

# Approximate Parametrisation of Confidence Sets<sup>\*</sup>

Zbyněk Šír

Charles University, Sokolovská 83, Prague, Czech Republic, [sir@karlin.mff.cuni.cz](mailto:sir@karlin.mff.cuni.cz)

**Abstract.** In various geometrical applications the analysis and the visualization of the error of calculated or constructed results is required. This error has very often character of a nontrivial multidimensional probability distribution. Such distributions can be represented in a geometrically interesting way by a system of so called confidence sets. In our paper we present a method for an approximate parametrisation of these sets. In the section 1 we describe our motivation, which consists in the study of the errors of so called Passive Observation Systems (POS). In the section 2 we give a result about the intersection of quadric surfaces of revolution, which is useful in the investigation of the POS. In the section 3 we give a general method for an approximate parametrisation of the confidence sets via simultaneous Taylor expansion. This method, which can be applied in a wide range of geometrical situations, is demonstrated on a concrete example of the POS.

## 1 Motivation

Our research was motivated by concrete problem of the analysis and the visualization of the errors of so called Passive Observation Systems (POS).

### 1.1 Passive Observation Systems

The POS have been successfully constructed and produced in Czech Republic since the 1960's as an alternative to the classical radars. These systems, which do not transmit any signal (therefore passive), are based on the principle of the time difference. A pulse in the transmission of an object (a plane) is received at four (or more) observation sites. In practice any plane is forced to transmit some signals, at least in order to ensure its orientation. From the differences of the time of reception of the pulse the position of the object can be determined.

The POS have two main advantages comparing to the standard radars. As they do not transmit any signal they can not be itself detected and have very low energy consumption.

In addition the error of the POS has a different characteristic comparing to the classical radars. For this reason a simultaneous use of the POS and the classical radars can be very interesting. For more details about the principle of the POS and for the basic information about their precision see [1, Chapter 5].

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## 1.2 Geometry of POS

The construction of POS creates many difficult problems on the level of the electrical engineering, but the underlying geometry is quite simple. Let a pulse transmitted by an object  $X$  be received at the sites  $A$  and  $A'$  respectively at times  $t_A$  and  $t_{A'}$ . Multiplying the difference  $t_A - t_{A'}$  by the speed of the signal (typically the speed of light) we get the difference  $d_{AA'}$  of distances from the object  $X$  to the sites  $A$  and  $A'$ . The object  $X$  must therefore lie on one of sheets of the two-sheet hyperboloid of revolution, which is determined by its foci  $A, A'$  and the measured difference of distances  $d_{AA'}$ . The sign of  $d_{AA'}$  indicates which of the two sheets must be taken.

Repeating the same procedure for two other pairs of sites  $(B, B')$  and  $(C, C')$ , we get in all three hyperboloids on which the object  $X$  must lie and its position can be therefore determined as their intersection. The space coordinates  $[x_1, x_2, x_3]$  of  $X$  are then computed from the measured distance differences  $d_{AA'}$ ,  $d_{BB'}$  and  $d_{CC'}$ .

The difference vector  $[d_{AA'}, d_{BB'}, d_{CC'}]$  can be easily computed from  $[x_1, x_2, x_3]$ , and the corresponding mapping  $F : [x_1, x_2, x_3] \rightarrow [d_{AA'}, d_{BB'}, d_{CC'}]$  can be explicitly expressed. If the sites  $A, A'$  have the space coordinates  $[a_1, a_2, a_3]$  and  $[a'_1, a'_2, a'_3]$  respectively, then for example

$$d_{AA'} = \sqrt{(x_1 - a_1)^2 + (x_2 - a_2)^2 + (x_3 - a_3)^2} - \sqrt{(x_1 - a'_1)^2 + (x_2 - a'_2)^2 + (x_3 - a'_3)^2}.$$

On the other hand the inversion mapping  $F^{-1}$  can not be in general expressed explicitly and the position of  $X$  must be computed from  $[d_{AA'}, d_{BB'}, d_{CC'}]$  numerically as a solution of a system of algebraic equations of the total degree 8.

In practice a network of observation sites should be used. But the smallest operational system consists of four sites only. In this case one site  $O = A' = B' = C'$  is considered as central one and the position of the object  $X$  is computed from the distance differences  $[d_{AO}, d_{BO}, d_{CO}]$ . In the sequel we will restrict ourselves to this simplest case. As we will show, in this case an explicit inversion formula for  $F^{-1}$  can be always given.

