Calibration Tests For Underwater Acoustic Transponders Using Kalman Filter

Mohd Razali bin Mahmud Centre for Hydrographic Studies Fakulti Ukur dan Harta Tanah Universiti Teknologi Malaysia

Abstract

Two tests were carried out for the calibration of underwater acoustic transponders using Kalman filter. Both tests used the Long Baseline Acoustic Positioning method. The first test is known as the relative calibration test and uses observables only from underwater acoustic transponders. This test is for the adjustment of the shape and scale of the transponder array. It requires at least three coordinates to be fixed. The second test, known as the absolute calibration test, involves the observables from four systems (i.e. DGPS, underwater acoustic system, range-range system and hyperbolic system) integrated together to give geodetic position and orientation to the transponder array. The results of both tests are presented.

1.0 INTRODUCTION

Long Baseline Acoustic Positioning System provides local control and high position repeatability independent of water depth for the surface or sub-surface vessel, particularly when the site is out of the range of a suitable surface positioning systems. In this technique, the position of the survey vessel is related to the transponder array on the seabed. Hence, the only practical method of accurately determining the surface and sub-surface positions beyond the line of sight is with referenced to an array of bottom moored acoustic transponder. During operation each transponder in interrogated from the navigating vehicle. The resulting range measurement, to each of the transponders are processed to determine the position of the vehicle relative to the transponder array. However, navigation of surface and sub-surface vessels in a transponder array depends on a prior precise determination of the transponder coordinates (i.e. calibration).

There are several methods commonly employed. Early methods were based on clover-leaf and baseline crossings techniques. These methods require large amounts of ship time and critical ship path. Later methods have been developed based on range data from a number of random ship positions which acoustic markers are then positions from an interative least squares fitting procedure. The method employed in this paper is based on Kalman filter.

2.0 KALMAN FILTERING

The theory of mathematical filters has mainly been developed by statisticians and electrical engineers. The description of the filter type most commonly used presently for scientific and engineering applications is by Kalman (1960). The filter which is now call the Kalman filter can be described as a computation techniques that enables real-time estimation of quantities of interest such as satellite's or ship's position. The filter may be applied to any situation where the desired parameters vary with

time. It is used to estimate the various states of a random process from a set of discrete measurements having a known linear connection to these states.

Kalman filter provides a set of algorithms for the estimation of the state vector (i.e. unknown parameters) at any point in time. Its filtering process follows a recursive sequence of mathematical models which remember past data, receive present data and calculate the best estimates of present and probable future positions, based upon the combination of past and present information. It produces optimal estimates of the state vector with well defined statistical properties. The estimates of the state vector are unbiased and have minimum variance, so long as observations and model are normally distributed.

2.1 State Vector

The state vector is a vector of desired parameters. It must include not only those parameters which we wish to estimate but also other parameters necessary to model the dynamic behaviour of the vessel. The elements of this vector might be position, velocity and acceleration.

2.2 Mathematical Models

Filtering makes use of information available from two sources i.e observations and some prediction on how the vessel is expected to move.

This leads to the use of two functional models:

- i) Measurement model
- ii) Dynamic model

Measurement Model

Let us assume that at time i we make some measurements l_i which are related to the state vector by the functional relationship:

$$F_i(\overline{x_i}) = \overline{l_i}$$

...(1)

If the relationships between measurements and state vectors are non-linear, then the relationships have to be linearised for use in the Kalman filter. The linearised functional model for observations at time i is given by:

$$A_i x_i = b_i + v_i$$

. . . (2)

where A_i ... is the Jacobian matrix or simply $\delta F_i / \delta x$

b, ... is the 'observed minus computed' quantities

v, ... is the vector of residuals

Dynamic Model

This model is based on some knowledge on how the state vector is expected to vary with time. In its most general form, the functional relationship of the dynamic model may be represented as:

$$F_{i-1,i}(\bar{x}_{i-1},\bar{x}_{i},t_{i-1},t_{i})=0$$

. . . (3)

where

 \overline{x}_{i-1} ... is the true state vector at time \mathbf{t}_{i-1}

 \overline{x}_i ... is the true state vector at time \mathbf{t}_i

The linearised dynamic model relating the state vector at times i-1 and and i is given by :

$$X_{i} = M_{i-1,i} x_{i-1} + y$$
 ... (4)

where

M _{i-t,i} ... is the transition matrix or dynamic matrix describing,

approximately, how the state vector changes from

time i-1 to i

y ... is the vector of unknown true errors in the model

and are assumed to be randomly distributed about a

zero mean

For most practical problems, it is convenient to consider the noise vector y as being given by:

$$y = Tg ... (5)$$

where

g ... is the vector of the quantities which cause the model to be in error with covariance C_g , i.e. $g \sim (0, C_g)$

