A Proposed Hybrid Method for Airtarget Acoustic Signature Diagnosis

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Abstract: The Airtarget acoustic signal always suffers from noise discrepancies depending on the environmental conditions. It is very useful to use Airtarget Acoustic Signature for detection, classification and recognition process. The merged noise with the acoustic signal differs from place to place and from day time to another. Without de-noising process the results of airtarget captured signal will be inaccurate. Therefore, a proposed preprocessing technique is introduced to analyze a noisy airtarget acoustic signal. Considering the benefits of the discrete wavelet transform in de-noising, it is used to preprocess the AT signature.

Keywords: Discrete Wavelet Transform (DWT), Discrete Cosine Transform (DCT), Acoustic Signature (AS), Airtarget (AT), signal to noise ratio (SNR), Signal statistics

1. Introduction

Acoustic signature (AS) has a wide usage in Airtarget (AT) detection and classification. It is very useful in detecting very low altitude ATs, which are difficult to be detectable by conventional radars [1]. Its importance arises from its ability to detect ATs that having sudden appearance as drones. Many military researches have been directed to detect ATs using their AS [2]. Generally, systems that use AS for AT detection and classification always suffer from turbulence noise. The most public noise is the wick and wind sound. They present a serious problem at capturing AT acoustic signals [3]. It provides a serious distortion in the analyzed acoustic signal. Therefore, signal de-noising should be used to enhance the AT detection and classification. Using many enhancements processing before signal analysis may cause losses of some information about AT signature. It may cause badness effect on detection and classification accuracy. It also increases the processing time which affect system performance.

Choosing a signal processing method is important to simplify the implementation and provides high accuracy for AT detection and classification. Some signal transformation technique helps in signal enhancement and analysis [4]. The famous transformation method is the wavelet transform. It has a wide usage in signal de-noising [5]. It can be used as 1D or 2D transformation. For signal analysis, as acoustic or biomedical signal it works as 1D signal processing, for image processing as image filtering and video coding it produces analysis in 2D of transformation [6]. Also it improves the performance of communication system for ground penetration radar by de-noising the signal [7].

2. Signal Preprocessing

The purposed method is to de-noise a received signal and to reconstruct an extracted original signal very closed to the original AS as possible. Almost the entire acoustic signal of ATs in real life suffers from noise like wick and wind noise. It should be preprocessed before being analyzed. The present work deals with a noisy AT signal analysis for detection and classification. As the noise increases in AS a hardness of feature extraction exists. The noise usually has low amplitude and high frequency where AS has high amplitude with low frequency. The received AT signal can be represented by the following equation:

\[ X[n] = S(n) + N(n) \]  

where \( X[n] \) is the received signal, \( S[n] \) is an input original signal and \( N[n] \) is noise combined with the signal.

A filtration process should be included in the proposed system. In the present system the DWT is used in processing stages. Due to its benefits in filtration, it has been used to recover the original AT-AS of the captured acoustic signal with noise for multilevel analysis and signal statistics.

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3. Proposed Hybrid System

The Proposed Hybrid System (PHS) deals with capturing of AT acoustic signal. The PHS consists of two main channels: preprocessing channel and feature extraction channel as shown in figure (1).

3.1 Recording Stage

The first stage is a recording stage consisting of a dynamic sensitive microphone to capture the AT acoustic signal from the field at run time. The microphone is characterized by high sensitivity 10-50mV/Pa followed by an instrumentation amplifier.

3.2 Preprocessing Channel

A preprocessing channel is the second stage in the PHS. The main objective from the preprocessing channel is to make a decision for the stopping level of the DWT process and ensures that a suitable sampling rate is selected for this end level. The DCT is used to get the spectral power density (SPD) of the amplified captured AS to determine a -3dB threshold value and to define the lowest frequency \( f_l \), highest frequency \( f_h \), BW and the allowable DWT (L) levels of the AT-AS feature extraction.

The highest frequency of the AS-BW is used to ensure that the sampling rate of the recording system is the most suitable for digital signal processing to satisfy the Nyquest rule:

\[
    f_{\text{sampling}} \geq 2f_h
\]

where \( f_{\text{sampling}} \) is the sampling frequency of analogue to digital converter of the recording system.

The \( f_l \) of the BW is used to determine the number of DWT (L) levels for the captured AS.

\[
    L = \text{Int}\left(1.4427 \times \ln\left(\frac{f_{\text{sampling}}}{f_l}\right)\right) - 1
\]

3.3 Feature Extraction Channel

The feature extraction is the second main channel in the PHS. It includes of three processes: DWT, signal statistical analysis and feature extraction. The DWT analyzes the AS up to the (L) level within the selected BW \( (f_l, f_h) \). The approximation coefficients of the multiple (L) levels are evaluated using the signal statistics tools: Mean, Variance, Median and Entropy to extract the main features characterizing the AT acoustic signal.

3.4 Classification Stage

A classifier based on the K-Nearest Neighbor (KNN) is to be used. It uses the extracted features of AS to classify the AT captured signal. The GUI stage is used to show the output results of the classifier as an interface between the PHS and the user.