## 1.3 Measurement error of the POS

Suppose, that a pulse is received at four observation sites  $O, A, B$  and  $C$  at times  $t_O, t_A, t_B$  and  $t_C$ . The error of the vector  $[t_O, t_A, t_B, t_C]$  of independently measured times can be well modeled by a multivariate normal distribution, characterized by its mean value  $[0, 0, 0, 0]$  and the variation-covariation matrix having on the diagonal the variations of the time errors at the four sites, which are not necessarily the same

$$\begin{bmatrix} \sigma_O^2 & 0 & 0 & 0 \\ 0 & \sigma_A^2 & 0 & 0 \\ 0 & 0 & \sigma_B^2 & 0 \\ 0 & 0 & 0 & \sigma_C^2 \end{bmatrix}. \quad (1)$$

The differences  $d_{AO}$ ,  $d_{BO}$  and  $d_{CO}$  have no more independent errors, but the error of the vector  $[d_{AO}, d_{BO}, d_{CO}]$  has still a normal distribution characterized by its mean value  $[0, 0, 0]$  and the variation-covariation matrix

$$c^2 \begin{bmatrix} \sigma_A^2 + \sigma_O^2 & \sigma_O^2 & \sigma_O^2 \\ \sigma_O^2 & \sigma_B^2 + \sigma_O^2 & \sigma_O^2 \\ \sigma_O^2 & \sigma_O^2 & \sigma_C^2 + \sigma_O^2 \end{bmatrix}, \quad (2)$$

where  $c$  is the speed of light. See [3] for the details about multivariate distributions and their characteristics.

If we compute the position  $[x_1, x_2, x_3]$  using  $d_{AO}$ ,  $d_{BO}$  and  $d_{CO}$  we transform the error distribution by the mapping  $F^{-1}$ . The transformed distribution will be no more normal. For this reason the mean value and the variation-covariation matrix are no more sufficient characteristics of this transformed error distribution.

In fact the analysis of such complex multivariate distributions is a difficult problem. This is due to the fact that the standard concepts used in the case of one dimensional distributions, are insufficient for the description of the geometry of the multivariate distributions. We are convinced that the methods of the applied geometry would be very useful in the analysis of both theoretical distributions and experimental data. See [2] for one possible approach based on the concept of the data depth.

#### 1.4 Confidence sets

The confidence sets (called also tolerance regions) are perhaps geometrically the most interesting characteristics of probability distributions.

**Definition 1.** For a given random variable  $U$  having the density function  $p_U$  and for a given probability  $\alpha \in (0, 1]$  we define the confidence set  $C_{U,\alpha}$  as a region for which

$$\int_{x \in C_{U,\alpha}} p_U(x) = \alpha \quad (3)$$

In other words a confidence set is a region in which the random variable  $U$  lies with the probability  $\alpha$ . In practice  $\alpha$  is set quite high, for example 0.99, and thus a confidence set is simply a region in which the random variable lies with a reasonable certitude.

It is clear from the definition, that for a given probability  $\alpha < 1$  there is in general more then one confidence set. There are natural additional properties which can be required of the confidence sets. First of all the confidence sets should be as small as possible in order to give good information about the probability density. For the same reason their boundaries should be the iso-lines (iso-surfaces) of the density function. In the case of multivariate normal distributions it is customary to use suitable ellipsoids as confidence sets. These ellipsoids satisfy both additional requirements (see for example [3, 45.9]).

The distribution of the error of the vector  $[d_{AO}, d_{BO}, d_{CO}]$  can be described by a system of ellipsoids (confidence sets) depending on the probability  $\alpha$  and

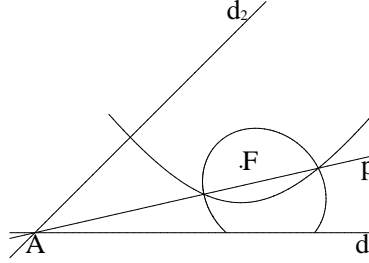
on the values  $[d_{AO}, d_{BO}, d_{CO}]$  (the error may in general depend on the value of  $[d_{AO}, d_{BO}, d_{CO}]$ ). Transforming this system by  $F^{-1}$  we will get a new system of confidence sets describing the error of the position  $[x_1, x_2, x_3]$ . The boundaries of these new confidence sets will be iso-surfaces of the new density function.