T ... is the coefficient matrix which describes how

is the coefficient matrix which describes how g propagates into the state vector.

Thus, equation (4) becomes:

$$x_i = M_{i-1,i} x_{i-1} + T_g$$

. . . (6)

The vector g is not actually known but its covariance, C_g can be estimated. As a result, the covariance matrix of g can be computed using Gauss's propagation of error law:

$$C_y = T C_g T^T$$

. . . (7)

2.3 Kalman Filter Algorithms

The derivation of the Kalman filter from standard least squares requirement is found in Cross (1987). The Kalman filter consists of the following parts:

- i) Time update equations (Prediction)
- ii) Measurement update equations (Filtering)
- iii) Smoothing equations (Smoothing)

Time Update Process

The time update is the prediction of the state vector and its covariance (error). The prediction equation of the state vector can be derived directly from the dynamic model of (4). However, since the vector y is not actually known, we make an assumption that it is zero. Thus, the prediction equation of the state vector is given by:

$$\overline{x}_{i}(-) = M_{i-1,i}\overline{x}_{i-1}(+)$$

...(8)

where the symbols - (i.e. bar) denotes an estimated quantity, and the symbols (-) and (+) following a vector denote the value of that vector at the instant in time before and after a measurement update.

The transition matrix, $M_{i-1, i}$ allows calculation of the state vector at some time i, given complete knowledge of the state vector at i-1, in the absence of the dynamic model noise, y. Thus, equation (8) gives the state vector at time i predicted using information up to time i-1.

The predicted covariance matrix of the state vector can be obtained from:

$$C_{\bar{x}_i}(-) = M_{i-1,i}C_{\bar{x}_{i-1}}(+)M_{i-1,i}^T + C_y$$

. . . (9)

where C_y ... is the covariance matrix of dynamic model errors.

This prediction might be made to some time at which the state is required to be known but at which time there are no measurements. In this case, the same equations (i.e (8) and (9)) may be used again ad again to compute the predicted state at any number of epochs until another set of measurements is available.

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Measurement Update Process

The measurement update is the improvement of the prediction (both the state vector and its covariance) which gives the filtered state estimates. The filtered algorithms are given by the following equations:

$$\overline{x}_i(+) = \overline{x}_i(-) + G_i(b_i - A_i\overline{x}_i(-))$$

. . . (10)

$$C_{\vec{x}_i}(+) = (I - G_i A_i) C_{\vec{x}_i}(-)$$

. . . (11)

where

$$G_i = C_{\bar{x}_i}(-)A_i^T (A_i C_{\bar{x}_i}(-)A_i^T + W_i^{-1})^{-1}$$

. . . (12)

G is the so-called Kalman gain matrix. The gain matrix performs the role of combining the dynamic model and the observations. It controls the amount by which a particular set of observations affects the predicted state vector.

The updated state vector is obtained after the difference between the actual and predicted measurements has been computed, i.e $b_i - A_i \overline{x_i}(-)$. Equation (10) gives the best estimator of the state at time i using both $\overline{x_i}(-)$ and b_i .

Smoothing Process

Smoothing process is carried out after the measurements have been completed. It is a requirement that the predicted and updated state vectors and their corresponding covariance have been stored.

2.4 Non-Linear Measurement Model

The measurement update equation listed in (10) is used to solved problems with linear observations equations. However, for offshore positioning most measurement lead to non-linear equations.

As mentioned in the 'measurement model' section if the problems involved non-linear functional model, they have to be linearised as in (2) for used in the Kalman filter.

