

**Western Australia School of Mines
Department of Exploration Geophysics**

**The use of distributed sensor arrays in electrical and electromagnetic
imaging**

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Master of Philosophy
of
Curtin University**

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Declaration

To the best of my knowledge and belief this thesis contains no material previously published by any other person except where due acknowledgment has been made.

This thesis contains no material which has been accepted for the award of any other degree or diploma in any university.

Signature: 

Date: 9/06/2011

This thesis is dedicated to Philippa.
Thank you for always showing an interest.

Abstract

Electrical methods for exploring the earth, such as direct current resistivity, induced polarization and electromagnetism are used for numerous exploration, engineering and environmental applications. Common to all these applications is the desire to obtain the clearest possible image of the target. This thesis analyses and develops methods for improving signal to noise ratio for electrical methods.

The ability to recover subsurface information from electrical exploration methods is dependent on the limits of signal detection which is strongly influenced by instrumentation and the conductivity structure of the Earth. Multiple sensors can be used to collect data efficiently over a survey area. Such multi-receiver arrays can improve the signal-to-noise ratio. However, the use of multiple sensors can also be exploited to improve the signal fidelity from each sensor, which may then translate to more accurate geological models and/or greater depth of investigation. In this thesis a two step algorithm for the removal of harmonic noise and atmospheric transients is presented. The first step is the removal of harmonic noise from each sensor using a non-linear single value decomposition (SVD) inversion technique to model a modulated sinusoid to narrow band noise sources. The second step is spherical attenuation using an iterative technique of signal stripping then removing residual coherent noise across the array combined with robust statistical measures in the stacking process. I show that this approach can recover signals otherwise buried in noise and that under certain conditions, signal to noise ratio can be improved by more than 46 dB. The algorithms designed here are applicable to any type of electrical or time domain electromagnetic survey conducted with a multi-receiver array.

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Contents

Abstract	iv
Acknowledgments	v
Contents	vi
List of Figures	vii
1. Introduction	1
2. Noise Characteristics	4
2.1. Observation Noise Characteristics	8
2.2. Time Dependent EM Noise	10
2.3. Spherics Characteristics	14
2.3.1. Synthetic Spherics	17
2.4. Cultural Noise Characteristics	21
2.4.1. Synthetic Modulated Powerline Harmonics	24
3. Signal Processing	27
3.1. Pre-Whitening and Deconvolution	27
3.2. Stacking	32
3.3. Robust Statistics	36
4. Powerline Harmonics	41
4.1. Inversion	42
4.1.1. Non-linear SVD Inversion for Modulated Harmonics	43
5. Array Processing	49
6. Conclusion and Recommendations	58
References	60
Appendix 1. Data Acquisition	65
1A. MIMDAS IP	65
1B. IP Survey at Curtin University	66
1C. EM Survey at Curtin University	66
1D. Noise recording in Darwin	67

List of Figures

Figure 2.1.	A synthetic time domain EM transmitted waveform.	6
Figure 2.2.	Amplitude spectrum of a synthetic time domain EM transmitted waveform in the frequency domain.	7
Figure 2.3.	Generalized geomagnetic spectrum for horizontal magnetic field (H) and induced voltage (V) measurements reproduced from Spies and Frischknecht, 1991.	11
Figure 2.4.	Spherics activity recorded in Darwin, the summer of 1983, data Courtesy of J. Buselli, CSIRO.	15
Figure 2.5.	Amplitude spectrum of spherics activity recorded in Darwin summer of 1983, data courtesy of J. Buselli, CSIRO.	16
Figure 2.6.	Spheric event recorded in Darwin, the summer of 1983, data courtesy of J. Buselli, CSIRO.	19
Figure 2.7.	Unconstrained non-linear optimization to fit real spherics.	20
Figure 2.8.	IP survey at Curtin University showing modulated harmonics.	22
Figure 2.9.	MIMDAS IP data showing modulated harmonics and weak non-stationary components, data courtesy of T. Ritchie, Xstrata Pty Ltd.	23
Figure 2.10.	Synthetic data showing modulated harmonics.	25
Figure 2.11.	Amplitude spectrum of synthetic data showing modulated harmonics.	26
Figure 3.1.	Synthetic spherics.	29
Figure 3.2.	Spiking deconvolution operator derived from spheric events.	30
Figure 3.3.	Convolution of spiking deconvolution operator with spheric.	31
Figure 3.4.	Stacking for a robust estimate of the received signal.	35
Figure 3.5.	Signal rectification.	37
Figure 3.6.	Stacking prior to robust statistics and iterative stacking.	38
Figure 3.7.	Robust statistics as a means of removing low frequency noise.	39
Figure 3.8.	Robust statistics as a means of removing high frequency noise.	40
Figure 4.1.	Harmonic noise removal; the initial estimate the harmonic frequency, stage one.	46
Figure 4.2.	Harmonic noise removal; the initial estimate the harmonic frequency, stage two.	47
Figure 4.3.	Comparison of the received signal before and after powerline harmonic attenuation.	48
Figure 5.1.	Array processing flow chart.	54
Figure 5.2.	Array processing a field example: Curtin electrical data array processing.	55

List of Figures

- | | | |
|-------------|---|----|
| Figure 5.3. | Array processing a field example: Curtin electromagnetic data array processing. | 56 |
| Figure 5.4. | Array processing a field example: MIMDAS data array processing, data courtesy of T. Ritchie, Xstrata Pty Ltd. | 57 |

1. Introduction

Electrical methods for exploring the earth, such as direct current resistivity, induced polarisation (IP) and electromagnetism are used for numerous exploration, engineering and environmental applications. Common to all applications is the desire to obtain the clearest image of the target signal as possible.

In mineral exploration there is a continuing pursuit of both new ore bodies, and more precise 3D imaging of existing ore bodies. Since the shallow targets have mostly been discovered, the minerals industry is increasingly focused on deeper targets. Electromagnetic exploration methods need to be improved to maintain acceptable resolution for deeper targets. One way to achieve this is to improve signal to noise ratio at the EM receivers. Our ability to measure rapidly changing small EM fields in and above the earth is impeded by the presence of natural and cultural electromagnetic noise. Improved acquisition and processing for the electromagnetic methods are required if greater depth of penetration for the technique is to be achieved.

Environmental and engineering electromagnetic surveys are often carried out in proximity of industrial and residential areas. The electromagnetic noise in these areas is more complex as it originates from many different sources such as inductive load imbalances and switching transients. These noise sources are termed “cultural” as they relate with activity of mankind and are different from the “natural” noise sources which relate to natural Earth processes and solar activity. The important difference between these two sources of noise is their level of coherency over short periods of time. Given the inertia of the electrical grid, cultural noise tends to be more or less stationary and coherent for periods under one second. Natural noise sources are more or less stochastic events. Fortunately, the short-term coherency of cultural noise allows it to be estimated, then subtracted from data records. This approach is explored in details in the following chapters.

The complexity of the new targets have caused a resurgence in; (i) the use of multi-electrode receiver systems in DC electrical resistivity, (ii) IP methods (e.g. White et al., 2001) and (iii) the use of multiple receiver systems for the time-domain electromagnetic method. This increase in the number of receivers offers the potential for deeper penetration, improved resolution, and detection of weak conductors. These benefits can be extracted from richer multi-receiver datasets with the help of signal processing algorithms that use the additional information to improve the target response. A major objective of this thesis is to design and test a multichannel processing algorithms designed to improve the data measured from arrays of time domain EM receivers.

To assist the reader, this thesis has been divided into five chapters. Chapter two examines noise sources that can contaminate electrical data and their impact on the quality of the electrical data that can be recorded. This chapter also introduces algorithms designed to generate synthetic noise series to be used for testing of the noise suppression algorithms developed in Chapters 3, 4 and 5.

Chapter three provides a review of traditional signal processing methods for time domain electromagnetic and electrical methods. Also I introduce the possible use of spiking deconvolution as a tool for spherics elimination. This chapter concludes with a novel approach to stacking and the use of robust statistics.

Chapter Four investigates inversion methods for the removal of powerline harmonics. In this chapter I introduce an innovative approach of single value decomposition (SVD) inversion for the removal of modulated powerline harmonics.

Chapter Five reviews remote reference techniques for noise minimization, and presents a two step array processing algorithm for the removal of harmonic noise and atmospheric transients. The first step is the removal of harmonic noise from each sensor using the non-linear SVD inversion technique presented in Chapter Four. The second step is spherics attenuation using an iterative technique of signal stripping then removing residual coherent

noise across the array. This is combined with robust statistical measures in the stacking process.

2. Noise Characteristics

The objective of a geophysical survey is to obtain information about the spatial distribution of one or more of the Earth's physical properties from a limited set of measurements of a related physical field. The electrical properties of the subsurface may be determined using either static or time varying electromagnetic fields (CSEM). Methods that rely on static fields rely on a steady flow of current in the ground. Controlled source electromagnetic methods, on the other hand, rely on the propagation of electromagnetic fields from a time varying electric or magnetic source (West and Macnae, 1991; Parasnis, 1997). Signal to noise ratio is fundamentally important when measuring late time responses from conductive targets. Variations in the received signal due to the target must be observable in the presence of natural and cultural noise sources. The received signal is composed of the actual target signal, time dependent EM noise, geological noise and observation noise. The statistical properties of these three forms of noise are different, and relevant suppression techniques are required.

This chapter considers sources of time dependent EM noise and observation noise. It also provides a detailed review of the intrinsic noise of the receiver (observation noise), spherics and modulated harmonics (time dependent EM noise).

For the purpose of testing processing algorithms with a controlled data set, synthetic received waveforms are generated and imprinted with synthetic noise. The advantage of using a synthetic receiver and noise sources is that the exact noise contributions are known, as well as allowing for specific noise situations to be tested.

The transmitted and received signal for a DC resistivity, IP or time domain EM surveys is essentially an alternating current (AC) or time-varying waveform. Such alternating/changing waveforms are imperative as they allow for the influence of static errors, instrumental drift and electronic, low frequency noise in the first amplification stages to be removed (Becker and Cheng, 1987).

In this research, only the full digitized waveform is recorded, as this provides an uninterrupted record of any noise sources as well as providing detail about the transmitter. This leads to a better understanding of the signal and noise behavior as there is no ambiguity over signal at any time, which leads to improved interpretation.

For the purpose of testing processing algorithms with a controlled data set, synthetic transmitted and received waveforms can be generated for DC/IP and time domain EM data sets. The synthetic transmitted waveform for DC/IP and time domain EM can be approximated by a 50% duty cycle waveform with period T , having an exponential rise with time constant t/τ , where τ is a time constant determined by the transmitter impedance (size, resistance, Earth resistivity), and the transmitter design with duration $T/4$. The turnoff ramp can be approximated as linear with a duration from 0.03 to 0.6 τ . The last half period of the waveform is the inverse of the first half (Asten, 1987).

For time domain EM, the received waveform (for a confined conductor) is the derivative of the transmitted waveform convolved with the Earth's exponential response:

$$\text{Earth response}(t) = -1(1 - \exp(-t/\tau_e)), \quad (2.1)$$

where τ_e is the time constant for the Earth's decay (Asten, 1987). Figures 2.1 and 2.2 show synthetic transmitted and received waveforms for time domain EM using equation (2.1) in the time and frequency domain respectively.

For a layered Earth the late time Z-component response may be represented as power law decay t^{-x} , with x equal to 2.5 for a homogenous Earth, 4 for a conducting layer over a resistive half space and 1 for a resistive layer over a conducting half space (Lee and Lewis, 1974; Kaufman and Keller, 1983).

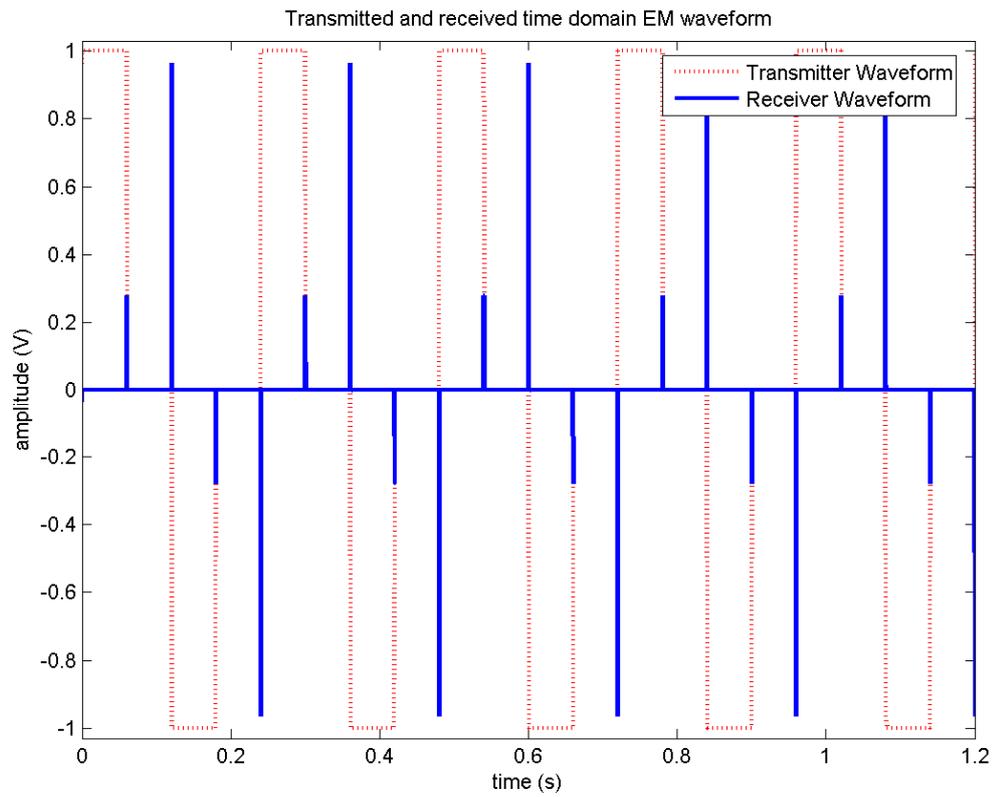


Figure 2.1. A synthetic time domain EM transmitted square waveform. Duty cycle 50 %, base frequency 4.166 Hz, input current 1 A current, sampling frequency 10 kHz. The received waveform is the derivative of the transmitted waveform convolved with the Earth's response (assuming a confined conductor), Equation (2.1).

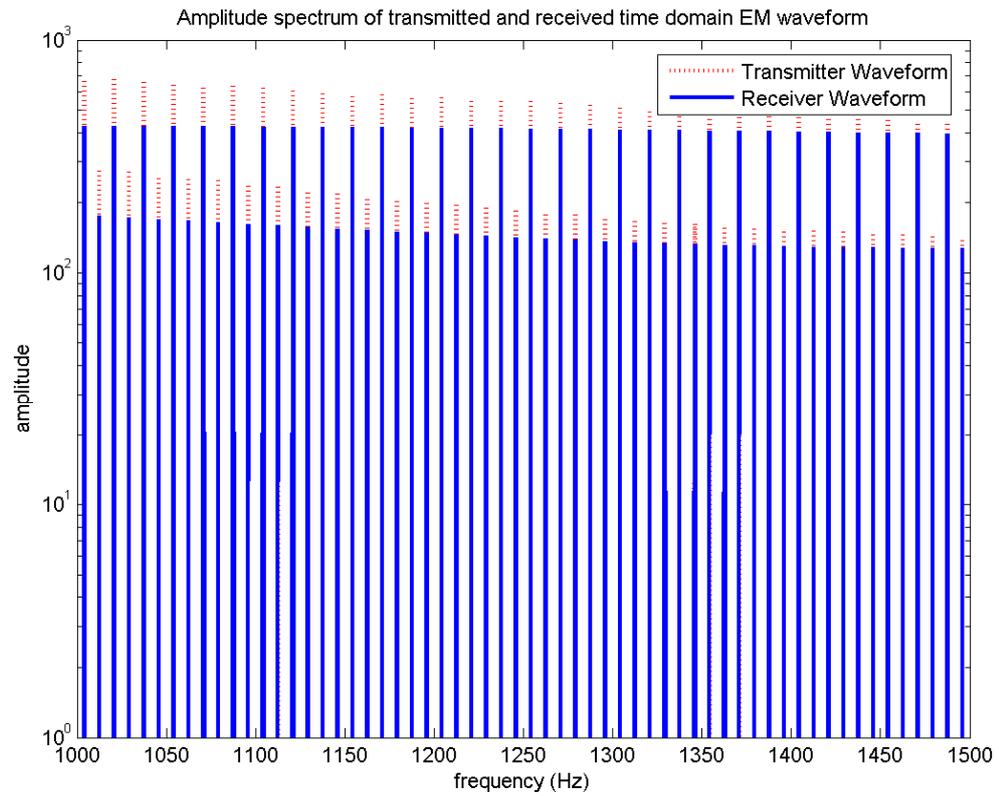


Figure 2.2. Amplitude spectrum of a synthetic time-domain EM transmitted waveform in the frequency domain (assuming a confined conductor). Duty cycle 50 %, base frequency 4.166 Hz, input current 1 A current, sampling frequency 10 kHz. Peaks in the amplitude spectrum occur at twice the base frequency.