## 2 Explicit inversion formula

The importance of an explicit formula for  $F^{-1}$  is obvious from the previous section. As  $X$  is obtained as intersection of three quadric surfaces, the resulting system of equations has degree 8. Therefore there is seemingly no possibility to obtain an explicit expression for  $F^{-1}$ . However for concrete examples, we were able to reduce the degree of the problem and even obtain a simple explicit formula. A deeper investigation of this fact has shown, that this simplification is due to the following interesting property.

### 2.1 Intersection of quadric surfaces of revolution

**Proposition 1.** *Let  $S_1, S_2$  be two quadric surfaces of revolution, each of which obtained by rotating a conic section around its main axis. (The only axis for a parabola and the axis passing through the foci for an ellipse or an hyperbola.) Suppose that  $S_1$  and  $S_2$  have a common focus. Then their intersection can be decomposed into curves of degree 2.*



*Proof.* Let  $F$  be the common focus. Clearly the axes of  $S_1$  and  $S_2$  intersect in the point  $F$  and therefore they lie in a plane. The previous figure represents this plane and its intersections with all mentioned objects.

We can characterize the surfaces  $S_1$  and  $S_2$  using the focus-directrix property of the generating conic sections. Obviously the surface  $S_1$  is precisely the set of points in the space, having a constant ratio of distances to the focus  $F$  and a directrix plane  $d_1$ , perpendicular to the main axis:  $S_1 = \{X, \frac{|XF|}{|Xd_1|} = r_1\}$  for some fixed ratio  $r_1$ . For  $r_1 = 1$  we get a paraboloid, for  $r_1 < 1$  an ellipsoid and for  $r_1 > 1$  a two-sheet hyperboloid. In the same way the surface  $S_2$  can be characterized as the set  $S_2 = \{X, \frac{|XF|}{|Xd_2|} = r_2\}$  for some plane  $d_2$  perpendicular to the axis of  $S_2$  and for some fixed ratio  $r_2$ .

For the points of the intersection  $X \in S_1 \cap S_2$  we thus get  $\frac{|Xd_1|}{|Xd_2|} = \frac{r_2}{r_1}$ . This equality characterizes all the points lying in two planes passing through

the intersection  $d_1 \cap d_2$ . One of these planes is denoted  $p$  on the figure. As the intersection of a quadric surface with a plane is of degree 2, the intersection  $S_1 \cap S_2$  must have a component of degree 2. As  $S_1 \cap S_2$  is itself of degree 4, the proposition is proved.  $\square$

So the intersection of two hyperboloids, which is in general a curve of degree 4, will have components of degree 2 (conic sections) if the two hyperboloids share a focus.

Consequently the degree 8 system describing a general POS will be decomposed if two of hyperboloids have a common focus. If particular if the three hyperboloids have a common focus - the site  $O$  - the problem will be reduced twice and the resulting system can be decomposed to the degree 2 systems. In this case therefore an explicit inversion formula can be always obtained.

We will not describe this explicit formula in general, but we will study in detail one particular example. In this example the situation is simplified even more by the additional condition, that the four sites  $A, B, C, O$  are coplanar.

## 2.2 Example

Let us consider the POS in which the four sites lie in one plane and have the coordinates:  $O = [0, 0, 0]$ ,  $A = [30, 0, 0]$ ,  $B = [-15, 26, 0]$  and  $C = [-15, -26, 0]$ . The mapping  $F$  is then expressed by formulae:

$$\begin{aligned} d_{AO} &= \sqrt{x_1^2 - 60x_1 + 900 + x_2^2 + x_3^2} - \sqrt{x_1^2 + x_2^2 + x_3^2} \\ d_{BO} &= \sqrt{x_1^2 + 30x_1 + 901 + x_2^2 - 52x_2 + x_3^2} - \sqrt{x_1^2 + x_2^2 + x_3^2} \\ d_{CO} &= \sqrt{x_1^2 + 30x_1 + 901 + x_2^2 + 52x_2 + x_3^2} - \sqrt{x_1^2 + x_2^2 + x_3^2} \end{aligned} \quad (4)$$

We implicitise these equations and obtain implicit algebraic equations of the three hyperboloids  $H_{AO}$ ,  $H_{BO}$  and  $H_{CO}$ . For example the implicit equation of  $H_{AO}$  is

$$4d_{AO}^2 (x_1^2 + x_2^2 + x_3^2) - (900 - 60x_1 - d_{AO}^2)^2 = 0.$$

Due to the Proposition 1 any two of these hyperboloids intersect in two conic sections. Because of the symmetry with regard to the plane  $x_3 = 0$  these conics lie in the planes perpendicular to the plane  $x_3 = 0$ . Their projections to this plane will be therefore lines.