The design matrix, A_i and the gain matrix, G are computed based on the predicted state vector, $\overline{x}_i(-)$. Thus the filtered estimate of the state vector is given by:

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$$\overline{x}_{i}(+) = \overline{x}_{i}(-) + G_{i}(l_{i} - comp(\overline{x}_{i}(-)))$$

$$\dots (13)$$

where

 $comp(\overline{x}_i(-))$ is the vector of observation computed using the predicted state vector at time i is the vector of observation at time i.

2.5 Summary of the Kalman Filter AlgorithmsA summary of the algorithms used for the linear dynamic model and non-linear measurement model is given below. Here, the estimate $x_i(+)$ given in (13) can be improved by repeatedly calculating $\bar{x_i}(+)$, G_i , and $C_{\bar{x}}(+)$ and each time linearising about the most recent estimate. As many iterations can be performed as are necessary to reach the point where there is no significant change in consecutive iterates. However, it should be recognised that each iteration contributes to the computation time required to mechanise the filter.

A detailed explaination about the algorithms can be found in Mahmud (1991).

1) Initialise
$$\overline{x}_{i-1}(+) = x_0$$
 ... (14)

and
$$C_{\vec{x}_{i-1}}(+) = C_{x_0}$$
 ... (15)

i.e an a prior estimate of the state vector and its covariance is assumed to be known.

2) Increment i, i := i + 1

3)
$$\overline{x}_i(-) = M_{i-1,i}\overline{x}_{i-1}(+)$$
 ... (16)

4)
$$C_{\bar{x}_i}(-) = M_{i-1,i}C_{\bar{x}_{i-1}}(+)M_{i-1,i}^T + C_y$$
 ... (17)

5) Start iteration

$$G_i = C_{\bar{x}_i}(-)A_i^T (A_i C_{\bar{x}_i}(-)A_i^T + +W_i^{-1})^{-1}$$

. . . (18)

6)
$$\overline{x}_i(+) = \overline{x}_i(-) + G_i(l_i - comp(\overline{x}_i(-)))$$
 ... (19)

7)
$$C_{\bar{x}_i}(+) = (I - G_i A_i) C_{\bar{x}_i}(-)$$
 ... (20)

8) Return to step 5 (i.e. until a certain stopping criterion is met)

$$\overline{x}_i(-) := \overline{x}_i(+)$$
 ... (21)

and

$$C_{\overline{x}_i}(-) := C_{\overline{x}_i}(+) \qquad \qquad \dots (22)$$

9) Return to step 2.

3.0 DATA SIMULATOR

The tests carried out in this paper involved simulated data. Hence it is possible to monitor accurately the performance of the Kalman filter because the true positions of the ship are known. Figure 1 shows the simulated ship's path and the six transponders station. The true coordinates of the six transponders are shown in Table 1. Errors of \pm 100 metres were introduced to all the transponders coordinates except those chosen as fixed coordinates in the relative calibration test.

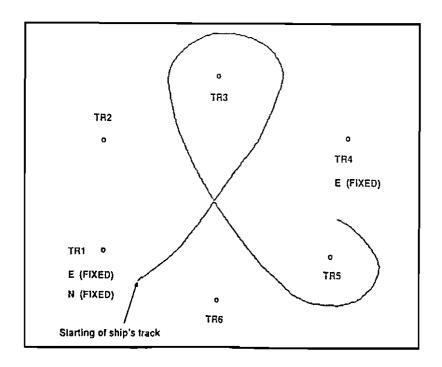


Figure 1. Simulated ship's path and transponders position

	Coordinates (metres)									
Station	Eastings	Northings								
TR1	642544.9554	6264513.4918								
TR2	642585.8109	6267405.8134								
TR3	644967.0925	6269116.0998								
TR4	647551.9617	6267446.0043								
TR5	647185.9108	6264329.8162								
TR6	644905.5715	6263209.8709								

Table 1. True coordinates of transponders

4.0 RELATIVE CALIBRATION TEST

4.1 Observables and Provisional Coordinates of TranspondersThe provisional coordinates of the transponders are shown in Table 2. Both coordinates for station 1 and the easting coordinate of station 4 were held fixed to give scale and orientation to the array of transponders. Note that in a real situation, the fixed coordinates may be known from previous surveys or arbitrary values can be assigned. In this case the true simulated coordinates were used. Hence it was expected that once all the errors had converged, the coordinates obtained would be the true simulated coordinates. Only the observables form underwater acoustic transponders were used in this test.