2.1. Observation Noise Characteristics

Observation noise may be generated by the motion of the survey sensor in the Earth's magnetic field, the intrinsic electronic noise of the sensor, and the receiving systems hardware and data sampling software. Motion induced noise can be a large problem, as spurious signals can be induced in magnetic sensors by even slight movement of the sensor in the Earth's magnetic field. The Earth's field is approximately hundreds of thousands times larger than the typical fields measured in EM sounding. Therefore vibration induced by wind can often cause appreciable noise voltages known as wind noise or microphonics (Macnae *et al.*, 1984; Spies and Frischknecht, 1991).

In electronic circuits, total noise free performance is impossible to achieve as noise is a consequence of molecular motion. Intrinsic noise within the electronic components of a circuit include:

1. White noise: random energy containing all frequencies in equal proportions within the bandwidth, with random phases (Sheriff, 1999),
2. Pink noise: dominated by low frequencies, pink noise has the same distribution of power for each octave, with energy equal to $1/f$
3. Burst noise: noise switches between two or more discrete values at random time.

White noise is present in all resistors and semiconductors. Such noise results from the random thermal energy of electron conduction. The average power in a given bandwidth depends on temperature (Smith and Sheingold, 1969). Pink noise results from flicker or $1/f$ noise. Semiconductors exhibit pink noise due to random fluctuations in the number of surface re-combinations (Ryan and Scranton, 1984). Burst-noise, also known as popcorn noise, only occurs in poorly manufactured semiconductors. It is due to random on/off recombination action in the semiconductor material, leading to erratic switching of an affected device's current gain (Ryan and Scranton, 1984).

Most commonly the intrinsic noise of a time domain EM sensor is white noise. Synthetically this can be represented by generating an array of Gaussian distributed random numbers, normally distributed with a mean of zero.

2.2. Time Dependent EM Noise

Time-dependent EM noise is characterised by differences observed in recordings made at the same receiver station at different times. This type of noise comprises of spherics, noise of geomagnetic and ionospheric origin and cultural noise. Figure 2.3 shows a typical spectral plot for geomagnetic variations averaged over a long time. Geomagnetic noise is one of the major sources of noise in EM soundings. Below 6 Hz, the natural EM noise field is primarily geomagnetic and of ionospheric origin. Below 1 Hz, the signal arises mainly from within and outside the ionosphere as a result of complex interactions between plasma emitted from the Sun with the Earth's permanent magnetic field, known collectively as ultra low frequency (ULF) waves (previously known as micropulsations or magnetic pulsations). Oscillations with quasi-sinusoidal waveform are called pulsations continuous (Pc). The waveforms that are more irregular are called pulsations irregular (Pi). Each type is subdivided into frequency bands roughly corresponding to distinct phenomena (McPherron, 2002). Below 0.1 Hz, the amplitude of these signals is approximately inversely proportional to frequency (Figure 2.3). These signals are stronger during the morning and are also stronger in equatorial regions, so they are clearly related to the attitude of the sun. Over the range of 0.1 to 1 Hz, the steep fall off in the amplitude spectrum is caused by attenuation of the EM fields passing through the ionosphere, which is conductive over a broad range of frequencies (Spies and Frischknecht, 1991; San Filippo and Hohmann, 1983). ULF waves are relatively unimportant in EM soundings, except at very low frequencies. They can be an annoyance in IP surveys particularly Pc1 (Ritchie, 2003, pers. comm), which has period from 0.2 to 5 s (McPherron, 2002).

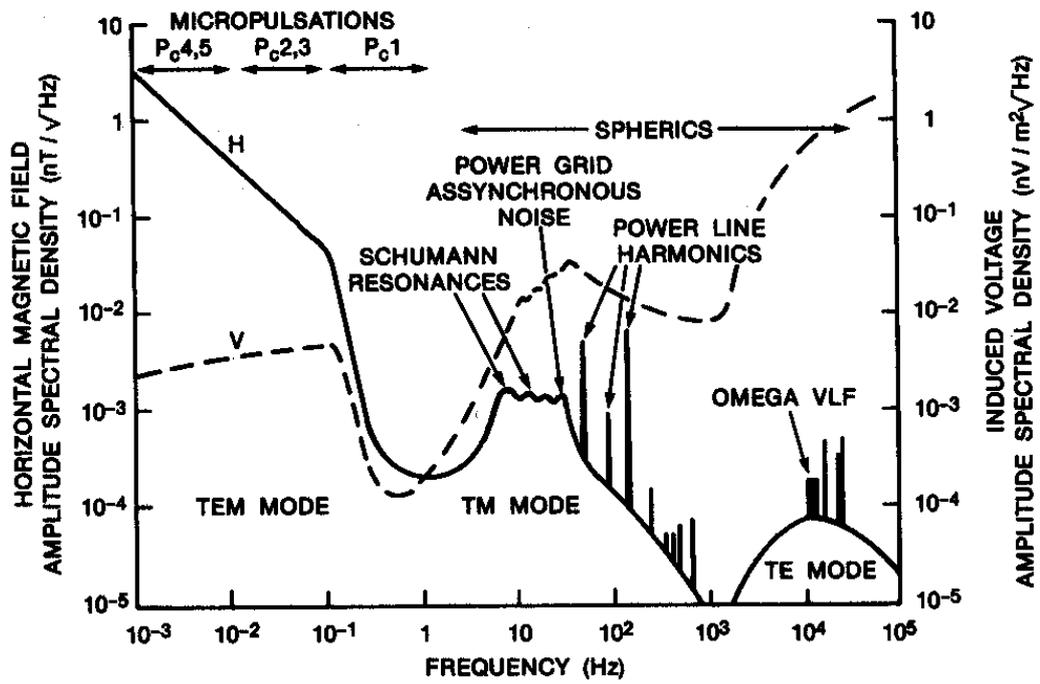


Figure 2.3. Generalized geomagnetic spectrum for horizontal magnetic field (H) and induced voltage (V) measurements reproduced from Spies and Frischknecht, 1991.

The main source of geomagnetic noise in the frequency range above 1 Hz is atmospheric lightning discharges around the Earth, referred to as spherics (Figure 2.3; Spies and Frischknecht, 1991). Spherics are impulsive events that occur over a broad range of frequencies, and they will be discussed in greater detail in Section 2.3. The extremely low frequency (ELF) components of spherics travel around the Earth's ionospheric cavity and interfere constructively and destructively, resonating at 8, 14, 20, 26 and 32 Hz, the Schumann resonances. The ionosphere has strong absorption between 500 Hz and 2.3 kHz, resulting in a low spectral density of geomagnetic noise in this region (Macnae *et al.*, 1984; Spies and Frischknecht, 1991).

Cultural noise arises mainly from power distribution grids. Depending on the country, the mains frequency has its fundamental at either 50 or 60 Hz but higher order harmonics can prove as equally damaging within the measured time series. While the United States, Canada Japan and parts of South America opted to run their power grids at 60 Hz, the majority of the world opted to run them at 50 Hz. This difference arose from the different types of generators developed during the introduction of electricity. In developed countries, the inertia of the power grid is very substantial and thus the frequency drift of the fundamental frequency is limited to less than 2% of its nominal value. In areas where the power grid is less developed and has minimal power factor correction, the fundamental frequency observed on the grid can vary up to 10 % of its nominal value (Macnae *et al.*, 1984; Spies and Frischknecht, 1991; Zonge and Hughes, 1991). Such cultural noise will be discussed in details in Section 2.4.

Communication signals are another major source of noise which occupy the higher portion of the electromagnetic spectra. The Navy uses Very Low Frequency (VLF) communication transmissions that produce strong signals in the range of 15 to 24 kHz (McNeill and Labson, 1991). The advent of global positioning systems (GPS) brought an end to military positioning signals such as the United States OMEGA VLF communications, and thus they are no longer a noise source but it remains important to be aware of these signals when processing older data sets. Radio and radar stations broadcast at much higher frequencies,

but may often overload induction EM systems that do not incorporate sufficient filtering and shielding. They may also preclude the possibility of making soundings (Spies and Frischknecht, 1991).

2.3. Spherics Characteristics

Spherics energy is found within the range of 5 to 25 kHz and is the principal noise source limiting the depth of investigation and resolution for IP and EM surveys. Worldwide spherics contribute to the average background noise power in the frequency ranges below 500 Hz and between 2.3 and 10 kHz (Spies and Frischknecht, 1991). The major global thunderstorm centres are located in Brazil, central Africa and Malaysia, which average over 100 storms per day. During the day, the center of the storm activity shifts westward with the sun, resulting in fairly consistent levels of spheric activity. Near storm centers, the peak of the activity occurs in the early afternoon. There is a pronounced seasonal variation, with magnitudes being one order lower during winter months. The level of spheric activity decreases away from the equator as the major sources are in the tropical regions. Lightning discharges are estimated to occur with a frequency of about 100 per second (Jewell and Ward, 1963). The EM noise is quasi continuous below 300 Hz, but predominantly impulsive in nature at higher frequencies (Spies and Frischknecht, 1991)

Over a horizontal layered Earth, the magnetic field from distant sources will be nearly horizontal. Typically when EM soundings are made, the noise component of the vertical magnetic field will be 6 to 10 times smaller than the noise component of the horizontal field (Spies and Frischknecht, 1991). Figure 2.4 shows the magnitude of the vertical and horizontal magnetic fields for spherics recorded in Darwin in the summer of 1983. Details of the survey can be found in Appendix 1. If the DC shift is removed, and taking the cumulative sum of the squared amplitudes of the vertical and horizontal magnetic fields the horizontal component is observed to be 7.5 times the magnitude of the vertical component. Figure 2.5 shows the amplitude spectrum of the vertical and horizontal magnetic fields at Darwin for the same recording period.

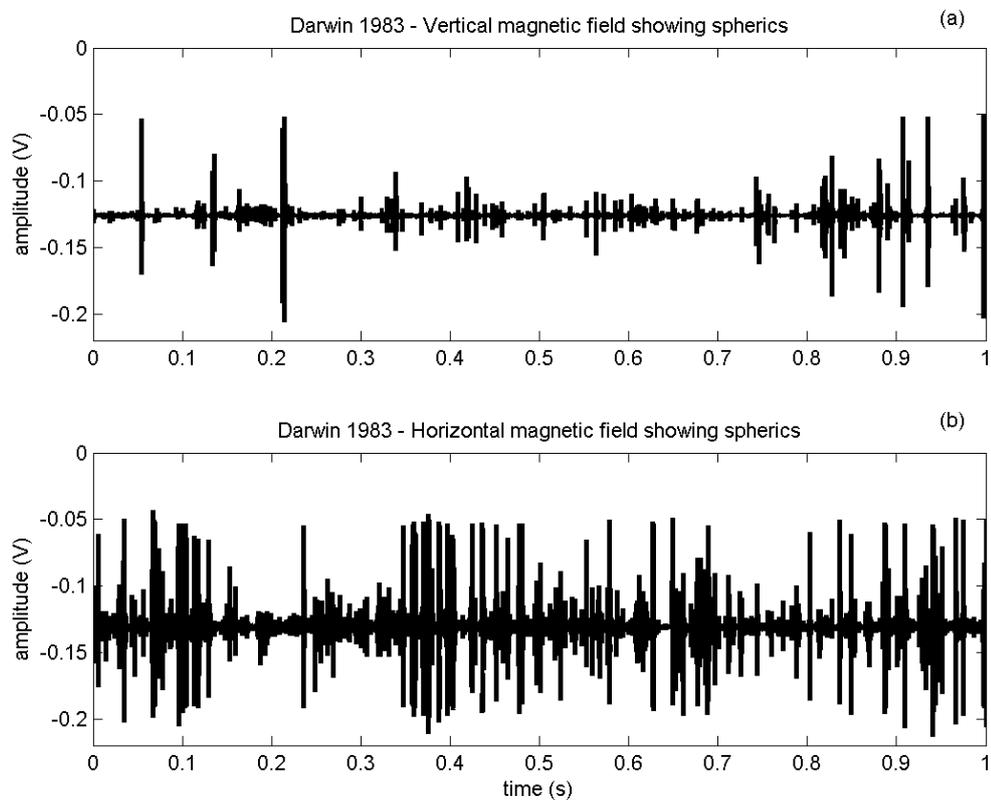


Figure 2.4. Spherics activity recorded in Darwin, the summer of 1983, data courtesy of J. Buselli, CSIRO.
(a) Vertical component of magnetic field showing spherics. (b) Horizontal component of magnetic field showing spherics. Notice the impulsive nature of spherics. The magnitude of the horizontal component of the magnetic field is 7.5 times the vertical component. Details of this data recording are in Appendix 1.

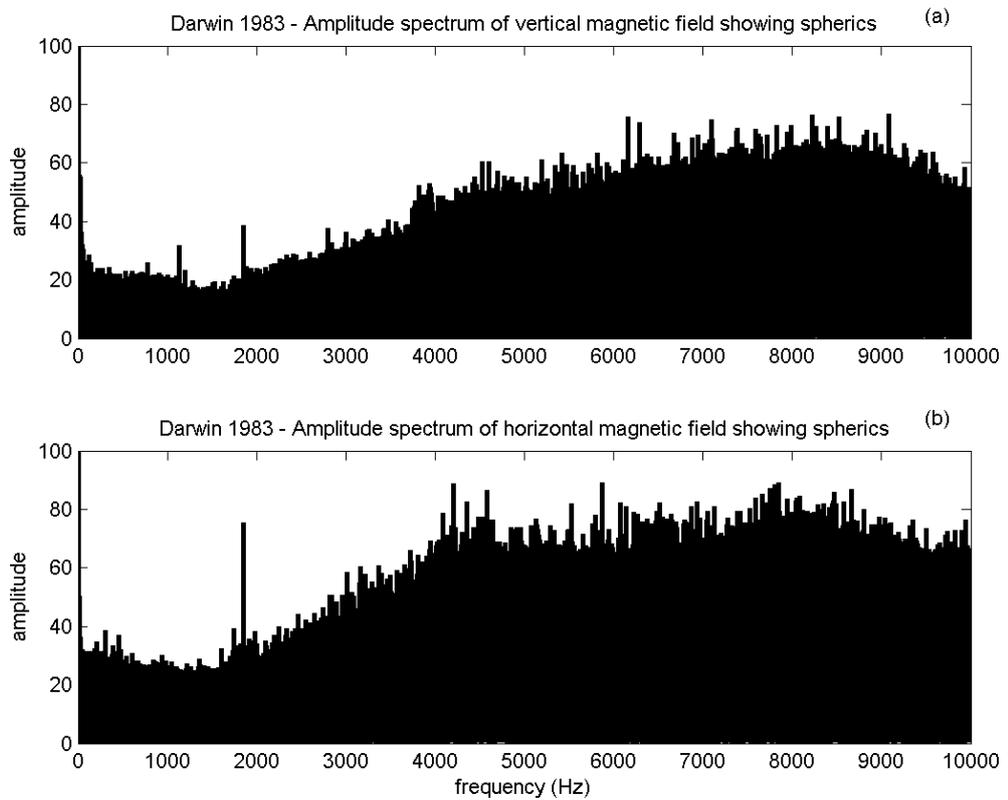


Figure 2.5. Amplitude spectrum of spherics activity recorded in Darwin summer of 1983, data courtesy of J. Buselli, CSIRO.
(a) Amplitude spectrum of the vertical component of magnetic field showing spherics.
(b) Amplitude spectrum of the horizontal component of magnetic field showing spherics. The spectrum indicates that the storm activity was nearby as the high frequency components are dominant and have not been absorbed. These plots also show the broadband nature of spherics as a noise source. Details of this data recording are in Appendix 1.