For the determination of the 8 intersections of the hyperboloids  $H_{AO}$ ,  $H_{BO}$  and  $H_{CO}$  we first evaluate the resultant with respect to  $x_3$  of the implicit equations of  $H_{AO}$  and  $H_{BO}$ . Because of the previous observations this resultant (of degree 4 in  $x_1, x_2$ ) can be factorised in two linear factors (each of them of with multiplicity two) describing two stright lines  $p_1$  and  $p_2$ . In a similar way from the equations of  $H_{AO}$  and  $H_{CO}$  we get two lines  $q_1$  and  $q_2$ . As intersection of this two pairs of lines we get four points  $X_{i,j} = p_i \cap q_j$ ,  $i, j = 1..2$ , each of them being projection of two symmetrical intersections of the three hyperboloids. The signs of  $d_{AO}$ ,  $d_{BO}$  and  $d_{CO}$  will indicate which of the four points  $X_{i,j}$  must be taken. The last coordinate  $x_3^i$  can be calculated from the equation of any of the three hyperboloids.

Let us give the explicit formula of one of the 4 pairs of solutions of our example system (4):

$$\begin{aligned} x_1 &= \frac{d_{AO}(d_{BO}^2 + d_{CO}^2 - 1802) + (900 - d_{AO}^2)(d_{BO} + d_{CO})}{60(d_{AO} + d_{BO} + d_{CO})} \\ x_2 &= \frac{d_{AO}(d_{CO}^2 - d_{BO}^2) + (d_{AO}^2 - 2d_{BO}d_{CO} - 2702)(d_{BO} - d_{CO})}{104(d_{AO} + d_{BO} + d_{CO})} \\ x_3 &= \pm \frac{\sqrt{P_6(d_{AO}, d_{BO}, d_{CO})}}{d_{AO} + d_{BO} + d_{CO}} \end{aligned} \quad (5)$$

where  $P_6(d_{AO}, d_{BO}, d_{CO})$  is a polynomial of degree 6 in  $d_{AO}$ ,  $d_{BO}$  and  $d_{CO}$ . The  $x_3$  is usually supposed to be positiv, as the object (plane) is usually "over" the observation sites.

A similar explicit form of  $F^{-1}$  can be in general obtained for any POS having the four sites in a plane. In this case, the first two coordites  $x_1, x_2$  can be expressed as rational functions in  $d_{AO}$ ,  $d_{BO}$  and  $d_{CO}$ , but the expression of  $x_3$  will involve a square root.

If the four sites are not coplanar, an explicit formula can be still obtained, but square roots will appear in the expressions of all coordinates.

For a general POS, based on three independent pairs of sites  $(A, A')$ ,  $(B, B')$  and  $(C, C')$ , no closed expression of  $F^{-1}$  can be obtained.

### 3 Approximate representation

The explicit inversion formula is not available for the POS in the general position. In some other cases the inversion formula can be too complicated. For this reason we will describe in this section a general method for the approximation of  $F^{-1}$ .

#### 3.1 General setting

Let us consider the following general setting. Suppose that  $x = [x_1, \dots, x_n]$  is a set of parameters which is transformed by a local diffeomorphism  $F$  to a second set of parameters  $y = [y_1, \dots, y_n]$ :

$$F : [x_1, \dots, x_n] \rightarrow [y_1, \dots, y_n] \quad (6)$$

Suppose in addition that an algebraic implicitisation of  $F$  is available. We mean by this a system of algebraic equations

$$G(x, y) = 0 \quad (7)$$

which hold if and only if  $y = F(x)$ .

Next suppose that in the space of parameters  $y$  the system of confidence sets (for example a system of ellipsoids) is described. We want to obtain a description of the transformed system of the confidence sets in the space of the parameters  $x$ .

