between A's caption and B's image when we are using  $j^{th}$  image/caption pair as the ground truth image/caption pair.

#### Comparison:

Searching for audio recall score: [0.07421875 0.24902344 0.36621094] Searching for image recall score: [0.06933594 0.22363281 0.31933594] Searching for audio recall score: [0.09375 0.26757812 0.36328125] Searching for image recall score: [0.08984375 0.22949219 0.35449219]

#### Baseline

Result Using New Loss Function

#### Modification of Model Architecture

We proposed to augment the spectrogram feature vector by explicitly modelling temporal information in another branch parallel with the baseline architecture. Suppose our spectrogram is an image of size (num\_freq, num\_time, num\_channel). First, three sets of vertical frequency filters are learned to collapse the frequency axis. Then another set of filters are learned for local temporal information. Horizontal max pooling is then applied, before the output is fed into the next layer, which is called "Temporal Convolution Networks".

This kind of network is basically a 1D fully convolutional network with three modifications, causal convolution in residual block, Causal convolution is used to model the relationship flytly-1...,y-k-k, while dilation is used to exponentially increase the receptive field. Residual block is used between previous layer and current layer to learn modifications to the identity mapping, rather than full transformation.

#### An illustration of a single TCN stack is shown below



Figure 1. Architectural elements in a TCM, 10 A delated usual consoleration for the following control of the second secon

We used two stacks of TCN, which, combined, gives a total receptive field of over the whole temporal dimension, so the last temporal vector of the second TCN stack is used as our temporal feature vector. This temporal vector is concatenated with the caption-wised pooled vector of the baseline architecture, with the number of feature maps reduced. The concatenated feature vector is then fed through a dense layer, in the hope to learn a meaningful correlation among the concatenated vector.

#### esults

There are two ways to use our developed machine learning model.

1) The input query is an image, and the output is related captions.

2) The input query is a caption, and the output is related images.

For better understanding and visualization, we transform audio captions into ASR text captions and also provide ground truth images with returned text captions in pairs here. And an example of search result is below. Shown on the top is an image with its ground truth audio transformed into text caption form. Below are its five highest scoring audio files transformed into text caption form.

#### Input Image



Groud Truth AGR Text: picture taken at surset over looking into scieduills in a field

#### Search Results:



1 3



a black and object photo taken of abodelils in the middle of a field there are also several gover likes and there are closes to



I so in 4 do to the second at If you landing up into the second a simplific the sixtle I



se sh se sh se monet in a also open field there are several wind turbines in the field turbine



## Interactive Computer Graphics Projects collection



- Created a 2-D animation of a dancing Illinois Victory Badge.
- •Built a flight simulator game. Generated the terrain using the Diamond-Square algorithm. Used Blinn-Phong illumination model and Phong shading with a colormap. Added a weather-change feature.
- •Wrote an app that loads the Utah teapot and renders it with environment mapping. Added the Quaternions option for the player to modify the teapot position.

• Rendered a system of rainbow-colored bouncing spheres in 3D with the effect of gravity and friction.



# Collaborative Filtering Based Course Workload Prediction system



$$p_{a,i} = \overline{r}_a + \frac{\sum_{u \in K} \left(r_{u,i} - \overline{r}_u\right) \times w_{a,u}}{\sum_{u \in K} w_{a,u}}$$

	cs125	cs233	cs241	cs374
student1	6 (hrs/week)	10	25	16
student2	8	12	N/A	20
				•••

User	Predicted courses	rating for each course
	1 CS225, CS374, PHYS211	4, 4, 3
	2 GEOG101, ECON103, CS125, STAT400, NPRE100	3,3,2,4,2
	3 PHYS212, PHYS213	5,4
	4 CS461, CS357, CS374	3,5,5
	5 CS425, CS421, CS498WEB, HORT107	4,3,4,1
	6 CHLH101, CS225, CS233, CS241	2,5,3,3
	7 ECE220,ECE385	4,4
	8 MATH286, PHYS213, CS473, CHEM102, PHYS214, CS446	3,4,4,3,5,3
	9 CS410, CS411	4,5
1	0 CS241, CS126, CS427	3,5,4

true\_val = [12,6,11,3,10,6,4,8,1,18,1,3,20,16,5,8,19,2,12,9]
pred\_val = [12,12,7.65,4.57,9,6.6,7.1,6.5,3.4,20.1,2.1,6.2,20.5,13,5.5,8.96,23.2,10.5,8,9.5]
rms = sqrt(mean\_squared\_error(true\_val, pred\_val))
print(rms)

170275273425225

A terminal-based application leveraging user-based collaborative filtering algorithm to help students wisely plan their semester course schedule based on their expected course workloads.

Type in your email for example, 'tracy@gmail.com' >>yma34@illinois.edu

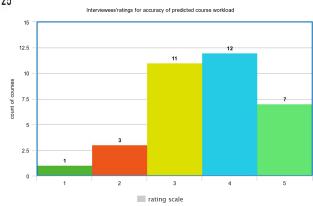
\_Type in your taken courses with corresponding workload sperated by spaces

for example, 'cs125:10 cs233:11 cs225:16 cs241:25'
->>cs125:10 cs233:8 cs225:20 cs241:25

(51, 220)

Type in course you plan to take for example, 'cs242 cs210' >>cs374 ('yma34@illinois.edu', 50) ('cs374', 11)

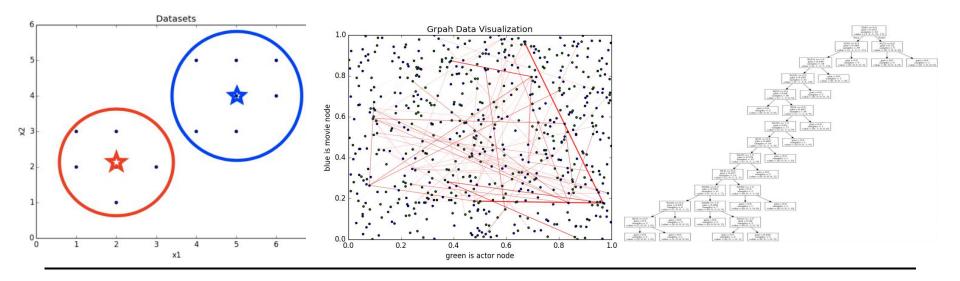
(15.75, 11.507705316731125, 1.475216166015602) prediction workload is 23.550690896583774hours



## Data Science(ML + Data Mining) Projects collection



- Pattern Discoverist: a program that performs frequent pattern mining on datasets and outputs the regular/closed/max patterns.
- •ORIdentifier: an Apriori-based program to identify all the outlier resilient itemsets from given data files.
- •RFC: a general-purpose classification framework using decision trees and random forests.
- QPong: an Epsilon-greedy Q-learning reinforcement agent on a single-player version of Pong.



Skills: Python

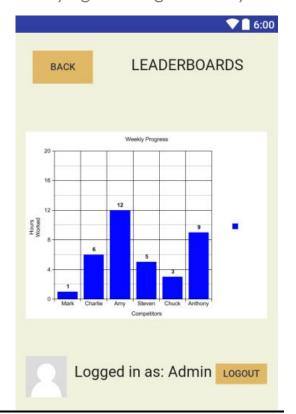
## Productivity Enhancer mobile app



An Android mobile application that enhances individual studying/working efficiency.







A mobile app that allows users to visualize their personal GitHub information, GITAPP such as their repositories, followers, etc. Implemented in Java.