2.3.1. Synthetic Spherics

A heuristic approach was taken to generating synthetic spherics. Figure 2.6 shows an example of a spheric event recorded in Darwin in the summer of 1983. Details of the data recording are in Appendix 1. A lightning strike can generally be described as being a minimum phase impulse response. A minimum phase signal has the energy concentrated at the onset. When there are multiple global sources and spheric events, the minimum phase assumption may no longer hold as the signals can merge. Very strong transients usually come from nearby or intermediate distances < 1000 km. Figure 2.6 (b) shows that the spheric response is comprised of a multitude of frequencies. In this case the dominant frequencies are centered at 5 kHz and 8.5 kHz. I theorized that a spheric event can be approximated by exponentially damped sinusoids as:

$$Sph = \left(\frac{t}{5}\right)^4 \exp\left(A\frac{t}{5}\right) \left[\sin\left(B\frac{t}{5} + C\right) + \sin\left(D\frac{t}{5} + E\right) + \sin\left(F\frac{t}{5} + G\right) \right] + H\left(\frac{t}{5}\right)^3 \exp\left(I\left(\frac{t}{5} + J\right)\right) \left[\sin\left(K\frac{t}{5} + L\right) + \sin\left(M\frac{t}{5} + N\right) + \sin\left(O\frac{t}{5} + P\right) \right]. \quad (2.2)$$

Using Equation (2.2) and field data recorded at Darwin, the constants A through P were solved using a Matlab Mathworks Inc. function file for unconstrained, non-linear optimisation. Figure 2.7 shows the original spheric events and the modeled spheric events using Equation (2.2).

Equation (2.2) can then be simplified to:

$$Sph = \left(\frac{t}{5}\right)^4 \exp\left(A\frac{t}{5}\right) \sin\left(R^*B\frac{t}{5} + R^*C\right) + H\left(\frac{t}{5}\right)^3 \exp\left(I\left(\frac{t}{5} + R^*J\right)\right) \sin\left(R^*K\frac{t}{5} + R^*L\right), \quad (2.3)$$

where R^* is a random number between zero and one, with uniform distribution; R^{**} is a random number, normally distributed with a mean of zero, variance one and standard deviation one; and A through G are constants used to define the spheric response.

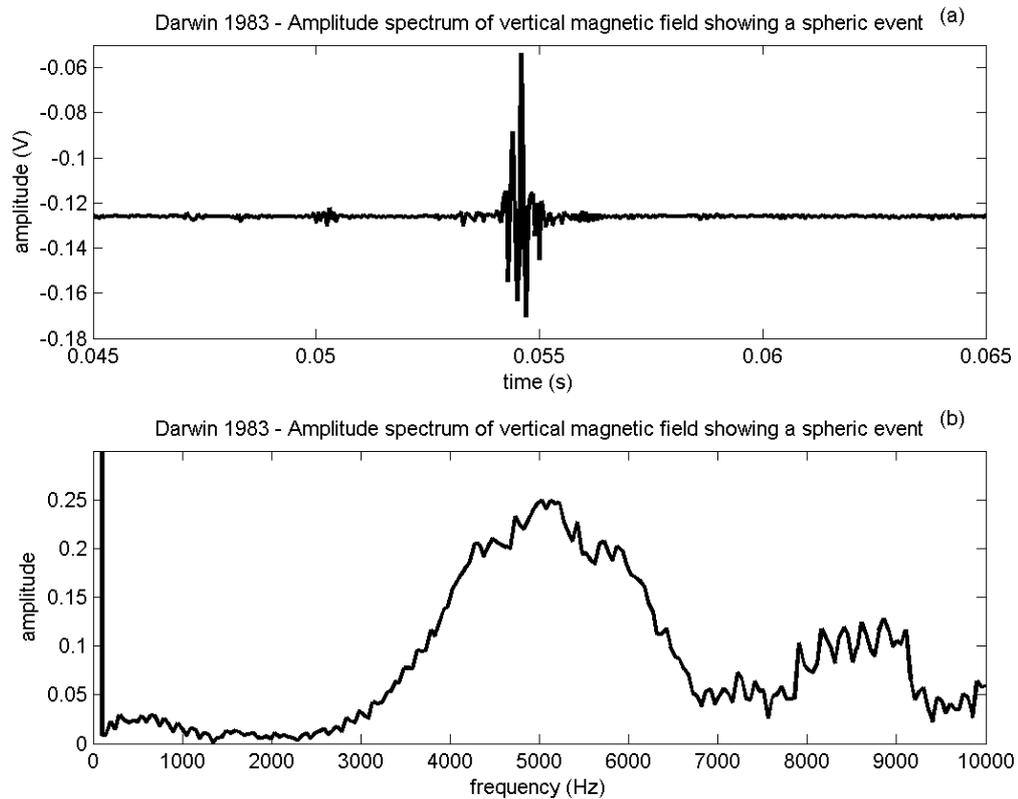


Figure 2.6. Spheric event recorded in Darwin, the summer of 1983, data courtesy of J Buselli, CSIRO.

(a) Vertical component of magnetic field. (b) The amplitude spectrum of the spheric event. The time series data show the impulsive nature of spherics. The amplitude spectrum indicates the frequency of the spheric event is concentrated at around 5 kHz and 8.5 kHz. Details of this data recording are in Appendix 1.

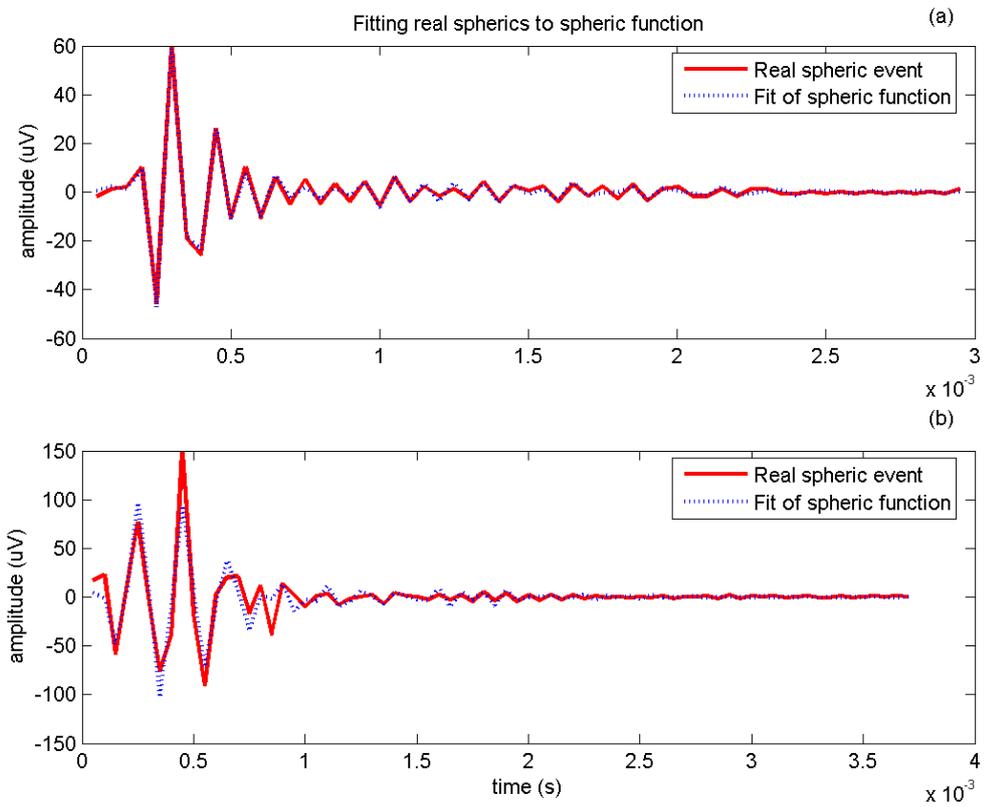


Figure 2.7. Unconstrained non-linear optimisation to fit real spherics.
Example of two spherics captured from the Darwin vertical component data with fit
for Equation (2.2). Data courtesy of J. Buselli, CSIRO.

2.4. Cultural Noise Characteristics

The powerline voltage waveform is generally sinusoidal. However, the current waveform is often complicated. Motor loads operate non-synchronously and produce more damaging, but weaker non-stationary components, such as switching transients, sidebands and sub-harmonics. The simple switching of current loads and the electronic cycle by cycle switching of rectifiers in many power systems produce broadband transients and high frequency harmonics. Electric rail roads in Europe are often a source of large amplitude, 13 2/3 Hz noise (Macnae *et al.*, 1984; Spies and Frischknecht, 1991; Zonge and Hughes, 1991). In surveys located at mine sites, the Person Emergency Device (PED) may contaminate the received signal. PED transmissions occur over the range of 362 to 386 Hz. PED signals are not sinusoidal, and thus have some power at higher harmonics; notably the 3rd, 5th and 2nd (Duncan, 2002).

Generally the powerline EM field consists of a series of Dirac-delta functions at the mains frequency and its harmonics, then a multitude of much weaker and more contaminating non-stationary components, Figures 2.8 and 2.9 show examples of data contaminated with harmonic noise. Figure 2.8 is an example of data collected in an urban environment (electrical data collected at Curtin University, details in section Appendix 1) and Figure 2.9 is an exploration example (IP data collected with Mount Isa Mines' distributed acquisition system (MIMDAS), details in section Appendix 1). Where powerline harmonics are modulated, the time series shows a variation in amplitude, known as the modulation envelope, seen in Figure 2.8 and 2.9. The amplitude spectrum modulation is indicated by multiple spikes instead of clear spectral lines, Figure 2.8 and 2.9.

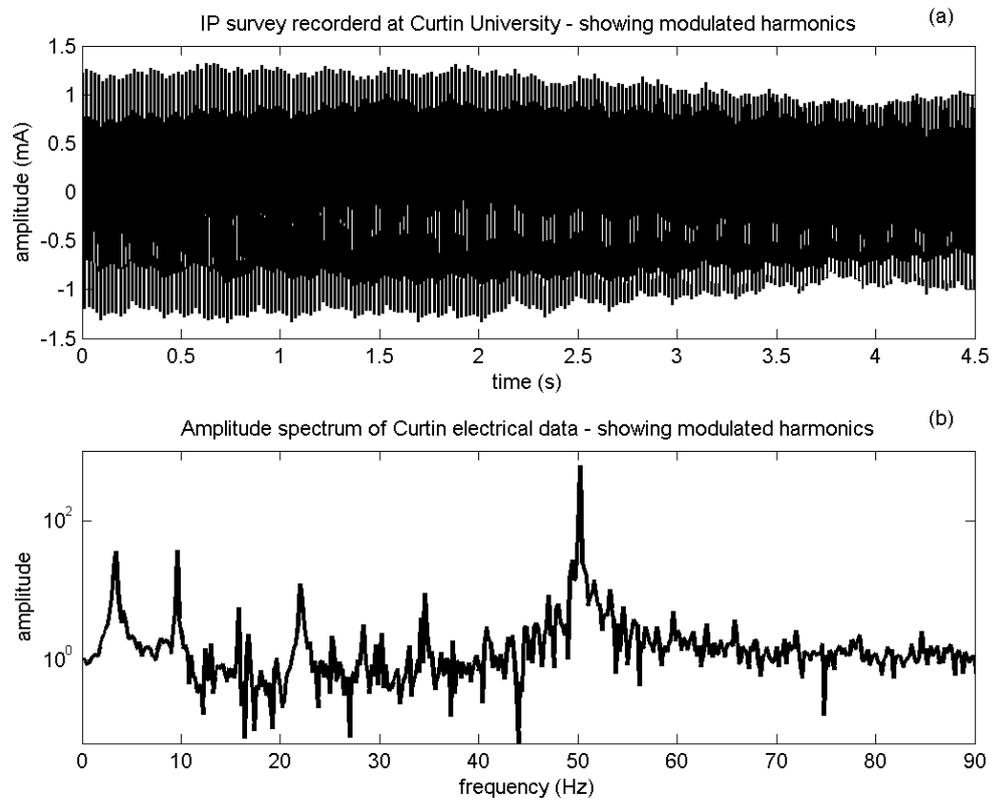


Figure 2.8. IP survey at Curtin University showing modulated harmonics. (a) Section of time series. (b) The amplitude spectrum of IP survey. The time series shows varying overall amplitude, indicating harmonic modulation. Peaks in the amplitude spectrum at 3.4 and 9.65 Hz are due to the received signal occurring at increments of the base frequency 3.125 Hz. The peak at 50 Hz is due to the powerline harmonics. Details of this data recording are in section Appendix 1.

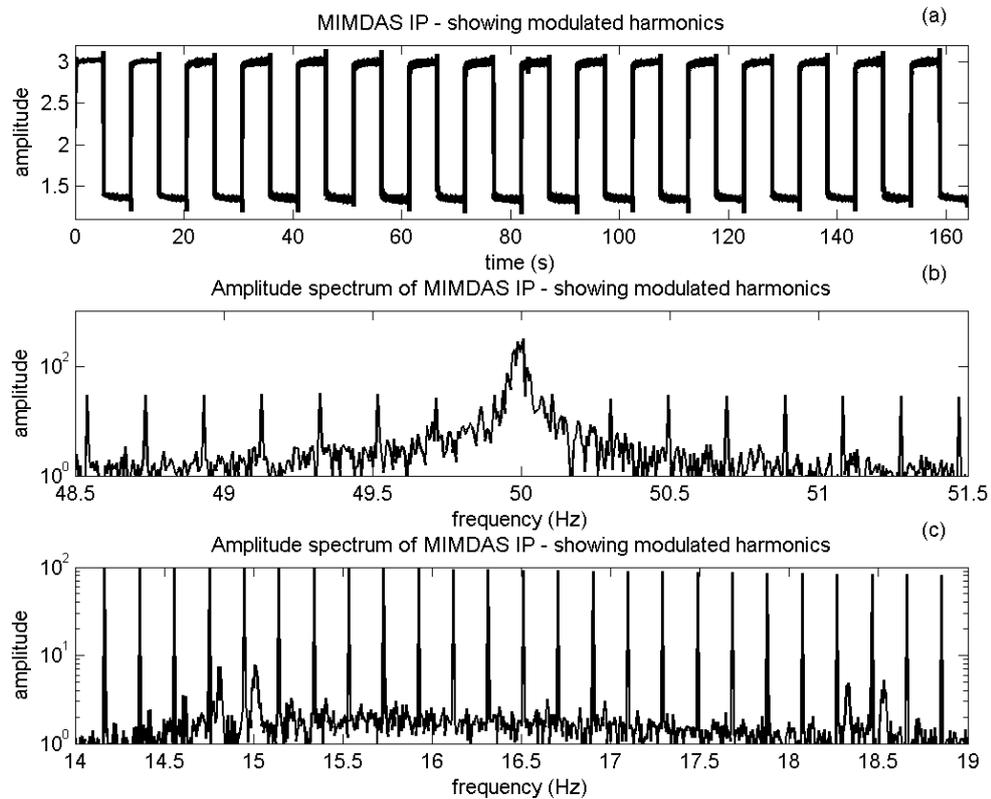


Figure 2.9. MIMDAS IP data showing modulated harmonics and weak non-stationary components, data courtesy of T. Ritchie, Xstrata Pty Ltd.

(a) Section of time series. (b) Sections of the amplitude spectrum. Peaks in the amplitude spectrum for the IP received waveform occur at twice the base frequency of 0.0997 Hz. The peak at 50 Hz is due to powerline harmonics. Peaks in the amplitude spectrum at 14.8, 15, 18.3 and 18.5 Hz are due to weak non-stationary components. There are multiple peaks at 14.8 Hz and 50 Hz, indicating modulation. Details of this data recording are in section Appendix 1.

2.4.1. Synthetic Modulated Powerline Harmonics

A heuristic approach was taken to generating synthetic amplitude modulated powerline harmonics. A stationary sinusoid can be represented by:

$$S(\vec{m}, t) = \sum_{i=1}^n a_i \sin(2\pi f_i t) + b_i \cos(2\pi f_i t), \quad (2.4)$$

where f_i is the frequency of the sinusoid, t is time, a_i and b_i are the amplitudes of the sine and cosine components (the ratio of these is directly related to the phase of the sinusoid), and n is the number of sinusoid functions needed to model the sinusoidal noise.