	Coordinates (metres)										
Station	Eastings	Northings									
TR1	642544.9554 (FIXED)	6264513.4918 (FIXED)									
TR2	642485.8109	6267305.8134									
TR3	645067.0925	6269016.0998									
TR4	647551.9617 (FIXED)	6267546.0043									
TR5	647085.9108	6264429.8162									
TR6	644805.5715	6263309.8709									

Table 2. Provisional coordinates of transponders

4.2 Relative Calibration Results

Tables 3, 4, 5 and 6 show the precision results of epoch 1, 100, 200 and 300 respectively. They show the results of the ship's parameters, error ellipse and relative error ellipse of the transponders position including their coordinates and standard errors. Note that the notation 'NONE' in the tables means that no value is given due to the station coordinates being held fixed and the notation 'DIFF' is the difference between consecutive coordinate results. The coordinates of the transponders shown in the tables can be compared with the true values shown in Table 1. Figures 2, 3, 4, 5 and 6 show the graphs of errors in coordinates against the fix number. The results showed that all the errors of the transponders coordinates converged from ± 100 metres to less than 1.1 metres after 100 epochs except the easting coordinate for transponder 3 (i.e. less than 2.6 metres). Clearly it can be seen that all the transponders coordinates managed to converge satisfactorily as the number of epoch increases. The values of the standard errors for distances and the semi-major and semi-minor axes of the relative error ellipses in Tables 5 and 6 for epoch 200 and 300 respectively, are less than 0.1 metres. These values can be used to decide to end the calibration process in a real situation. For this particular test, it was found that the calibration process could be terminated when the values of standard errors of distances and the semi-major and semi-minor axes of the relative error ellipses were less than 0.1 metres since the error of the transponder coordinates had converged to a satisfactory limit.

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Table 3. Precision results for epoch 1

Table 4. Precision results for epoch 100

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Table 5. Precision results for epoch 200