4. System Verification

To simulate the PHS F-16 aircraft acoustic signal has been downloaded from internet as an example as follows:

4.1 The F-16 Original Signal Analysis

According to the sequence of signal processing in the PHS, the F-16 AS sample will pass through the preprocessing channel to determine the \( (f_l, f_h) \), BW and L.

The F-16 AS- SPD is analyzed using the DCT as shown in figure (2).

The output is tested by measuring the SNR for the original F-16 AS. It is found that the SNR of the zero-level is equal to -12.06 dB whereas for the fifth level is equal to -10.012 dB at the fifth level. As shown in figure (3) by the solid blue line. It is seen that the SNR increases with the level increase up to the selected level of the DWT-BW for the tested AT-AS.
statistics on the approximation coefficients of the F-16 AS up to the fifth level.

| Table 1: The F-16 acoustic signal statistics |
|-----------------|-----|-----|-----|-----|
| DWT levels      | Mean | Median | Variance | Entropy |
| Level 1         | 0.016286 | 0.0356193 | 0.035491 | 4.604269 |
| Level 2         | 0.023023 | 0.0468731 | 0.049252 | 4.826551 |
| Level 3         | 0.035356 | 0.0838753 | 0.087104 | 5.048071 |
| Level 4         | 0.045939 | 0.1616917 | 0.167181 | 5.168349 |
| Level 5         | 0.064780 | 0.2505101 | 0.262907 | 5.255075 |

4.2 Noisy Signal Analysis for F-16 aircraft

A wind acoustic signal has been merged with the tested sample of the F-16 AS for studying the performance of the proposed PHS. The SNR has been measured for the noisy F-16 AS and it is found for the zero level is equal to -12.96 dB and at the fifth level was equal to -11.667 dB as shown in figure (3) by the dashed red line.

![Figure 3: The measured SNR for the original F-16 AS and the de-noised F-16 AS](image)

Table (2) shows the results of the signal statistics of the merged acoustic signal up to the fifth level.

| Table 2: The F-16 + wind for noisy AS statistics |
|-----------------|-----|-----|-----|-----|
| DWT levels      | Mean | Median | Variance | Entropy |
| Level 1         | 0.016521 | 0.0350580 | 0.050073 | 4.618930 |
| Level 2         | 0.023358 | 0.0317245 | 0.099199 | 4.847366 |
| Level 3         | 0.033012 | 0.0306344 | 0.149969 | 5.010394 |
| Level 4         | 0.046628 | 0.0455395 | 0.213297 | 5.164855 |
| Level 5         | 0.065782 | 0.0639236 | 0.379588 | 5.153133 |

5. Signal Reconstruction

By comparing the results of tables (1) and (2) it is found that the difference is too small at Mean, Variance, Median and Entropy as shown in figure (4). The statistical results of the original F-16 AS are plotted by the red line and the black line shows the statistical results of the de-noised F-16 AS.

![Figure 4: The comparison among statistical analysis for the original F-16 AS and the de-noised F-16 AS](image)

Another comparing between the F-16 original AS and the F-16 reconstructed noisy AS the variance has been measured. Table (3) and figure (5) show the variance for (Mean, Variance, Median and Entropy).

| Table 3: Comparison among the statistical parameters |
|-----------------|-----|-----|-----|-----|
| DWT Levels      | Mean | Variance | Median | Entropy |
| Level 1         | 8.5593E-06 | 1.0445E-04 | 1.2998E-05 | 8.0217E-03 |
| Level 2         | 1.7124E-05 | 1.3690E-03 | 2.2979E-05 | 6.6782E-03 |
| Level 3         | 3.4264E-05 | 2.1842E-03 | 3.9543E-05 | 2.0094E-02 |
| Level 4         | 6.8619E-05 | 1.3316E-03 | 5.0717E-05 | 2.4553E-02 |
| Level 5         | 1.3749E-04 | 8.3306E-03 | 1.2817E-04 | 7.4642E-02 |
6. Results

The comparison between the Statistical parameters results for the original F-16 AS and the de-noising F-16 AS is achieved by using the SNR and the variance measurements. It is found that the results of the variance values at the DWT five levels was too small as shown in table (3). The difference between the SNR at the DWT fifth levels was equal to -1.5639 dB. All of this results provides a success for the using the PHS in signal reconstruction with high performance.

7. Conclusion

The Proposed Hybrid System is characterized by two processing channels. The first channel is used to determine values of \( f_l, f_h \), bandwidth and the allowable levels of the acoustic signature to be used in the discrete wavelet transform. The second channel is level limited analysis to extract the main features of the acoustic signature to be used in the classification process. From the above discussion it is seen that the Proposed Hybrid System provides high performance in the noisy surrounding. This benefit is shown from the difference between the measured SNR values for the original F-16 AS and the de-noised F-16 AS at the DWT fifth level was equal to -1.5639 dB.

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