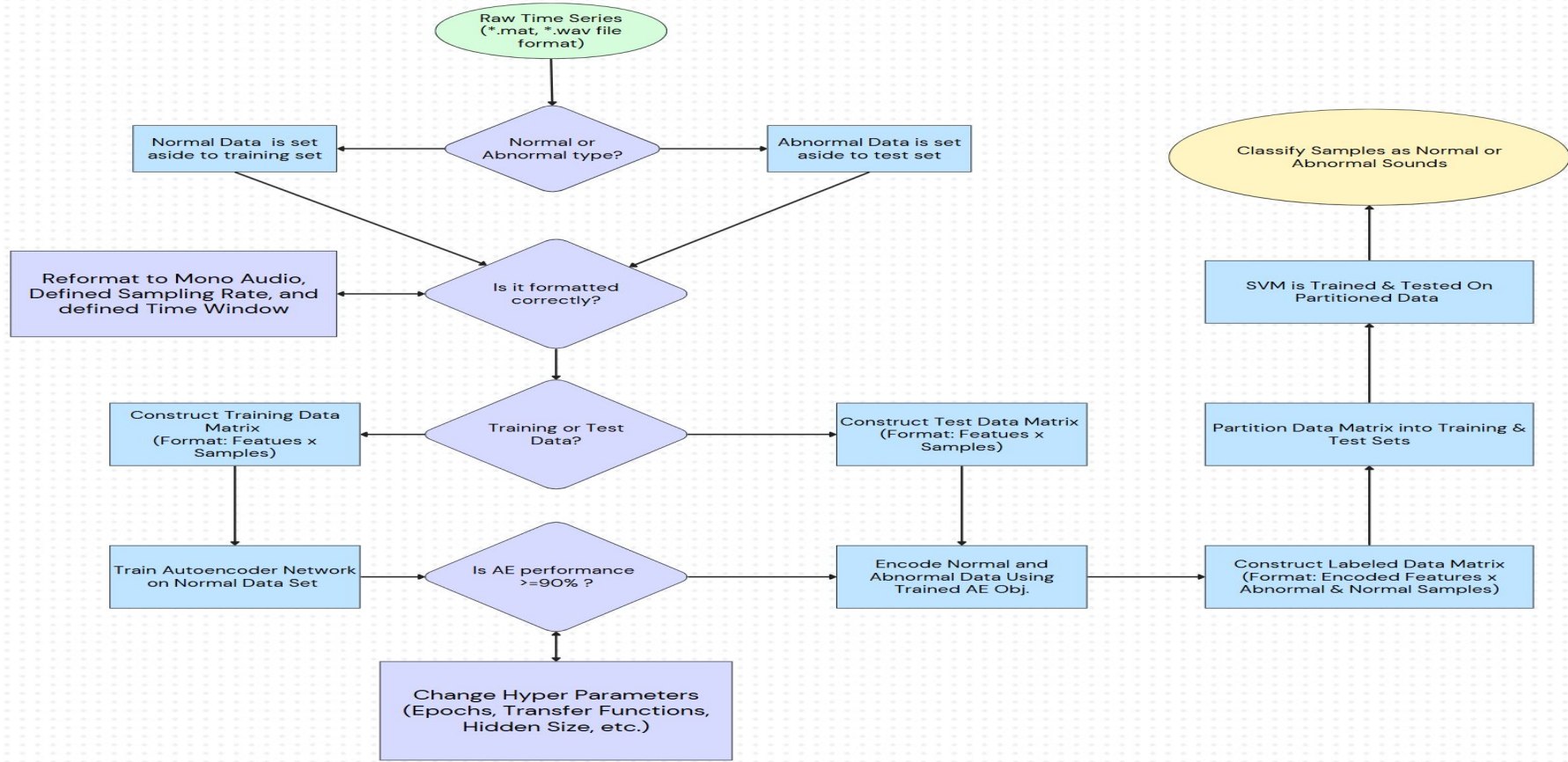


The University of Texas at El Paso

Senior Design Review: Acoustic Feature Analyzer

Eric M. Alonzo | Naval Surface Warfare Center - Carderock Division





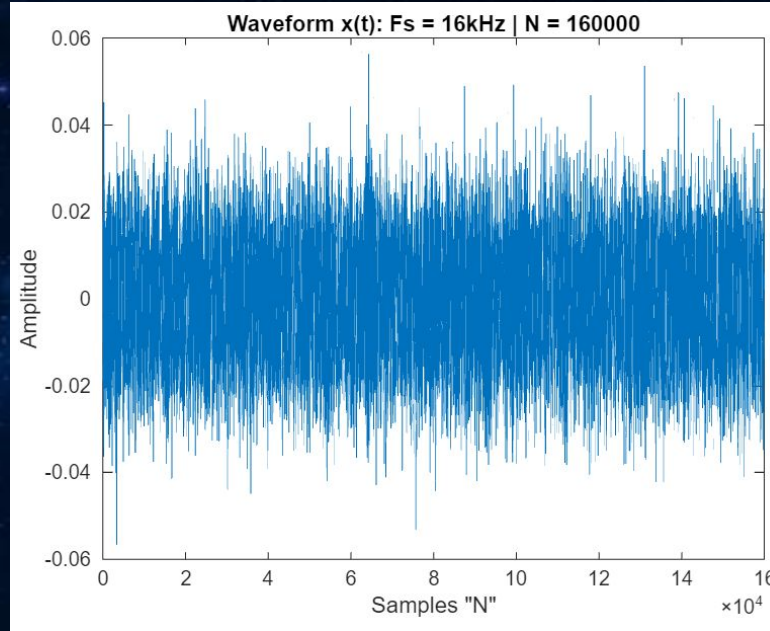
Model/Structure	Type	Accuracy	Interpretability	Training Time	Suitability for Audio	Key Strengths	Key Weaknesses
Autoencoder (AE)	Unsupervised	High	Low	Moderate	Great	<ul style="list-style-type: none"> - Learns compressed representations - Good for anomaly detection - Can denoise audio 	<ul style="list-style-type: none"> - May learn irrelevant patterns - Hard to interpret latent space
Variational Autoencoder (VAE)	Unsupervised	High	Low	Long	Great	<ul style="list-style-type: none"> - Captures complex distributions - Can generate synthetic audio - Regularizes latent space 	<ul style="list-style-type: none"> - Latent features are abstract - More sensitive to tuning
Convolutional Autoencoder (CAE)	Unsupervised	High	Moderate	Long	Excellent	<ul style="list-style-type: none"> - Captures frequency/spatial patterns - Works well with spectrograms - Can handle noise 	<ul style="list-style-type: none"> - Needs large data for good features - May blur fine details
Gaussian Mixture Model (GMM)	Unsupervised	Moderate	High	Moderate	Good	<ul style="list-style-type: none"> - Good for modeling sound distributions - Can separate mixed audio sources - Handles uncertainty well 	<ul style="list-style-type: none"> - Assumes Gaussian distributions - Prone to overfitting with high-dimensional data



Competitive Analysis

Design Requirements

Model can be trained on time series data (Determined by model structure).



Design Requirements

Chosen model must be unsupervised.

- No Labeling Row/Column
- Purely Numerical Array

Rows of Time Samples "N"