Table 6. Precision results for epoch 300

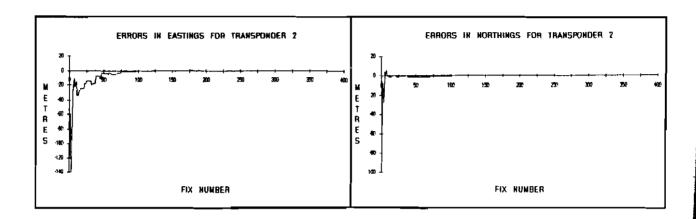


Figure 2. Relative calibration: Errors in coordinates for transponder 2

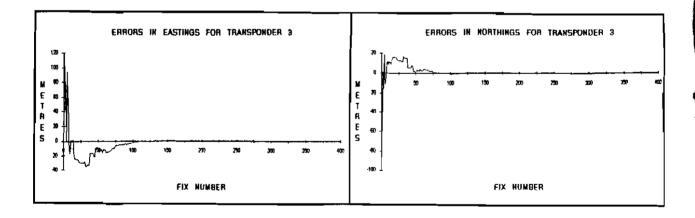


Figure 3. Relative calibration: Errors in coordinates for transponder 3

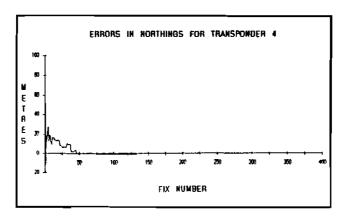


Figure 4. Relative calibration: Errors in coordinates for transponder 4

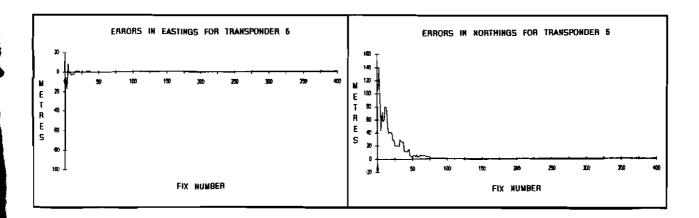


Figure 5. Relative calibration : Errors in coordinates for transponder 5

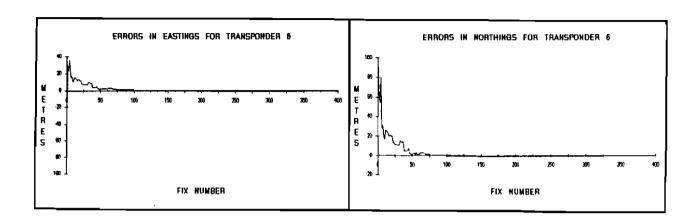


Figure 6. Relative calibration: Errors in coordinates for transponder 6

5.0 ABSOLUTE CALIBRATION TEST

5.1 Observables and Provisional Coordinates of TranspondersThe provisional coordinates of the transponders for the absolute calibration test are shown in Table 7. In this test there was no fixed coordinate defined. All the observables from four systems (i.e. DGPS, underwater acoustic system, range-range system and hyperbolic system) integrated together were used in this test.

···	Coordinates (metres)									
Station	Eastings	Nothings								
TR1	642644.9554	6264413.4918								
TR2	642485.8109	6267305.8134								
TR3	645067.0925	6269016.0998								
TR4	647651.9617	6267546.0043								
TR5	647085.9108	6264429.8162								
TR6	644805.5715	6263309.8709								

Table 7. Provisional coordinates of transponders

5.2 Absolute Calibration Results

Tables 8, 9, 10 and 11 show the precision results for epoch 1, 100, 200 and 300 respectively. The graphs of errors in coordinates against the fix number in Figures 7, 8, 9, 10, 11 and 12 show that all the errors of the transponder coordinates converged from ± 100 metres to less than 1.1 metres after 100 epochs except the easting coordinate for transponder 3 (i.e. less than 2.7 metres) and the northing coordinate for transponder 4 (i.e. less than 4.9 metres). Clearly it can be seen that all the transponders coordinates managed to converge satisfactorily as the number of epoch increases. Tables 10 and 11 show that the values of standard errors of distances and the semi-major and semi-minor axes of the relative error ellipses for epoch 200 and 300 respectively are less than 0.1 metres. Note that in certain circumstances the above interpretation may be quite different. The small values of the relative error ellipse does not mean that the calibration process has reached its satisfactory limit due to the transponders achieving correct 'relative positioning' but not achieving correct absolute position in the required coordinate system. This can happen when there are more acoustic than other observables. Thus the quality and number of observables received from DGPS, range-range system and hyperbolic system are important to achieve high quality absolute calibration.