Harmonic noise with amplitude modulation can be represented as a summation of sinusoids with a sinusoidal function defining the modulation envelope with the harmonic component at a single sensor, it can be then be given as;

$$S(\vec{m}, t) = \sum_{i=1}^n a_i \sin(2\pi h_i t) \sin(2\pi f_i t) + b_i \sin(2\pi h_i t) \cos(2\pi f_i t), \quad (2.5)$$

where h_i is the frequency of the modulation envelope. Figures 2.10 and 2.11 show modulated harmonics produced by Equation (2.5)

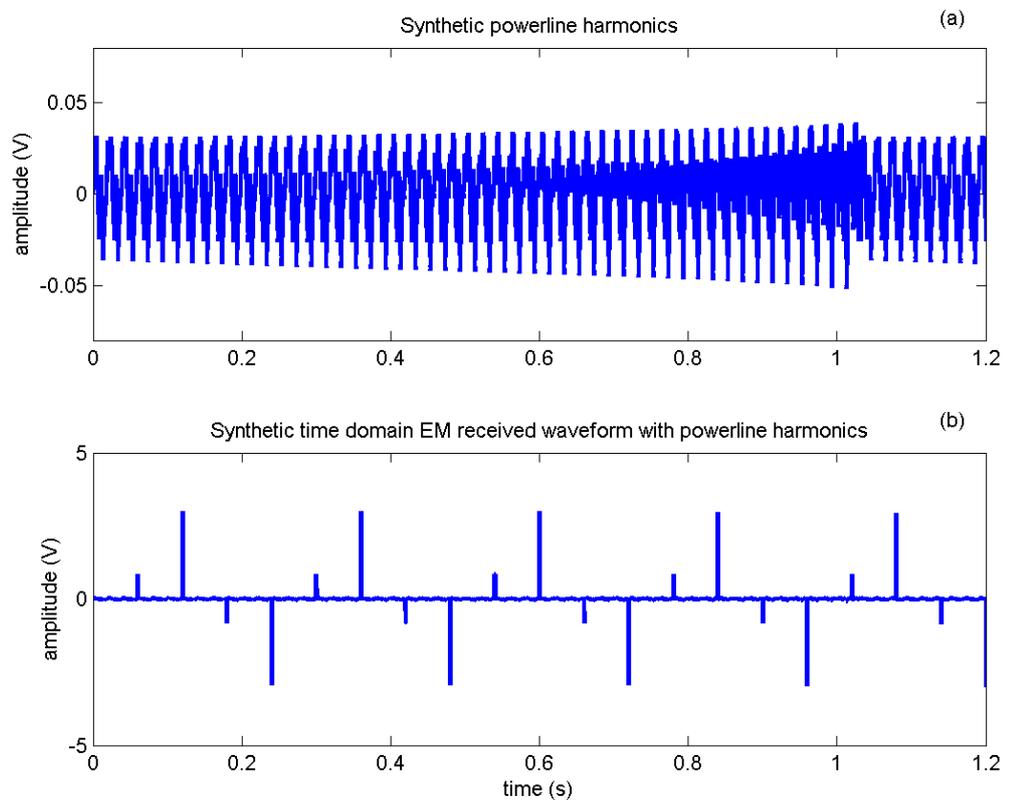


Figure 2.10. Synthetic data showing modulated harmonics.
(a) Synthetic modulated harmonics generated using Equation 2.5 (b) Synthetic modulated harmonics with synthetic a time domain EM waveform, generated using equation 2.1 and 2.5. The time series show varying amplitude, indicating harmonic modulation.

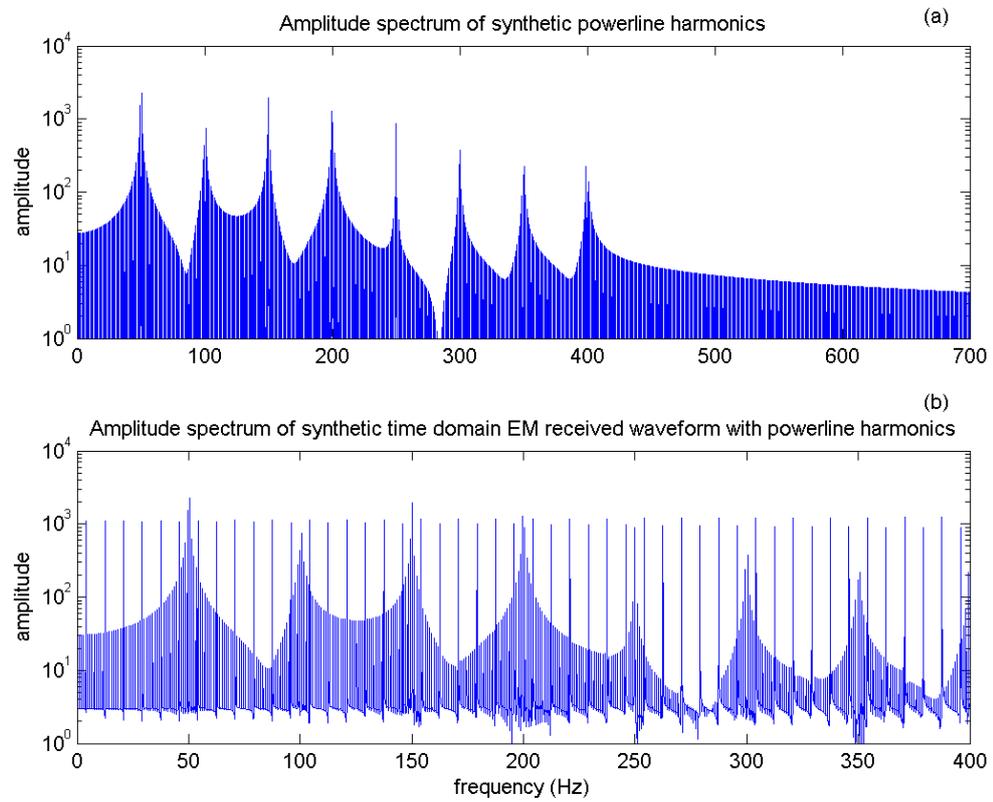


Figure 2.11. Amplitude spectrum of synthetic data showing modulated harmonics. (a) Synthetic modulated harmonics generated using equation 2.5(b) Synthetic modulated harmonics with synthetic time domain EM waveform, generated using equation 2.1 and 2.5. Peaks in the amplitude spectrum at 50, 100, 150, 200, 300 350 and 400 Hz due to the powerline harmonics.

3. Signal Processing

There has been comprehensive research into noise sources and noise reduction techniques for electrical methods used in geophysics. Chapter two reviewed the characteristics of observation noise and time dependent EM noise, primarily spherics and powerline harmonics. At present, a number of data processing steps can be performed on the data, to improve the signal to noise (S/N) in the presence of these noise sources:

- 1) Stacking (averaging) the signal over time and increasing the transmitted field. The target signals increase linearly with the magnitude of the inducing field and is phase coherent with it. Therefore, relative noise amplitude is decreased by both averaging in time and increasing the transmitted field. Instrument averaging or stacking rejects random, events such as electronic noise and distant spheric activity.
- 2) Pruning (data rejection) of non-stationary intense transients, such as local spherics or powerline transients.
- 3) Modification to weighting of a stack to improve the rejection of line spectrum noise.
- 4) Signal modification by pre-emphasis/de-emphasis to accommodate the non-white character of the spheric noise spectrum.

These steps are discussed in greater detail in Buselli and Cameron, 1996; Macnae *et al.*, 1984; and McCracken *et al.*, 1986b.

3.1. Pre-whitening and Deconvolution

Natural EM noise originates mainly from spherics. Spherics energy is not uniformly distributed through the system bandwidth. It is concentrated at the high frequency end of a typical EM system bandwidth (Figure 2.5). It is possible to redistribute the power in the transmitted spectrum and compensate in the receiver by filtering the signal and noise. Such techniques are known as pre-emphasis pre-whitening and deconvolution. The main obstacle to their use in EM is the necessary preservation of phase characteristics in filters (Macnae *et al.*, 1984). When applying a filter to the transmitter current, with a realisable

inverse, if the inverse filter is applied to the received signal, the earth response will be the same while the noise is filtered via the inverse filter. Optimisation (or pre-emphasis) is a complex process for any EM system, as the noise spectrum is far from constant and the process is limited by the achievable voltage and current limits of the transmitter. Optimisation can be formalised and the conditions for this are that the noise spectrum is known and constant, the relative precision on different channels is stated, and the norm of the transmitter is fixed (Macnae et al., 1984 and Duncan, 2002).

Spiking deconvolution has application in seismic processing for whitening the amplitude spectrum of the seismic data prior to pre-stack processing (Yilmaz, 2001). I theorized that spiking deconvolution could be used to minimise spheric events from the received waveform. A synthetic spheric event is produced using Equation 2.3 (Figure 3.1) and a deconvolution operator is created by taking the inverse of the spheric event (Figure 3.2). In this example two spheric examples are shown. The first is a minimum phase spheric event and the second is a mixed phase spheric event. A spheric event will be minimum phase if the recording of the event is taken close to the source of the event. The phase of a spheric event will alter as it travels from its source, and if there are multiple spheric events within the same time window, they interfere with each other and thus also reflect this interference on their combined phase. The deconvolution operator is convolved with the original data. Ideally, spheric events would become Dirac-Delta functions (Figure 3.3(a)). This approach, however, proved unsuccessful as it requires the assumption that the spheric event is minimum phase. When this condition fails, the convolution of the deconvolution operator with the time series does not produced the Dirac delta function see Figure 3.3 (b).

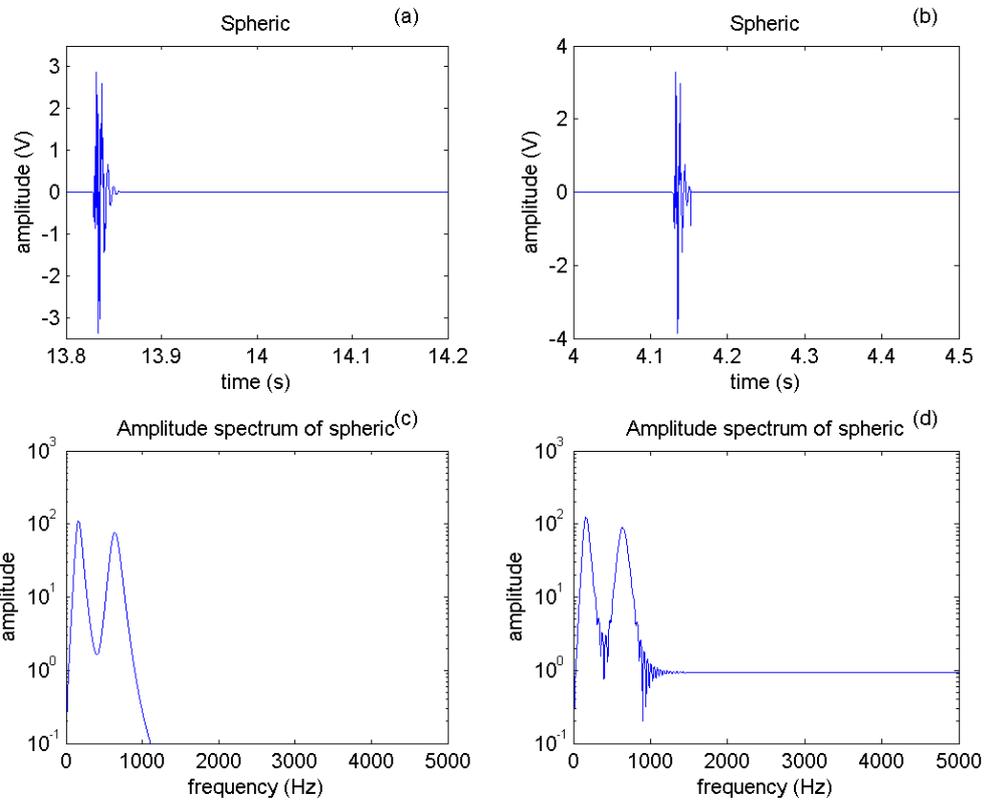


Figure 3.1. Synthetic spherics

(a) Minimum phase spheric event (b) Mixed phase spheric event (c) Amplitude spectrum of minimum phase spheric event (d) Amplitude spectrum of mixed phase spheric event. The two synthetic spheric events have similar amplitude spectrums and though their phase spectrums differ. The phase spectrum controls the shape of the waveform.

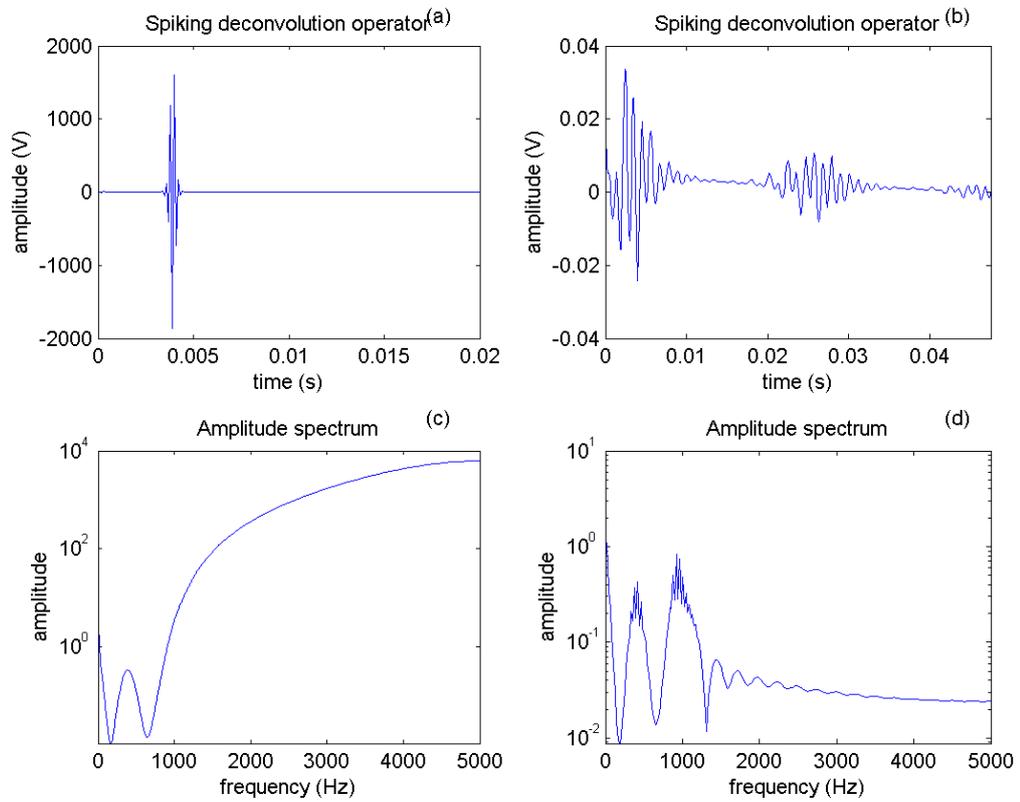


Figure 3.2. Spiking deconvolution operator derived from spheric events. (a) Operator derived using a minimum phase spheric event (b) and its amplitude spectrum (c) Operator derived using a mixed phase spheric event and its amplitude spectrum (d).