	160000x8 double							
	1	2	3	4	5	6	7	8
1	-0.0153	-0.0193	-0.0126	-0.0077	-0.0063	-0.0082	-0.0123	-0.0125
2	-0.0145	-0.0099	-0.0061	-0.0075	-0.0155	-0.0106	-0.0076	-0.0112
3	-0.0093	-0.0094	-0.0077	-0.0138	-0.0120	-0.0106	-0.0062	-0.0099
4	-0.0116	-0.0070	-0.0132	-0.0170	-0.0171	-0.0094	-0.0110	-0.0076
5	-0.0037	-6.1035e-05	-0.0079	-0.0144	-0.0146	-0.0132	-0.0090	-0.0098
6	0.0087	0.0070	-2.7466e-04	-0.0051	-0.0099	-0.0135	-0.0084	-0.0051
7	0.0028	0.0142	0.0166	0.0031	-0.0072	-0.0048	0.0121	-0.0018
8	-0.0073	0.0103	0.0097	0.0118	0.0040	-0.0127	-0.0097	-0.0121
9	-0.0111	-0.0171	-0.0134	-0.0078	-8.5449e-04	-0.0127	-0.0181	-0.0204
10	-0.0126	-0.0203	-0.0221	-0.0203	-0.0176	-0.0051	-0.0082	-0.0032
11	-0.0047	-0.0172	-0.0236	-0.0185	-0.0176	-0.0105	0.0027	0.0020
12	-0.0107	-0.0111	-0.0171	-0.0200	-0.0095	-0.0126	-8.8501e-04	-0.0026
13	-0.0017	-0.0049	-0.0140	-0.0154	-0.0179	-0.0093	-0.0077	-0.0031
14	-0.0063	-0.0084	-0.0090	-0.0201	-0.0184	-0.0162	-0.0101	-0.0028
15	-0.0111	-0.0069	-0.0022	-0.0076	-0.0113	-0.0110	-0.0074	-0.0101
16	-0.0128	-0.0011	-0.0023	-7.3242e-04	-5.7983e-04	-0.0067	-0.0039	-0.0040
17	-0.0109	-9.4604e-04	0.0077	0.0043	0.0050	-3.0518e-04	0.0034	-0.0063
18	-0.0031	0.0076	0.0186	0.0138	0.0086	0.0035	2.7466e-04	-0.0037
19	0.0066	0.0148	0.0117	0.0163	0.0144	0.0066	-5.7983e-04	0.0031
20	0.0091	0.0091	0.0099	0.0156	0.0203	0.0085	0.0079	0.0098
21	0.0028	7.6294e-04	-0.0013	0.0093	0.0169	0.0058	0.0077	0.0055
22	-0.0043	-0.0106	-0.0139	-0.0137	-0.0020	5.4932e-04	0.0010	2.1362e-04
23	-0.0172	-0.0145	-0.0199	-0.0255	-0.0074	0.0026	0.0031	-0.0055
24	-0.0210	-0.0234	-0.0251	-0.0208	-0.0089	0.0014	4.8828e-04	-0.0060
25	-0.0295	-0.0294	-0.0215	-0.0211	-0.0056	-0.0060	-0.0033	-0.0146
26	-0.0176	-0.0194	-0.0163	-0.0190	-0.0148	-0.0089	-0.0095	-0.0186
27	-0.0117	-0.0091	-0.0240	-0.0212	-0.0147	-0.0133	-0.0127	-0.0121
28	-0.0094	-0.0188	-0.0220	-0.0226	-0.0123	-0.0178	-0.0113	-0.0084
29	-0.0094	-0.0069	-0.0089	-0.0170	-0.0117	-0.0208	-0.0189	-0.0071

Columns of Waveforms "x(t)"

Design Requirements

Chosen unsupervised model must work with “relatively” small data sets.

- **200 \geq Number of Sound Files in Data Set \geq 50000**
- Visuals provided use 200 total samples of audio files with 160000 points, a sampling rate (Hz) of 16000 Hz, 7 Channel Audio, and a total time length of 10 seconds.
- Shown tests take on the 1 second chunks of each sound sample.