5HIF	PARAI	IE TERS								l suit	FMRI	METERS	1						
EPOC		EAST (NG		HORTHING		INE	HEADING		(knots)	EPOCI			TING	MORTHIN		INE	KEADING	SPEED	(knot
ı	ı	43320.522	626	3715.126	2	0 0	49,39	5 8.09	5	100		645965	.927	6267845.24	9 2	19 48	15.01	8.131	
COOR	DINATI	OF BEACON	5							COOR	DINA	TE OF 8	EACONS						
HO.	STM.	EASTIN	_	SID.ER	DIFF		NORTHING	STD.ER	DIFF	MO.	SIN		ASTING	STD.ER	DIFF		NORTHING	STD.ER	DIFF
	tri	642548.1	49	35.756	96.813		264513.341	34.664	99.849							,	157	0.089	0.032
] 2	trz	642468.8		48.695	16.913		267378.527	11.347	72.713	1 !	trl		545.157	0.040	0,003 0,624		64514.157 67406.397	0.085	0.009
3	trā	645086.5	-	47.480	19.417		269075.027	15.654	58.927	3	tr? tr3		:595.339 !964.465	0.142 0.248	0.129		69116.797	0.056	0.028
4	tr¢	647544.3	• •	33.110			267450.891	37.444	95.113	1 7	tr#		551.027	0.163	0.224		67450.824	0.210	0.290
š	tr5	647165.7		9.322	79.847		264444.944	49.076	15.128	;	tr5		186.133	0.042	0.004		64330.703	0.153	0.068
6	tr6	644928.4		13.162	122.884	•	263276.211	48.008	33,660	، ا	tré		906.579	180.0	0.016		63210.355	0.056	0.010
ABS0	LVTE 1	RROR ELLIF	SES							ABSO	LUTE	ERROR	ELL LPSE:	ı					
MO.		MOLTA	MAJO		11MOR		ENTATION			WO.		MOTATION		-	WIKOR		MOTTATION		
ı		tri	49,79		0.701		45.89			١,		tri		. 090	0.039	17	1.38		
2		tr2	49.99		0.699		76.91			1 2		tr2	(. 160	0,044	6	1.71		
3		tr3	49.99		3,561		108.24			3		tr3	(. 252	0.030	10	0.71		
4		tr4	49.98		0.525		138.52			4		tr4	(.265	0.026	14	2.28		
5		tr5	49.94	•). 63B	i	169.27			5		tr5		, 153	0.04L		3.36		
4		tr6	49,77	4 (0.744		15.31			4		tré	(,086	0.045	6	3.83		
RELA	T JVE	STANDARD EI	RORS"							RELA	11YE	STANDA	PP ERROI	\$					
¦ LI	HE.		. ERRO		•		E ERROR ELL		}	1 1 11	WE		¦ STO. E	RROR OF	; R	ELATIVE	ERROR ELLI	PSES .	ł
; fx		- •	TANCE	BEARL		10%		DR1ENTATION	i	; FRI		TO	; DISTAN			AJOR		RIENTATION .	ţ
tr	1		. 109	0.02			18.876	61.47		tri	1	tr2	0.03	7 0.00	0 0	.146	0.055	83.83	
tr	1		. 594	0.010			36.530	77.28		tri	1	tr3	0.06	7 0.00	0 0	. 272	0.063	112.70	
tr	1	tr4 45	. 325	0.009			48.731	179.83		tri	Į.	tr4	0.04	0.00	0 0	.316	0.040	149.32	
tr	-		. 638	0.013			33.453	17.44		l tri	1	tr5	B. 02	7 0.00	0 0	221	0.027	2.34	
tr	1		, 717	0.025			18.575	30.61		tri	i	tr6	0.03	3 0.00	0 0	.134	0.033	28,43	
tr			, 509	0.014			19,101	92.57		tra	7	tr3	0.19	1 0.00	0 0	. 180	0.090	119.85	
tr	-	•••	.716	0.000			36.211	107.70		tra	2	tr4	0.10	6 0.00	0 0	274	0.099	170.93	
tr	_		. 936	0.009			48.934	33.71		tra	?	tr5	0.04	8 0.00		263	0.048	33.65	
tr			.478	0.017	-		36.125	46.34		tr:	2	trá	0.02	8 0.00	0 0	232	0.027	60.17	
tr			.237	0.000			10.473	123.37		tr:	3	tr4	0.19	4 0.00	0 0	. 195	0.167	134.16	
tr	-		, 138	0.000			35.892	138.71		tri	3	tr5	0.09	2 0.00	0 0	.268	0.088	71.46	
tr	-		.434	0.009			48.593	149.31		tra		tr6	0.04	0.00	0 0	304	0.039	91.44	
tr			.869	0.013		137	18.748	153.98		tre	1	tr5	0.13	0.00	0 0	.203	0.093	126.89	
4-	4		.717	0.016		052 703	33.553	166.73 2.24		tre	ţ	tr6	0.07	6 0.00	0 0	.277	0.066	130.26	
tr t:		tr6 36					15.898												