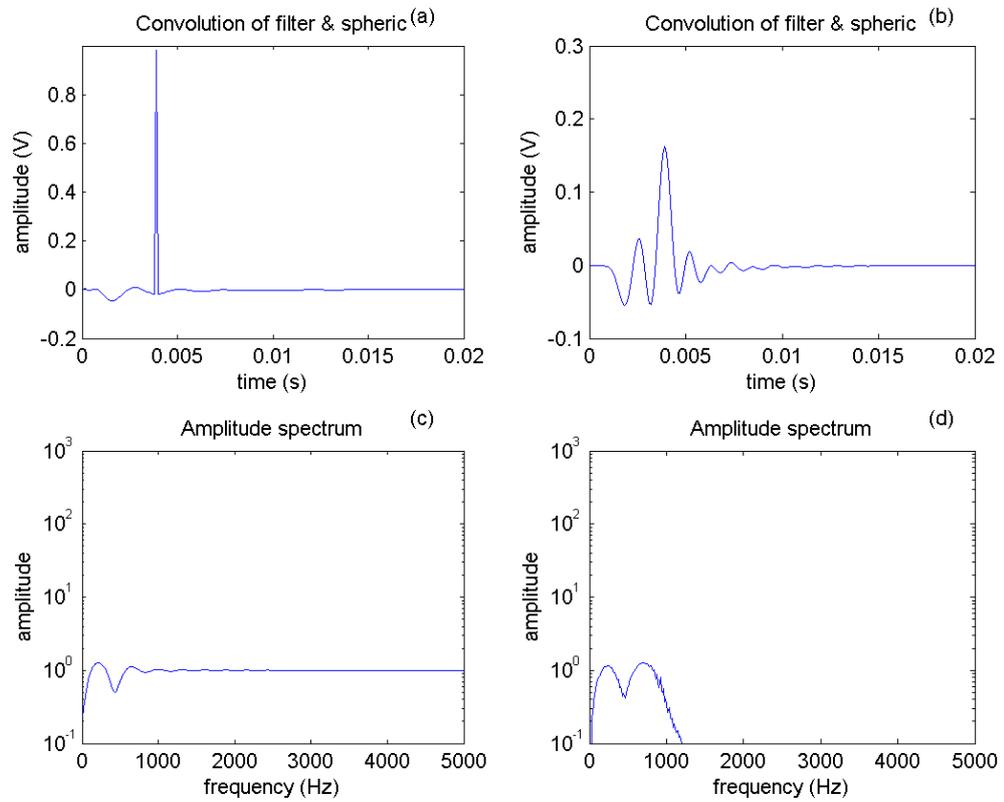


Figure 3.3 Convolution of spiking deconvolution operator with spheric. The desired output is the Dirac delta function (a) when using a mixed phase spheric to derive the deconvolution operator the operation fails (b). The desired output is a flat amplitude spectrum (c) the effect of the spheric has not been removed from (d).

3.2. Stacking

Stacking is the collective binning of a geophysical measurement in one location. The number of stacks is chosen to provide a balance between available time for acquisition, available computer memory (in the case of full waveform measurements) and the number of necessary measurements to produce repeatable measurements. In most cases, data are stacked by the instrument during recording using a boxcar integrator type approach. However, the boxcar integrator is essentially a single channel device and the detection of geophysical signals is generally done with a number of gates activated sequentially, usually with six to fifty channels. Gate widths and positions are set to fully recover the expected signal (Becker and Cheng, 1987). Data are gated by averaging groups of adjacent samples. The data groups are coherently sorted into channels with each channel containing a subset of the acquired data. The acquisition is synchronised with the transmitter so the delay between the commencement of the signal cycle and the positioning of the data group is constant. The sorting (or windowing) process is followed by rectification, consistent with the signal polarity changes and averaging of the rectified data. Usually these two processes are combined into a single digital filtering operation (Becker and Cheng, 1987). In the case of electronic rectification it is difficult to achieve exact equality in processing the plus and minus half cycles. The resulting bias may be eliminated by interchanging processing paths in the receiver at regular intervals during stacking. The output is then a sum of several smaller stacks, rather than one continuous stack. This procedure has no effect on the spectral sensitivity for uncorrelated noise (Macnae et al., 1984).

Field data normalised with respect to the transmitter current or calculated primary field and receiver coil moment. In certain cases, the finite transmitter turn off is also corrected (Nabighian and Macnae, 1991). For certain types of non-stationary noise, such as spherics or coherent noise caused by powerlines, it is possible to increase S/N ratios above those obtained by simple stacking for an equal time by use of techniques such as tapered stacking, randomised stacking or pruning (data rejection) of non-stationary intense

transients, such as local spherics or powerline transients (Macnae et al., 1984; Spies and Frischknecht, 1991; Nabighian and Macnae, 1991).

Tapered stacking, which is the convolution of the received signal with a tapered window, will increase the central acceptance peak and reduce the side lobe amplitude in comparison to a rectangular filter in the frequency domain (Macnae et al., 1984). If noise is white, simple stacking will result in a better S/N ratio than tapered stacking. However reduction of side lobes may be favourable if strong monochromatic noise is present (Macnae et al., 1984). Powerline noise, high frequency VLF interference and any other narrow bandwidth, coherent interference, is removed far more effectively with tapered stacking. Powerline filtering will improve at higher base frequencies (Duncan, 2002).

Large amplitude transient noise, such as local spheric activity or nearby powerline transient, often occur only occasionally on the time scale of a single measurement. Such noise is not stationary with respect to system averaging time, and it is easy to show that S/N ratio after stacking can be improved if, during the time of the transient, all input is totally rejected rather than added to the stack.

The simplest criterion for identifying a transient is to have an amplitude threshold trigger. The threshold is best set according to the conditions at the survey site. Some spheric and powerline transients have a precursor or gradual onset. To reject this part of the non-stationary noise, it is necessary to reject the immediate samples and previous half cycle of data. Rejection then continues for a set time window which depends on the amplitude and duration of the overload. It will generally include at least one cycle after the threshold is no longer exceeded to prevent any tail of the transient from biasing the stacked data. Only spikes with an amplitude large enough to be recognised as an outlier can be rejected. Therefore, spherics noise with an amplitude equal or less than the amplitude of the noise from other sources will not be recognised as a spheric and cannot be eliminated. It is essential to reject equal numbers of plus and minus cycles to avoid introducing zero bias (Macnae et al., 1984, Buselli and Cameron, 1996, and Buselli *et al.*, 1995).

McCracken *et al.* (1984) state it is possible to remove a spheric from the data stream and replace it with synthetic data, rather than nulling the time-series. This introduces less noise into the stacked data. Where both strong transients and narrowband noise are present, taking random stacks improves the data fidelity, except for noise at odd harmonics of the base frequency. To avoid zero bias and sensitivity to the even harmonics of the transmitter base frequency, it is necessary that equal numbers of plus and minus cycles to be stacked (Macnae *et al.*, 1984). It is possible to apply both pruning and tapered stacking to the same data set, but because pruning is a step function, it reintroduces the power side lobes that tapered averaging is designed to remove, and may also lead to bias in the output (Macnae *et al.*, 1984).

Iterative stacking in conjunction with single value decomposition (SVD) inversion for harmonic removal (Section 4.1.1) and robust statistics for attenuation of observation noise was the most effective. This will be explored in detail in Section 3.3 on robust statistics and in the array processing chapter (Chapter 5).

A method is reviewed where stacking was used to get a robust estimate of the target signal when the signal was highly contaminated with noise, as shown in Figure 3.4. This estimate can be used in other processing applications, such as removing an estimate of the target signal prior to estimating the frequencies of harmonics before SVD inversion. To determine a robust estimate of the target signal, a low pass filter was used to limit the signal to less than 40 Hz, followed by rectangular stacking with logarithmic binning (Figure 3.4). Rectangular stacking is used as opposed to tapered stacking, as the high frequency noise from the harmonics (and possible spherics) were previously removed via low pass filtering.

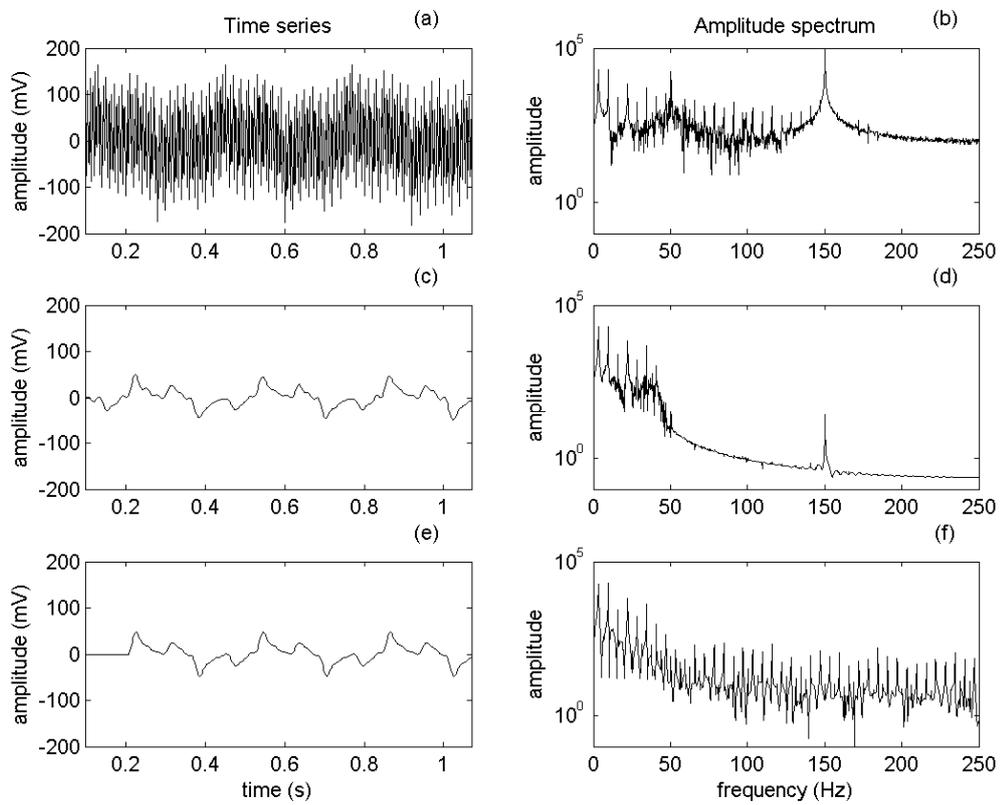


Figure 3.4. Stacking for a robust estimate of the received signal
 The received signal (a) and its amplitude spectrum (b) have been low pass filtered to < 40 Hz producing signal (c) and amplitude spectrum (d). The resultant signal (c) is then rectangular binned and logarithmically sampled, producing (e) and its amplitude spectrum (f). This binning provides a good estimate of the target signal. The target signal (e) may be subtracted from the received signal (a) and the resultant noise train used for estimating the frequencies of powerline harmonic prior to SVD inversion.

3.3. Robust Statistics

Recent research into spherics elimination has been conducted by Buselli and Cameron (1992), Buselli and Cameron (1996) and Buselli *et al.* (1993). They researched different ways of stacking noisy data, stacking using robust estimators such as median, trimmed mean and a range of M-estimators. They created an algorithm to recursively calculate a robust mean and the standard error in each time window measurement. The algorithm was implemented in the SIROTEM system and improved spherics noise rejection in time domain EM measurements.

Iterative stacking is used in this study. Powerline harmonics are removed via non-linear SVD inversion (Section 4.1.1). Then the remaining array correlated noise and intrinsic random noise are attenuated via a series of iterative stacking using robust statistics. Initially, a stack of the rectified transmitted signal is removed post harmonic removal. Signal rectification is the process of converting the transmitted signal to the same polarity (Figure 3.5). The rectified stack of the transmitted signal is passed through a low pass (30 Hz) Hanning filter (Figure 3.6), giving an estimate of the target signal. This can be removed from the rectified time-series (Figure 3.5(c)), giving an estimate of the noise within a stack (Figure 3.7). This noise stack can then be weighted and removed from the signal, thus attenuating the low frequency noise content (Figure 3.7). The remaining signal can then be stacked again, and an alpha-trimmed median binning applied to the stack. Alpha-trimmed median binning is the process of binning data and then excluding an equal amount of the upper and lower data in the bin prior to stacking, thus attenuating the high-frequency noise content (Figure 3.8). All this applied iteratively reduces both the array correlated and the random intrinsic noise. This method has the advantage of not limiting the signal to the intrinsic noise of the sensor, as with the remote reference system.

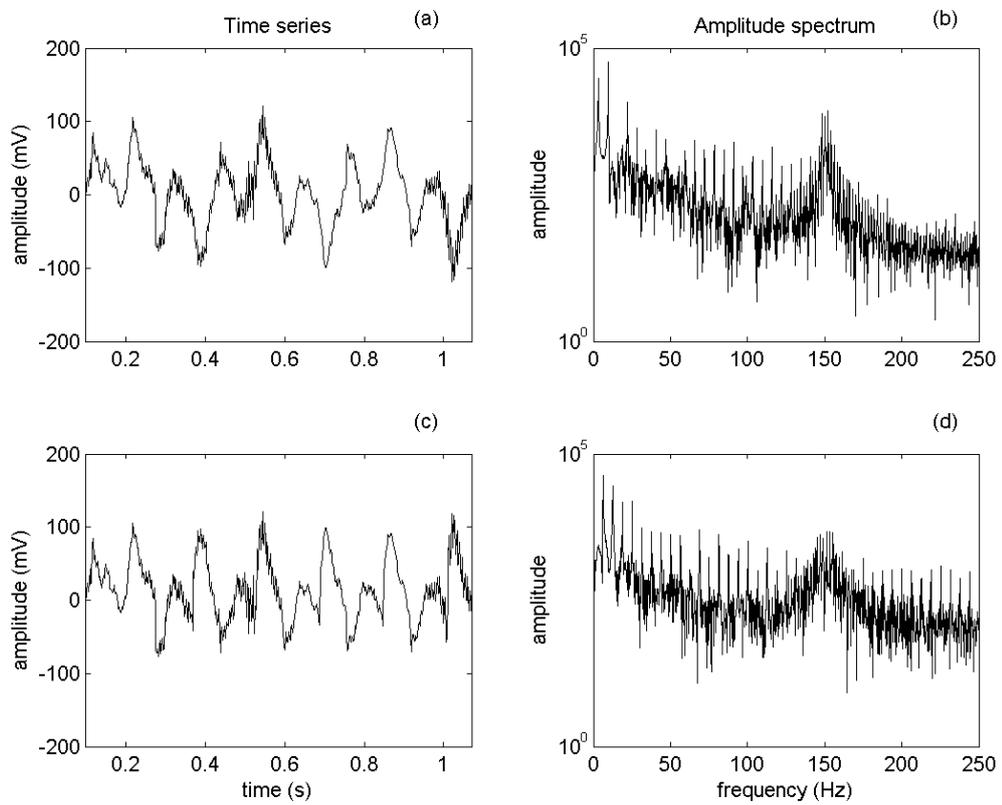


Figure 3.5. Signal rectification.

Time series (a) and its amplitude spectrum (b) is rectified to produce signal (c) and its amplitude spectrum (d). Rectification takes place by multiplying negative half-cycles by negative one.

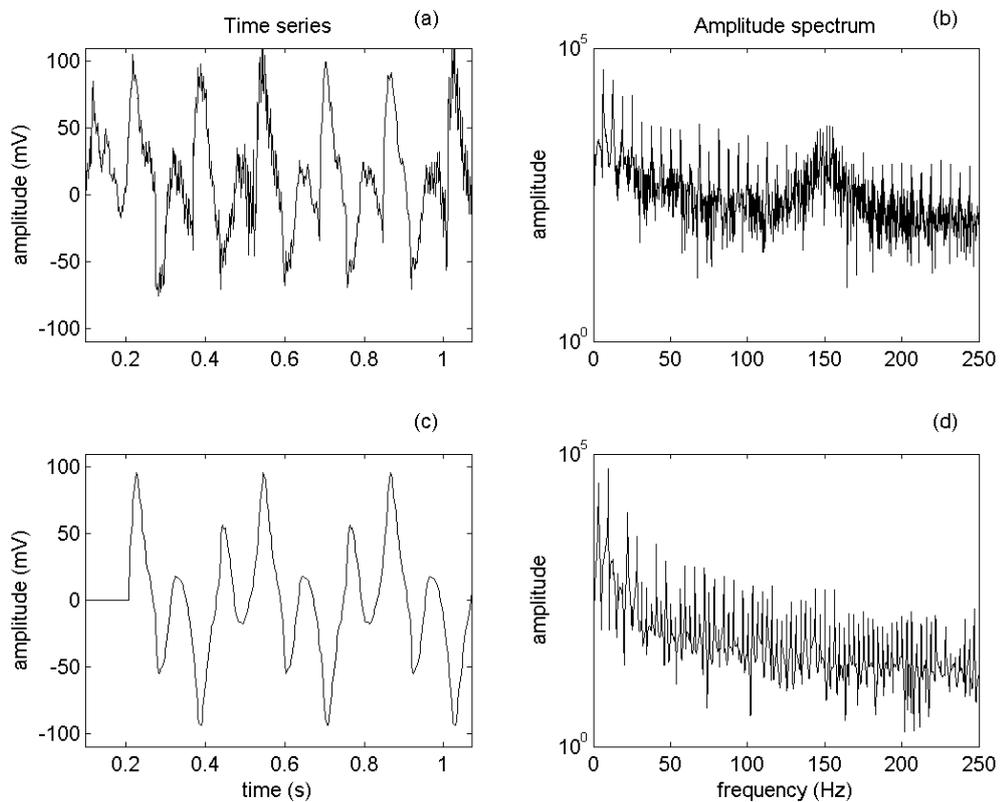


Figure 3.6. Stacking prior to robust statistics and iterative stacking. The rectified signal (a) and its amplitude spectrum (b) have been binned and convolved with a low pass (30 Hz) Hanning filter, then logarithmically sampled, producing time series (c) and its amplitude spectrum (d). This binning provides a good estimate of the target signal for each sensor, which is removed prior to robust statistics and iterative stacking.