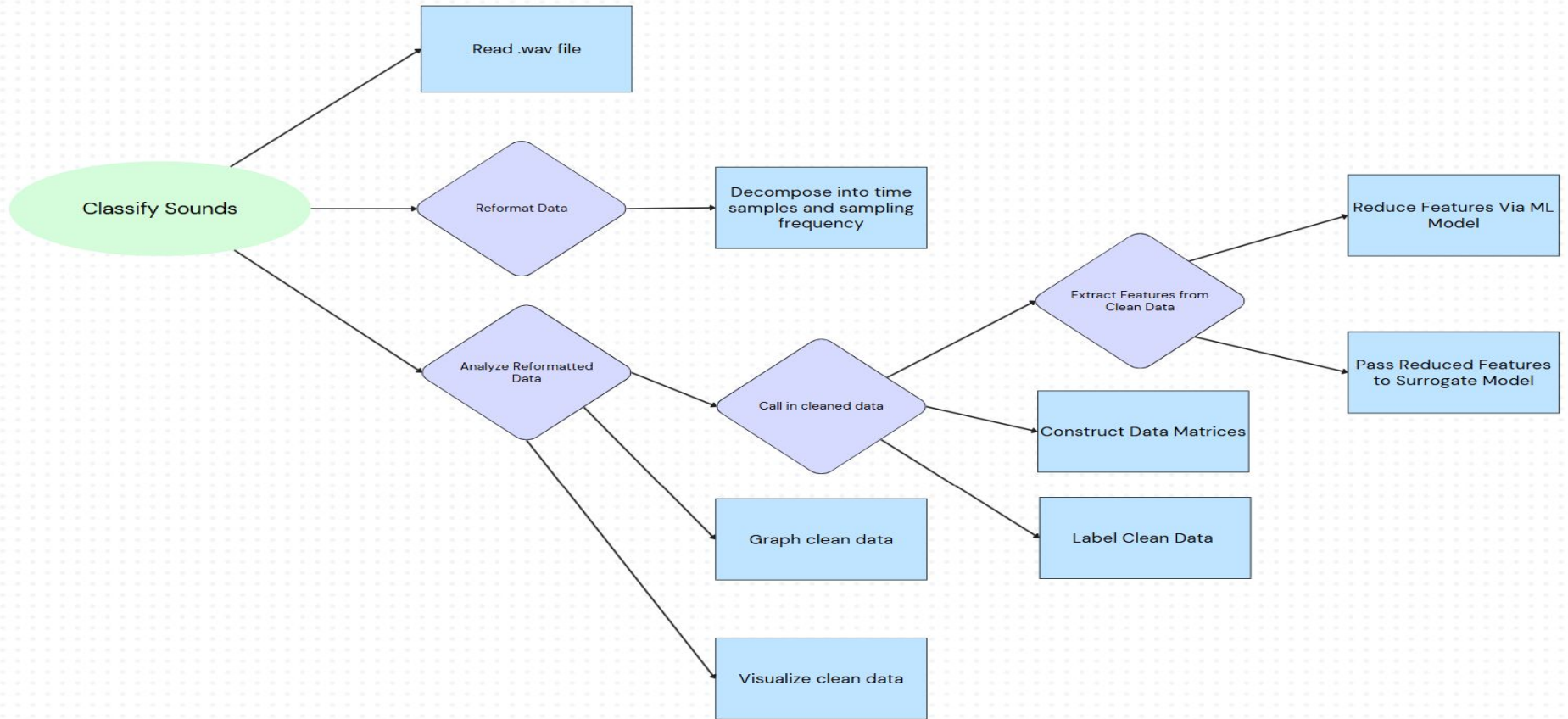


Design Requirements

Most Importantly: Chosen Model must categorize sounds into specified categories.

Confusion Matrix

	Actually Positive (1)	Actually Negative (0)
Predicted Positive (1)	True Positives (TPs)	False Positives (FPs)
Predicted Negative (0)	False Negatives (FNs)	True Negatives (TNs)



Q&A

Design Requirements

Most Importantly: Chosen Model must categorize sounds into specified categories.

Confusion Matrix

	Actually Positive (1)	Actually Negative (0)
Predicted Positive (1)	True Positives (TPs)	False Positives (FPs)
Predicted Negative (0)	False Negatives (FNs)	True Negatives (TNs)

Thank you!