Table 8. Precision results for epoch 1

Table 9. Precision results for epoch 100

nsr	PARAN	CICAS								antr	PHEN	METERS							
POCI		EASTING		ORTHING		EME	HEADING		(knots)	EPOCI		EASTING		MORTHING		TIME	HEADIH		(knots
200	6-	13939,342	6268	781.252	2	39 48	172.33	8.194		390		645794 . 510	62	64195.385		2 59 4	8 146.5	4 8.016	
g Q R	DTNATE	OF BEACO	4S							CODR	DIKAT	E OF BEACO	MŚ						
0.	STM.	EASTI		id.ER	DIFF	,.	ORTHING	STD.ER	DIFF	NO.	SIN.	EASTI		SID.ER	01	er 	NORTHING	STO.ER	DIFF
ı	trl	642545.	160	0.039	0.081	626	¢51¢,128	9.047	0.005	1,	iri	642545.	181	0.038	0.0	103	6264514.062	0.044	D. GO2
2	tr2	642585.	979	0.052	0.001		7406.618	0.042	0.002	2	tr2	642585		0.048	0.0	101	6267406.501	0.040	0.001
3	tr3	644967.		0.078	0.009		9116.330	0.034	0.000	3	tr3	644966.	920	0.071	0.0		6269116.345	0.832	0.005
1	tr4	647552.		0.052	0.004		7449.324	0.065	0.005	1 1	tr4	647552.		0.048	0.0		6267448.338	0.061	0.003
6	tr5 tr6	647186. 644906.		0.040 6.057	0.001 0.003		4330.343 3210. 05 6	0.064 0.037	8.001 0.001	5 6	tr5 tr6	6471B6. 644906.		0.039 0.055	0.0		6264330.268 6263218.028	0.060 0.035	0.001
NO S C	LUTE EI	RROR ELLI	PSE S							ARSO	LUTE	ERROR ELLI	PSE S						
Kā.		ATION	MAJOR		IOR	ORIEN	TATION			NO.		TATION	MAJ		RONI	•	MOLITATION		
1		tr1	0.054	0.1	30	145	.15					trl	0.0		.028		142.84		
2		tr2	0.053	0.4			.93			1 2		tr?	0.0		.019		75.80		
3		tr3	0.081	0.4	28	105	. } 6			3		tr3	0.0	74 0	.026		106.00		
•		tr4	0.079		25	143				4		tr4	0.0		.024		143.80		
6		tr5 tr6	0.66# 0.057		040 037		. 53 . 10			6		tr5 tr6	0.0 0.0		.039 .035		6.50 97.33		
RELA	FIVE 5	TANDARD E	RORS							RELA	T I VE	STANDARD E	RRORS						
Lli			, ERROR		1 "-"		ERROR ELLI		! !	i Ui			D. ERR		. !		YE ERROR ELL		!
FR			TANCE	BEARING	; na.			IRIENTATION	i	, FR		,	STANCE	MIRASO		MAJOR		ORIENTATION	i
tr:			0.037	0.000	0.0		0.036	81.98 113.22		tr			0.035	0.006		0.050	0.034	82.91	
tr:).037).029	0.000 0.000	0.1 0.1		0.036	150.01		tr	•	***	0.034	0.000		0.083	0. 033 0.027	113.81 149.98	
tr:	-		1.024	0.000	0.0		0.026	2.47		tr tr	-		0.027 0.025	0.000		0.090 0.076	0.027	2,68	
tr.			1.032	0.000	0.0		0.031	34.32		tr tr	-		0.023 Q.031	0.000		0.053	0.023	35.88	
tr			1.046	0.000	0.0		0.038	114.22		tr	_		0.042	0.000		0.059	0.036	114.17	
ţr	2 1	ir4 (.042	0.000	0.1	384	0.041	173.77		tr		-	0.038	0.000		0.079	0.037	174,13	
tr	- '		.034	0,000	0.0		0.033	30.80		tr		tr5	0.030	0.000		0.086	0.030	30.72	
tr			1.026	0.000	0.1		0.026	60.37		tr			0.025	0.000		0.079	0.025	60.60	
tr.			.053	0.000	0.0		0.053	47.38		tr			0.050	0.000		0.053	0.050	45.74	
Ir.			.042	9,000	0.0		0.042	65.36		tr		tr5	8,040	0.000		0.085	0.040	65.14	
tr.			0.030 0.044	0.000	0.1		0.030 0.944	89.02 110.45		tr	-		0.028	0.000		0.097	0.028 0.042	88.34 107.18	
tre			1.040	0.000	9.1		0.040	120.16		tr			0.043	0.000		0.081 0.053	0.037	119.97	
LI	• 1		.042	0.000		253	0.041	168.05		ţг	4		0.037	0.000		0.050	0.010	169.88	