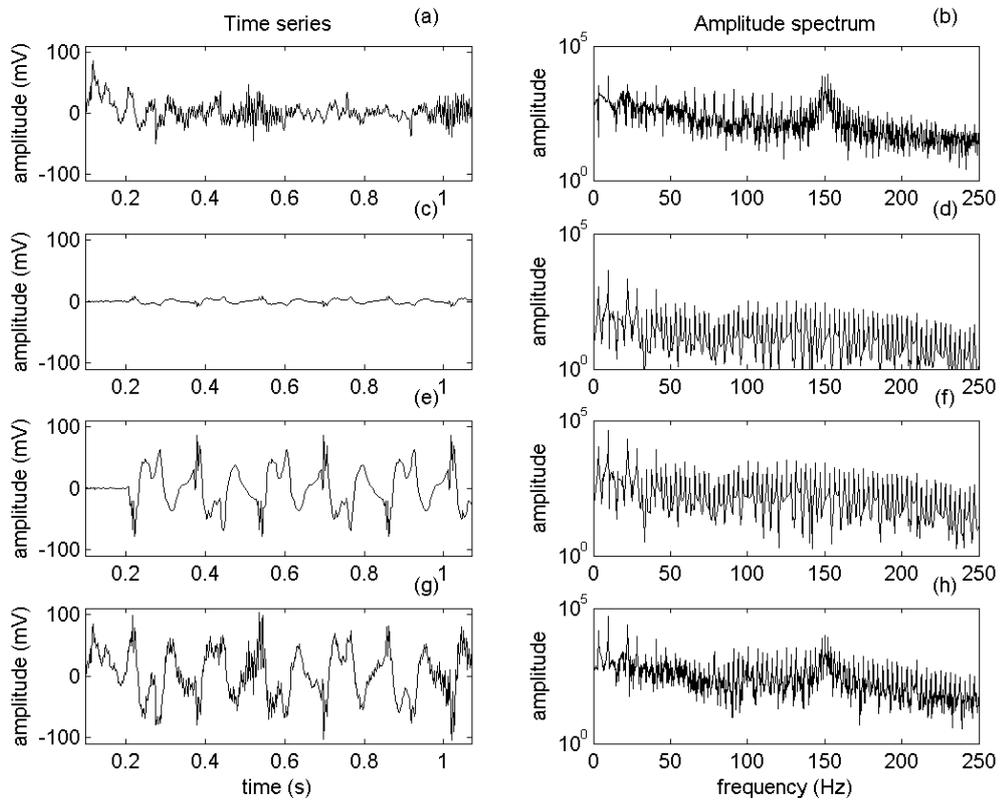


Figure 3.7. Robust statistics for removing low frequency noise.

A stack of the transmitted signal is removed leaving a noise estimate (a) amplitude spectrum (b). The noise estimate for each sensor is stacked producing (c). This noise estimate is then weighted according to the RMS noise of individual sensor (e), this weighted stack of noise is removed from (a) producing (g).

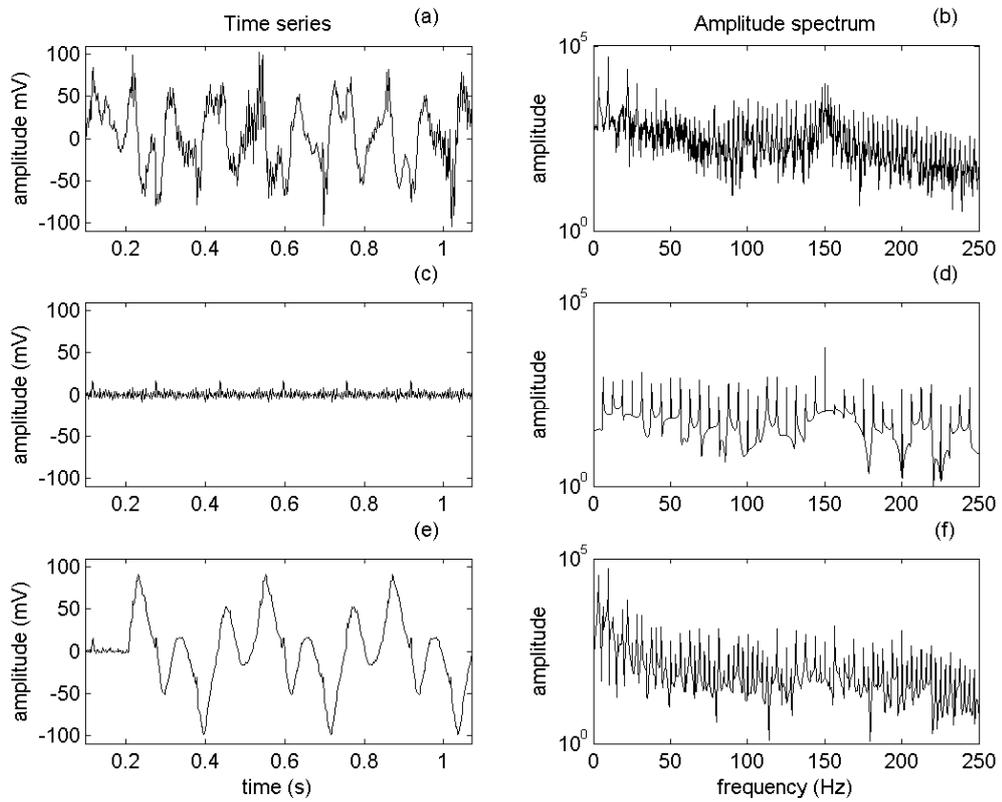


Figure 3.8. Robust statistics as a means of removing high frequency noise. After low frequency noise removal, the noise estimate (a) remains. This remaining noise train is stacked with alpha trimmed median binning (which excludes the upper and lower 10% median values), producing (c). The signal estimate prior to low and high frequency noise attenuation is added to (c) producing (e).

4. Powerline Harmonics

Stationary sinusoidal noise, such as that generated from power distribution grids, contaminates electromagnetic, induced polarisation, resistivity, magnetotelluric, seismoelectric and seismic data. Other sources of this distinct frequency noise are generators at mine sites or any rotating machinery. The base frequency of a the transmitted electrical waveform is

$$bf = \frac{hf}{n}, \quad (4.1)$$

where n is an even whole number and hf represents the local powerline transmission frequency. This permits optimal powerline harmonic removal as synchronous averaging in the time domain has a similar effect in reducing coherent noise as narrow band filtering in the frequency domain. In Australia, the power transmission line is emitted at 50 Hz, though in North America it is 60 Hz. The worst base frequencies to use are those that satisfy $bf = hf / n$, where n is an odd whole number (Macnae et al., 1984; Spies and Frischknecht, 1991; Nabighian and Macnae, 1991 and Duncan, 2002). Employing a sensible base frequency is common practice for any electrical surveys, but it does not completely remove harmonic noise and many other methods are in place to attenuate the noise.

Traditionally notch filtering is used to remove these discrete frequency noise sources. Notch filters deal with one frequency at a time and attenuates both the signal and noise at the notch frequency. Notch filters and circuits to reject phase coherent cultural signals are customarily included in a variety of modern exploration systems.

4.1. Inversion

Researchers have dealt with inversion techniques to solve for the phase and amplitude of stationary sinusoidal noise after initially determining the sinusoidal frequency. This is done by performing a search over a portion of the signal where the noise is strongest or there is no transmitted signal. Nyman and Gaiser (1983), Griffith (1988) and Dragoset (1995) use adaptive noise cancellation using least-mean squares. Linville and Meek (1992 and 1995) and Butler and Russell (1993, 2003) and Butler *et al.* (1996) have developed linear least-squares inversions. Xia and Miller (2000) use the Levenberg-Marquardt method with a singular value decomposition (SVD) approach.

Powerline harmonics deviate from their fundamental frequency up to **Error! Not a valid embedded object.** 0.03 Hz in developed countries, and up to **Error! Not a valid embedded object.** 5 Hz in less developed countries. This is known as amplitude modulation. The record length that traditional inversion techniques can operate on is restricted. If the record length is too long residual harmonics will be left in the time series. This is characterised by an increase in residual harmonics toward the end of the trace. The same result will be observed if an incorrect fundamental frequency is selected for signal processing. In the case of modulated harmonics, Butler and Russell (2003) suggest estimating and including the appropriate side-bands as additional fundamentals. Difficulties will be encountered in this method, for when modulation is a factor, the amplitude of the side-bands of the fundamental frequencies are often hidden within the lobes of the fundamental frequency peak and therefore not possible to estimate (Figure 2.8).

4.1.1. Non-Linear SVD Inversion for Modulated Harmonics

In this study the approach of non-linear SVD inversion is used for the removal of sinusoidal noise. Sinusoidal noise with amplitude modulation is described in section 2.4.1. where Equation (2.4) describes the behaviour of modulated powerline harmonic. There is a need to invert for the amplitude and frequency of each harmonic and the amplitude and frequency of the modulation envelope. However Equation (2.4) cannot be used as the forward model, because it is highly non-linear and the inversion will not converge. Instead, when modulation is dominant, the time series must be divided into multiple sections. The harmonic noise with amplitude modulation maybe represented as the summation of sinusoids with a quadratic function defining the modulation envelope as:

$$S(\vec{m}, t) = \sum_{i=1}^n a_i (1 + c_i t + d_i t^2) \sin(2\pi f_i t) + b_i (1 + c_i t + d_i t^2) \cos(2\pi f_i t), \quad (4.2)$$

Where f_i is the frequency of the sinusoid, t is time, a_i and b_i are the amplitudes of the sine and cosine components (the ratio of these is directly related to the phase of the sinusoid), c_i and d_i are the nominators of the quadratic amplitude modulation envelope and n is the number of sinusoid functions needed to model the sinusoidal noise.

A SVD damped Gauss-Newton method is used to solve for the model parameters m as

$$\vec{m} = [a_1 \quad b_1 \quad f_1 \quad c_1 \quad d_1 \quad a_2 \quad b_2 \quad \dots \quad a_n \quad b_n \quad f_n \quad c_n \quad d_n], \quad (4.3)$$

The forward model for the non-linear inversion is trivial. Therefore the Jacobian (A) is used for inversion and is defined by the actual partial derivatives of $S(m, t)$ with respect to the model parameters.

$$A = \begin{bmatrix} \partial S / \partial a_i & \partial S / \partial b_i & \partial S / \partial f_i & \partial S / \partial c_i & \partial S / \partial d_i \end{bmatrix}, \quad (4.4)$$

where the size of A is equal to the length of the time series (or sections of) by $(n \times m)$, and where;

$$\frac{\partial S}{\partial a_i} = (1 + c_i t + d_i t^2) \sin(2\pi f_i t), \quad (4.5)$$

$$\frac{\partial S}{\partial b_i} = (1 + c_i t + d_i t^2) \cos(2\pi f_i t), \quad (4.6)$$

$$\frac{\partial S}{\partial f_i} = 2a_i (1 + c_i t + d_i t^2) \cos(2\pi f_i t) \pi t - 2b_i (1 + c_i t + d_i t^2) \sin(2\pi f_i t) \pi t, \quad (4.7)$$

$$\frac{\partial S}{\partial c_i} = a_i t \sin(2\pi f_i t) + b_i t \cos(2\pi f_i t), \text{ and} \quad (4.8)$$

$$\frac{\partial S}{\partial d_i} = a_i t^2 \sin(2\pi f_i t) + b_i t^2 \cos(2\pi f_i t). \quad (4.9)$$

The initial estimate of f_i is a two stage process. A window of data from the time series is selected. For most desirable results the windowed data contains only the sinusoidal noise and the received signal, (i.e. noise from other sources is minimal). The transmitter signal is calculated. In the amplitude spectrum, peaks which correlate with the transmitter signal are set to nominal values. Window searches are conducted over the spectrum and the indices of peaks which are greater than 2 times the mean of their window are selected as the sinusoidal frequencies. The maximum peak and its neighbouring points are set to the mean, and the window is searched again. This ensures that all sinusoidal frequencies are accounted for (Figure 4.1).

These frequencies are then refined. The signal estimate is stripped from the time series and a finer pick for the frequencies is made using the whole trace. In the amplitude spectrum

each initial frequency estimate is windowed, and the index of the maximum amplitude within each window is selected as the frequency.

Utilizing the whole trace locates the frequency of the harmonic to $\pm 1/(\text{total time of the time series})$ (Figure 4.2). The initial approximations for a_i and b_i are completed by taking the cross correlation of the time series with the sine and cosine components for each f_i . c_i and d_i are initially set to zero for the idealistic non-modulation case.

The inversion for the model parameters is a two part process for each trace. The initial inversion is carried out, then a weighted inversion, with weighting for each increment in the time series, equal to $1/(\text{residual noise remaining})^2$. This weighting reduces the influence of the spherics on the inversion. The final estimates of the sinusoidal noise are subtracted from the time series (Figure 4.3).

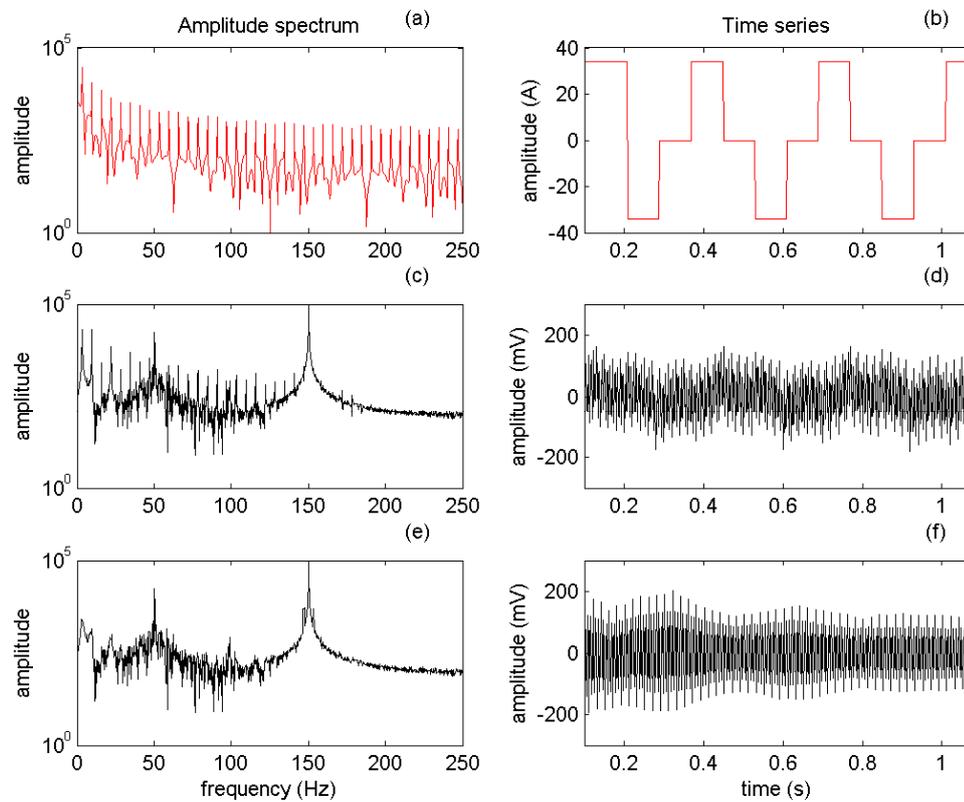


Figure 4.1. Harmonic noise removal the initial estimate the harmonic frequency, stage one.

The transmitter response (b) and its amplitude spectrum (a) are used to estimate the received signal (d). Peaks in the amplitude spectrum of the transmitted signal will correlated with peaks in the amplitude spectrum of the received signal (c). These peaks can be trimmed to an average background value, producing (e), an estimate of the noise spectrum and the corresponding time series (f). This noise spectrum (e) can be used to estimate harmonic frequencies. The noise spectrum is broken up into small frequency ranges or windows, each window is analysed and frequencies where the peak amplitude is greater than 2 times the mean of the amplitudes in the window are selected as the sinusoidal frequencies.

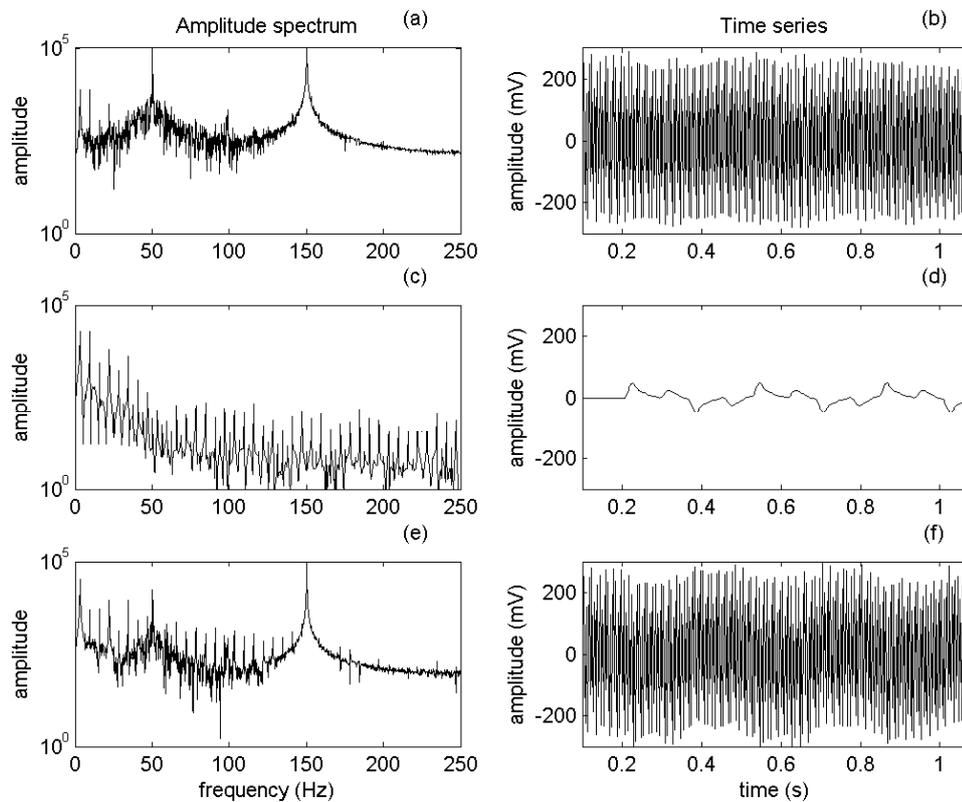


Figure 4.2. Harmonic noise removal the initial estimate the harmonic frequency, stage two.

Following the initial estimate of the harmonic frequencies (as seen in Figure 4.1). The frequency estimates are refined using all data from the time series for each individual sensor. An estimate of the received signal (d) and it's amplitude spectrum (c) is obtained by stacking the signal received at all the sensors (b), with corresponding amplitude spectrum (a). The estimate of the received signal (d) and (c) are subtracted from (b) and (a) producing (f) and (e) an estimate of the remaining noise and it's spectrum for the sensor. The frequencies which were selected in stage one (Figure 4.1) are windowed in the noise spectrum (e), the index of the maximum amplitude within the window is selected as the frequency. Utilizing the whole trace locates the frequency of the harmonic to $\pm 1/(\text{total time of the time series})$.

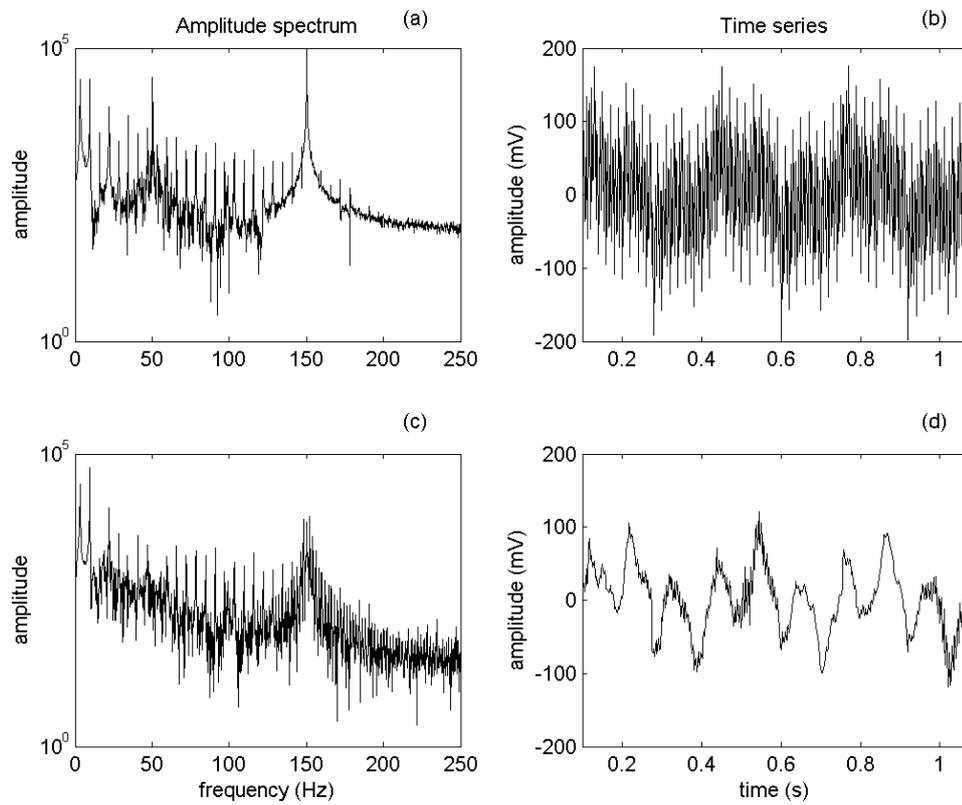


Figure 4.3. Comparison of the received signal before and after powerline harmonic attenuation.
The received time series (b) and its amplitude spectrum (a) in comparison to the received signal after harmonic removal (c) and its amplitude spectrum (d). Peaks in the amplitude spectrum (d) at 50 and 150 Hz, produced by powerline harmonics have been attenuated.

5. Array Processing

Data collected with a multi-sensor array has had little application in mineral exploration. In 1994, M.I.M Exploration Pty Ltd (MIMEX) designed and built a multichannel acquisition system named MIMDAS (Mount Isa Mines' distributed acquisition system). The foremost approach using arrays for electrical methods in mineral exploration prior to MIMDAS was the remote reference technique, where a second receiver is placed at a distance from the target signal receiver and measurements are made simultaneously at both receivers. The remote receiver can be used to estimate local noise sources, which can then be subtracted from the target signal receiver. The remote reference must be placed far enough from the transmitter (>10 km) that the observed remote fields consist only of geomagnetic fluctuations. The remote receiver may be the same instrumentation as the target receiver. Alternatively it can be a Superconducting Quantum Interference Device (SQUID) magnetometer. The technique extends low frequency (below 0.3 Hz base frequency) limits by one decade.

When using a remote reference, there is significant distance between the target receiver and the remote reference, therefore least-squares fit to find the transfer tensor between the target receiver field components and the horizontal field components at the remote receiver is performed before the noise can be subtracted. The tensor accounts for the lateral conductivity change and relative alignment errors between the base and remote reference. However, because the exact sensor relationship is frequency dependent and the power spectra of geomagnetic noise change over a finite time intervals, there is no guarantee that the coefficients calculated during an interval with the transmitter off will apply to data collected during the transmitter on time.

Data is required to be gathered in short segments so that natural field drift can be offset for each segment. The main limitation of this remote reference technique arises from the quantisation errors of the receiver instrumentation. S/N improvement of roughly 20 dB is expected (Wilt, *et al.*, 1983).

Another possible scheme uses a symmetric receiver configuration in which the noise is subtracted and the signal is added. This scheme is sensitive to alignment errors, as any minimum coupled source receiver measurement, and could only be used in layered Earth environments in which the signals would be mirror images at the two sites (e.g. San Filippo and Hohmann, 1983).

Spies (1988), Buselli *et al.* (1995 and 1997) researched spherics elimination techniques using a local noise prediction filter (LNPF) and a remote noise prediction filter (RNPF). A LNPF predicts the vertical component of the spherics noise from the two horizontal components. For a RNPF, noise is predicted for each component of spherics noise from simultaneous three component measurements of spherics at a separate station. They found a noise prediction filter can decrease the largest (horizontal) component of spherics noise by a factor of 18, with the average noise reduction factor between 2 and 6.

Spies (1991) patented a cancelling antenna, which is wrapped around the sensing antenna of the electromagnetic receiver equipment. The cancelling antenna is provided with an alternating current signal of the same frequency as the ambient powerline noise. The cancelling antenna produces an electromagnetic field that is 180 degrees out of phase and of equal amplitude to the ambient powerline noise, as measured by the sensing antenna. The alternating current is produced by a noise antenna, a phase-locked loop and an amplifier. The noise antenna receives the ambient powerline noise, the phase-locked loop locks onto and tracks the frequency of the noise, and the amplifier provides the necessary amplification (Spies, 1991).

Stephan and Strack (1991) developed a method of noise reduction known as the local noise compensation (LNC) technique. It is a noise cancellation technique for a multichannel far-zone time-domain EM acquisition with dense station spacing. An estimate of noise at the base station is calculated by removing a stacked estimate of the transient from each base station period. This noise estimate is then removed from all receiver stations. To apply the LNC, the following conditions must be met and tested at the beginning of the survey. The

noise must be correlated over a certain range around the base station, and this regional noise cannot be reduced by standard processing. The receiver systems at the base station and at the surrounding mobile stations must be identical in their system characteristics, and the recording times in the data header must be synchronised to identify corresponding records. The very localised noise belonging to only one receiver site (i.e. wind noise or noise from nearby powerlines) must be smaller than the regional noise. In situations where these conditions are met the, length of useable signal was increased by a factor of 10 (Stephan and Strack 1991).

Other disciplines, such as exploration seismology, passive sonar, radar, radio astronomy and tomographic imaging, have looked at array signal processing techniques (Haykin *Ed.*, 1985, and Robinson, 1972). To date array processing has had limited application in mineral geophysics. A comprehensive sequence of processing steps are shown here, which apply both a non-linear SVD inversion for the removal of powerline harmonics, and iterative stacking and robust statistics for the removal of array correlate noise and intrinsic sensor noise.

The iterative stacking and robust statistics exploit the use of an array of sensors by using the whole array to estimate the noise across the array, as data collected with a sensor array consists of three main components:

- 1.Signal correlated with both the transmitter output and sensor array: *the Earth response, including the signal of interest and other geological influences.*
- 2.Noise that is correlated within the sensor array, but uncorrelated with the transmitter: *harmonic noise (from the power grid) and atmospheric transients (spherics).*
- 3.Noise that is both uncorrelated within the sensor array and to the transmitter output: *the random intrinsic noise of each sensor.*

Using a sensor array, it is possible to recognise the different components received and remove noise uncorrelated with the transmitter. There is increased coverage and resolution. All data recorded is full waveform (both transmitter on and off time are recorded), this ensures that no information is lost for any time. The use of multiple receiving sensors can improve signal fidelity from each sensor, leading to more accurate geological models and

greater depth of investigation. Essentially all of the receivers behave as each other's remote references, though rather than getting a noise estimate from just one receiver, all of the receivers in the array can be used to get an estimate of the correlated noise across the array or an estimate of the transmitter correlated signal.

In this study the approach is to firstly remove the harmonic noise from each individual sensor (Section 4.1.1) and then the remaining sensor array correlated noise is removed using an iterative robust statistics stacking technique. The stacking consists of two steps. The first attenuates the low frequency noise component and the second the high frequency noise component (Section 3.3). Figure 5.1 shows the processing flow. This approach effectively attenuates modulated harmonics, spherics and the random intrinsic noise of each sensor. A significant advantage of this approach to compared to the remote reference method is that it does not add the intrinsic noise of the remote reference to the data and a sensor is not wasted.

Figures 5.2 to 5.4 show the application of the array processing workflow to three different data sets. The IP and EM surveys recorded at Curtin University, and MIMDAS IP survey. Details of these surveys can be found in Appendix 1.

The IP survey recorded at Curtin University (Figure 5.2) is strongly contaminated with powerline noise. This is indicated in by the relatively wide peaks at 50 Hz and 150 Hz which dominate the raw amplitude spectrum (Figure 5.2b). The inconsistent amplitude envelope of the time series indicates that the powerline noise is modulated (non-linear) (Figure 5.2a). After application of the non-linear SVD inversion for the attenuation of powerline harmonics the peaks at 50 Hz and 150 Hz are removed from the amplitude spectrum (Figure 5.2(d)). Following the array processing the only visible peaks in the amplitude spectrum are due to the transmitted signal (Figure 5.2(f)). The signal to noise improvement is a factor of 200.

Figure 5.3 shows the EM survey recorded at Curtin University. The time series show the time decay for one recorded event (Figures 5.3(a), (c) and (e)). There is minor

contamination by harmonics, seen as small peaks in the amplitude spectrum in the locality of 50Hz (Figure 5.3(b)). These are completely removed after application of the non-linear SVD inversion is for the removal of sinusoidal noise (Figure 5.3(d)). The array processing contributes the most to signal improvement with a dramatic improvement in decay curve fidelity (Figure 5.3(e)).

Figure 5.4, the MIMDAS IP exploration example is contaminated with harmonics and some other noise sources. Harmonic contamination is indicated by the presence of a peak at 50Hz in the amplitude spectrum (Figure 5.4(b)). After array processing the final signal to noise improvement is a factor of 30.

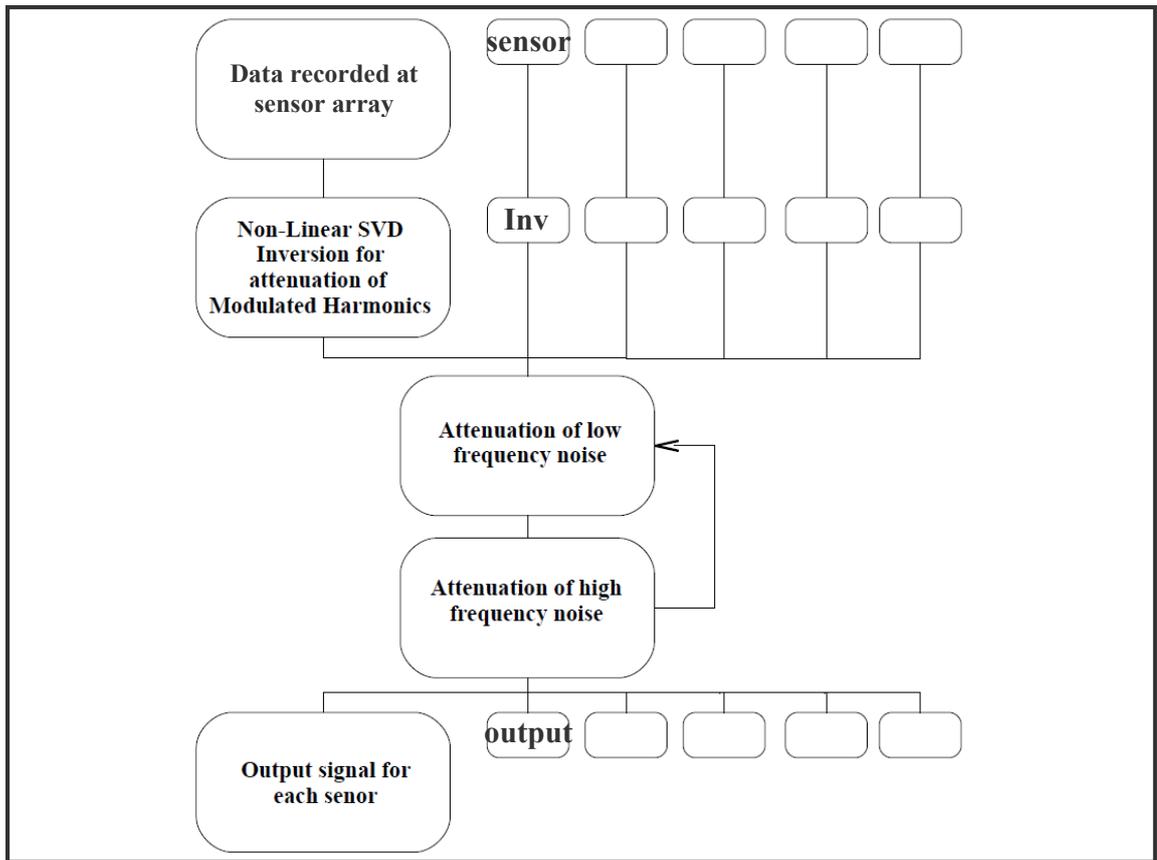


Figure 5.1. Array processing flow chart

The harmonic noise from each individual sensor is attenuated using a non-linear SVD inversion for modulated harmonics (Section 4.1.1) and the remaining sensor array correlated noise is removed using an iterative robust statistics stacking technique. The stacking consists of two steps. The first attenuates the low frequency noise component and the second the high frequency noise component (Section 3.3).