Table 10. Precision results for epoch 200

Table 11. Precision results for epoch 300

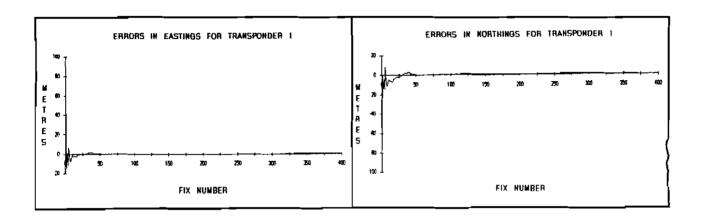


Figure 7. Absolute calibration: Errors in coordinates for transponder 1

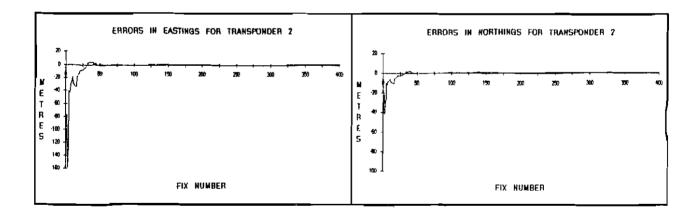


Figure 8. Absolute calibration: Errors in coordinates for transponder 2

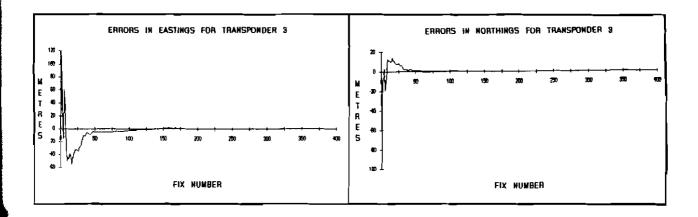


Figure 9. Absolute calibration : Errors in coordinates for transponder 3

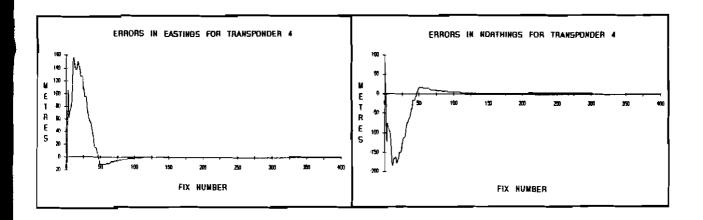


Figure 10. Absolute calibration : Errors in coordinates for transponder 4

7

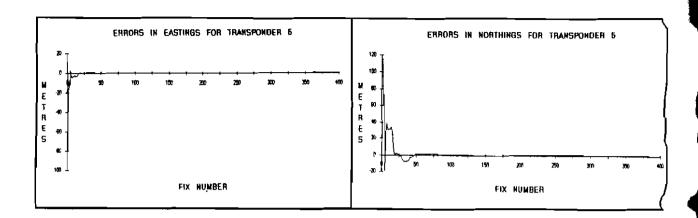


Figure 11. Absolute calibration: Errors in coordinates for transponder 5

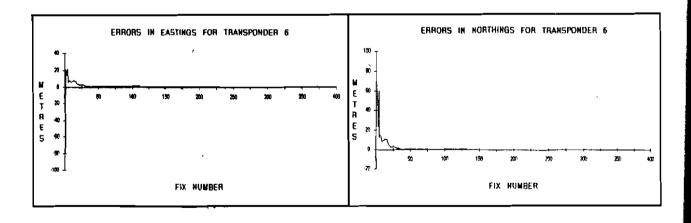


Figure 12. Absolute: Errors in coordinates for transponder 6

6.0 CONCLUDING REMARKS

The tests carried out for relative and absolute calibrations showed that in both cases the errors introduced in the transponders coordinates managed to converge to a satisfactory limit. This was only when the values of the standard errors of distances, semi-major and semi-minor axes of the relative error ellipses were less than 0.1 metres. Both these results suggested that absolute calibration could be carried out without relative calibration if the quantity of observations from the underwater acoustic system are the same or greater than the rest of the positioning systems employed.

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About the author

Mohd. Razali Mahmud is a lecturer at the Department of Land Surveying, Faculty of Surveying and Real Estate, UTM. Currently he is also the Director of Center for Hydrographic Studies (CHS), UTM. He has been with the department since 1982. He obtained B.Sc.(Hons.) Surveying and Mapping

Sciences from thw North East London Ploytechnic, United Kingdom and M.Phil. from the University of Newcastle upon Tyne, U.K. His research interest is in the areas of hydrographic surveying and offshore positioning.