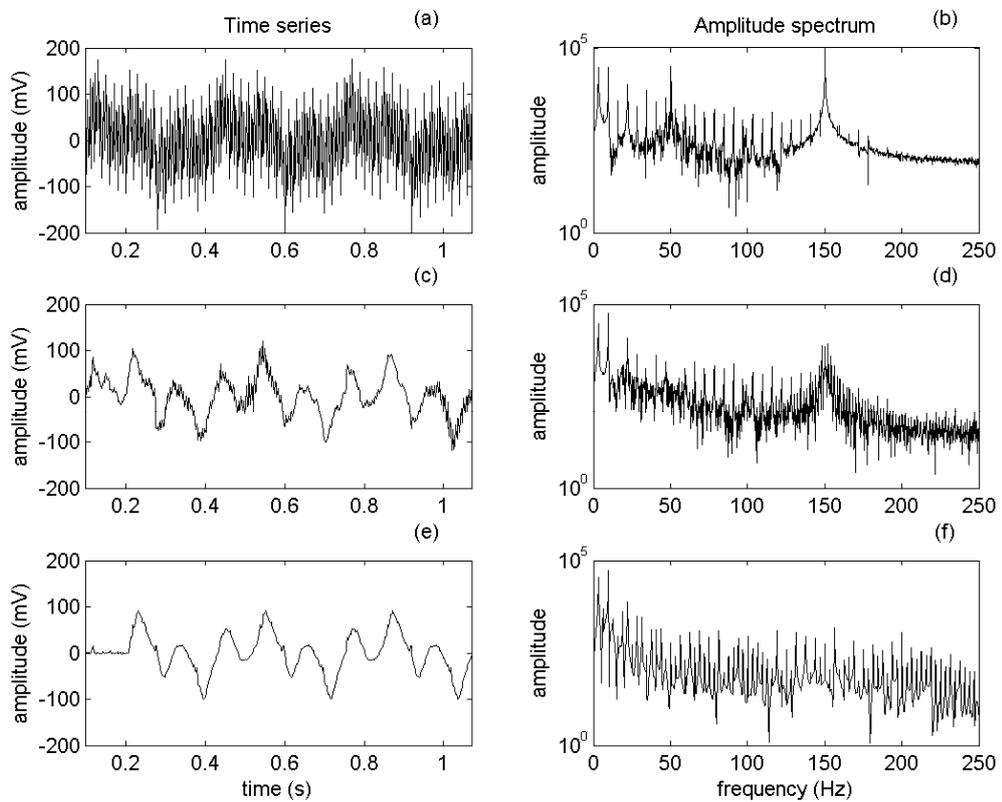


Figure 5.2. Array processing a field example: Curtin electrical data array processing. The initial time series (a) is contaminated with modulated harmonic noise, with the majority of the contamination from 150 and 50 Hz harmonics shown as spectral lines in (b). After SVD inversions to remove the harmonic frequencies the spectral peaks in the amplitude spectrum are greatly reduced (d) and the time series is more refined (c). Following harmonic removal the iterative robust statistics and alpha trimmed median stacking is applied producing (e). The amplitude spectrum is now predominately the transmitter correlated signal and there is a signal to noise improvement of 200. Details of the survey can be found in Appendix 1.

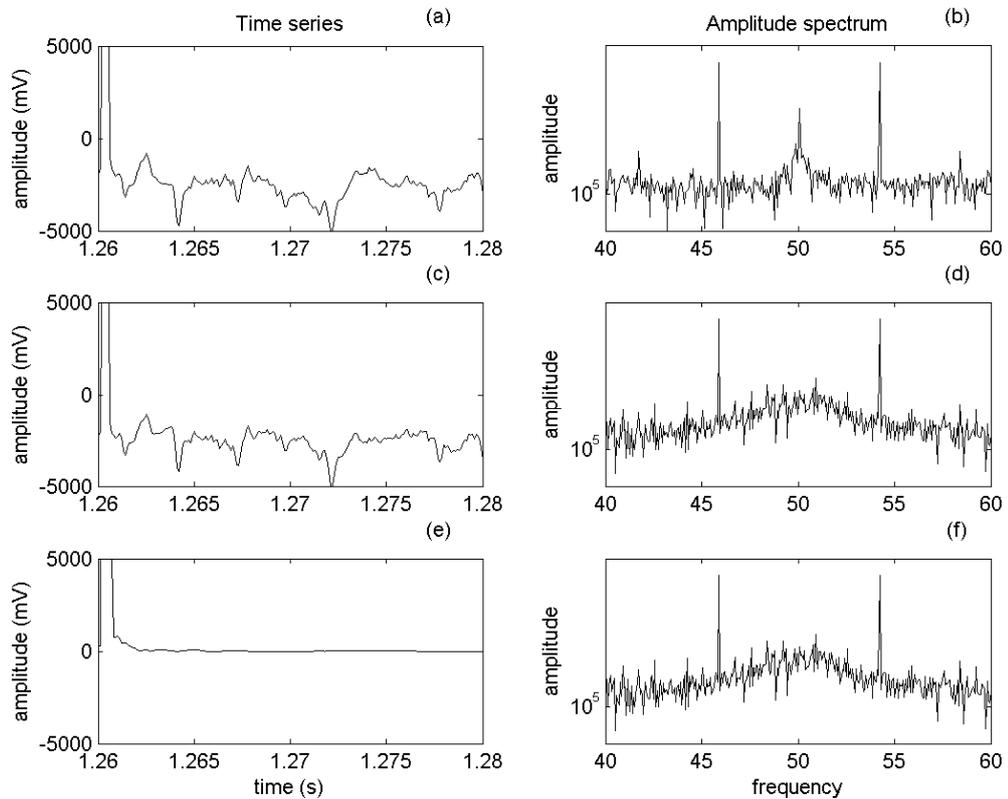


Figure 5.3. Array processing a field example: Curtin electromagnetic data array processing.

Part of the initial time series (a) is shown, there is some subtle harmonic noise, indicated by a small peak at 50 Hz in the amplitude spectrum (b). After SVD inversions the 50 Hz peak is removed, as shown in amplitude spectrum (d), there is only slight change to times series (c). Following harmonic removal, iterative robust statistics and alpha trimmed median stacking is applied producing time series (e). Details of the survey can be found in Appendix 1.

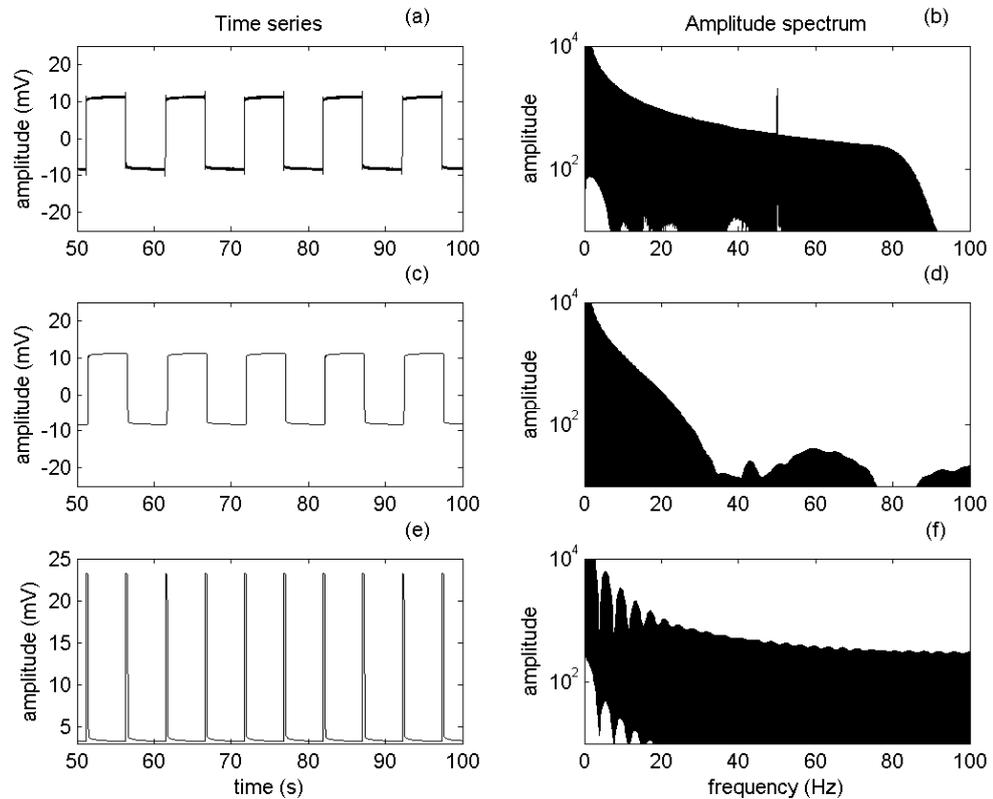


Figure 5.4. Array processing a field example: MIMDAS data array processing, data courtesy of T. Ritchie, Xstrata Pty Ltd.
The initial time series (a) is contaminated with modulated harmonic noise seen as a narrow spectral line at 50 Hz in amplitude spectrum (b). After SVD the 50 Hz spectral line is removed, as shown in amplitude spectrum (d) and the time series is uncontaminated (c). Following harmonic removal the time series is rectified then iterative robust statistics and alpha trimmed median stacking is applied producing the secondary field (e). There is a signal to noise improvement of 30. Details of the survey can be found in Appendix 1.

6. Conclusion and Recommendations

In this thesis I have analysed noise removal methods for electrical and electromagnetic time series measurements. The study found that array processing is favourable to the remote reference method, as each sensor in the array can be used to calculate the noise source rather than just one or two sensors as in the remote reference method. It also showed a multi-faceted approach to processing is beneficial. Chapter Three reviewed current stacking techniques and demonstrated a processing flow using iterative stacking in conjunction with robust statistics for S/N ratio improvement. Chapter Four reviewed inversion techniques for powerline harmonic removal and demonstrated a superior technique of SVD inversion which can remove modulated harmonics. Chapter Five combined an array processing flow which incorporated SVD inversion for the removal of modulated harmonics in conjunction with iterative stacking and robust statistics. The processing flow was applied to field examples of electrical and electromagnetic data, collected in both cultural and exploration environments. In areas where data was highly contaminated with cultural noise signal to noise ratio improvements of 200 were observed. These multifaceted array processing techniques lead to much improved signal to noise (greater than 46 dB), which ultimately leads to greater signal fidelity and greater depths of investigation.

Chapter Three reviewed pre-whitening/pre-emphasis and showed a method of spiking deconvolution for spherics removal. Spiking deconvolution as a method of spheric removal proved flawed. One of the caveats for spiking deconvolution is that the inverse of the deconvolution operator, the spheric event, should be minimum phase. A local spheric event will be minimum phase, but as the distance from the source increases and/or more than one spheric event arrive in the same time window, the minimum phase assumption fails and so does the method.

This research is a foreword to the yet greatly unexplored area of array processing for electrical and electromagnetic time series. Great gains can be made by intelligent use of

any array of sensors and many processing applications are still to be explored. There is advantage to reviewing existing processing methods in the seismic data processing industry, which has an established history for array processing.

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Appendix 1. Data Acquisition

All data recorded is full waveform both transmitter on and off time are recorded. The data was with multiple receivers and recorded simultaneously in time at each sensor. The premise is data collected with a sensor array consists of three main components:

- 1.Signal correlated with both the transmitter output and sensor array: *the earth response, including the signal of interest and other geological influences.*
- 2.Noise that is correlated within the sensor array, but uncorrelated with the transmitter: *harmonic noise (from the power grid) and atmospheric transients (spherics).*
- 3.Noise that is both uncorrelated within the sensor array and to the transmitter output: *the random intrinsic noise of each sensor.*

Using a mutiple sensor array, it is possible to recognise the different components received and remove noise uncorrelated with the transmitter. Three multiple sensor field surveys are presented: *MIMDAS IP*, IP Survey ar *Curtin University* and *EM survey at Curtin University*. The *MIMDAS IP* is an exploration dataset and the Curtin datasets were recorded in an urban environment. The *MIMDAS IP* data was provided by Terry Ritchie, MIM Exploration Pty Ltd (now Xstrata Plc). Streaming noise data, *Noise Darwin*, was provided by Jock Buselli, CSIRO Exploration and Mining, originally recorded on magnetic tape with an analogue tape recorder.

1A. MIMDAS IP

An exploration example: IP data were collected with MIM Exploration's (now Xstrata) proprietary distributed acquisition system "MIMDAS". MIMDAS, is a broadband, high resolution, distributed (multi-channel) acquisition system. MIMDAS is designed to acquire networked multi-channel electrical and EM geophysical data, though it could be used to acquire seismic data. Each channel or data acquisition unit (DAU) is networked in series with a computer interfaced central recording unit (CRU) and is capable of capturing

250,000 samples or 1 Mb of data at many hardware specific, but software selectable, sample rates. MIMDAS has four gain settings and is capable of 24-bit resolution at sample rates up to 4800 Hz and 19-bit resolution from 4800 Hz to 48 kHz inclusive (Garner and Thiel, 2000). For the *IP-MIMDAS* data the transmitter used a 100 % duty cycle waveform with peak-to-peak amplitude 13.7 A, base frequency of 0.0977 Hz, sampling frequency 200 Hz and a total of 32 stacks for each recording. There were 84 dipoles 100 m apart with line spacing of 200 m covering a survey area 1.4 km by 1 km.

1B. IP Survey at Curtin University

Urban Example: A simulated low power electrical/IP survey was conducted in Perth, Western Australia. A transmitter with a 50 % duty cycle waveform with, base frequency of 3.125 Hz, injected current with a peak-to-peak amplitude of 34 mA into the ground with a 5 m grounded dipole. There were a total of 30 stacks. The receiving system consisted of 24 grounded dipoles (spacing 1 m), located 10 m from the transmitter, connected to AC coupled pre-amplifiers (each with a gain of 30) relayed to an OYO DAS-1 seismic acquisition system with a sampling frequency of 500 Hz. The AC coupled amplifiers in the OYO DAS-1 and the preamplifiers remove low frequencies in the signal, thus, simulating a signal that might be received by an IP/EM system.

1C. EM Survey at Curtin University

Urban Example: A NanoTEM NT-20/2T transmitter with a 50 % duty cycle waveform with base frequency of 4.166 Hz, total of 130 stacks, injected current with a peak-to-peak amplitude of 3.2 A via a 100 m by 100 m transmitter loop. The receiving system consisted of 8 scalar receivers in-loop, with areas 10000 m², the receivers were developed in house by Anton Kepic and are known as Curtin University of Technology Coils (CUT). The receivers were connected to AC coupled pre-amplifiers (each with a gain of 30) relayed to SMARTem acquisition system with a sampling frequency of 10000 Hz.

1D. Noise Recording in Darwin

Noise Example: Three data sets were provided by Jock Buselli of CSIRO. Data were recorded on magnetic tape with an analogue tape recorder. Data were then digitised and recorded on a cassette tape, and later transferred to CD. Data were recorded in Darwin, February 1983. Sampling frequency was 20000 Hz, the receivers consisted of two 100 m by 100 m surface loops with gain 10 (20 dB) and gain 100 (40 dB), and a z-component scalar receiver with area 10000 m², with gain 100 (40 dB).