### 3.2 Implicit representation

If the confidence sets in the space of parameters  $y$  are described implicitly we can obtain an implicit description in the space of parameters  $x$  in a straightforward way. Suppose, that the boundaries of the confidence sets in the space of parameters  $y$  are given by implicit equations

$$E_{\alpha, \bar{y}}(y) = 0 \quad (8)$$

depending algebraically on the measured value  $\bar{y}$ . Then substituting  $y = F(x)$  and  $\bar{y} = F(\bar{x})$  in this equations we get implicit representations of the boundaries of the confidence sets in the space  $x$  depending on  $\bar{x}$ .

The drawbacks of this methods are obvious. As the transformation  $F$  is not necessarily rational, we obtain in general a complicated (non algebraic) implicit representation depending in a complicated way on  $\bar{x}$ .

### 3.3 Approximation by the Taylor expansion

Another natural possibility is to approximate the inversion  $F^{-1}$  by its Taylor expansion in a suitable point  $\bar{y}$ :

$$F^{-1}(y) = F^{-1}(\bar{y}) + D_1 F_{\bar{y}}^{-1}(y - \bar{y}) + \frac{1}{2} D_2 F_{\bar{y}}^{-1}(y - \bar{y}) + \frac{1}{6} D_3 F_{\bar{y}}^{-1}(y - \bar{y}) + \dots \quad (9)$$

where  $D_i F_{\bar{y}}^{-1}$  is the  $i$ -th total differential of  $F^{-1}$  at the point  $\bar{y}$ . See [4, par. 3.14] for the details about the multivariate Taylor expansion. The value of  $\bar{x} = F^{-1}(\bar{y})$  can be calculated numerically from (7) and the operators  $D_i F_{\bar{y}}^{-1}$  can be obtained by the implicit differentiation of (7), or from the known partial derivatives of  $F$  at the point  $\bar{x}$ . This approximation can be used for an approximate representation of the confidence sets in the space of parameters  $x$ . In particular if we have a parametrisation of the boundaries of the confidence sets in the space of parameters  $y$ , we can compose this parametrisation with the Taylor expansion and this way obtain an approximate parametrisation of the boundaries of the confidence sets in the space of parameters  $x$ .

The disadvantage of this approach is that the Taylor expansion can give a sufficiently good approximation in the proximity of the point  $\bar{y}$  but will not be sufficient for more distant points.

### 3.4 Symbolic Computation of the Taylor expansion

We propose a different approach, which consists in the symbolic computation of the Taylor expansion simultaneously in all points. If the mapping  $F^{-1}$  can not be expressed explicitly, there is no hope to get a general expression of the Taylor expansion depending on the point  $\bar{y}$ . On the other hand it is possible to get such general expression depending on the target point  $\bar{x} = F^{-1}(\bar{y})$ .

The total differentials  $D_i F_{\bar{y}}^{-1}$  can be symbolically computed via partial differentiation of the equality

$$G(F^{-1}(y), y) = 0 \quad (10)$$

For example by taking all the partial derivatives of the first order  $\frac{\partial}{\partial y_i}$  for  $i = 1..n$ , we obtain a system of  $n$  linear equations for  $n$  unknown partial derivatives  $\frac{\partial F^{-1}}{\partial y_i}$ . The coefficients of these equations are polynomials in  $y$  and  $F^{-1}(y)$ . This system can be symbolically solved and we get  $\frac{\partial F^{-1}}{\partial y_i}$  in the form of a rational function of  $y$  and  $F^{-1}(y)$ . If we use in a similar way the higher partial derivatives of (10), we get the same kind of expression for the higher partial derivatives. See [4, par. 4.5] for more details about the implicit differentiation.

Substituting these expressions into (9), we obtain the Taylor expansion having all the coefficients dependent rationally on  $\bar{y}$  and  $F^{-1}(\bar{y})$ . In this expression we can simply substitute  $F(\bar{x})$  for  $\bar{y}$  and  $\bar{x}$  for  $F^{-1}(\bar{y})$  and we obtain the desired simultaneous Taylor expansion depending on  $\bar{x}$ .

### 3.5 Example

Let us demonstrate the described general procedure on the following example. Consider a two-dimensional version of the POS, which can be used for example for the location of ships on the surface of sea. In this case  $F^{-1}$  can be expressed explicitly, but using square roots.

Suppose that we have three observation sites with coordinates  $O = [0, 0]$ ,  $A = [30, 0]$  and  $B = [-26, 15]$ . For to be coherent with the general notation introduced in the paragraph 3.1, we will denote the distance differences  $d_{AO}$  and  $d_{BO}$  by  $y_1$  and  $y_2$ . The mapping  $F$  is then given by:

$$\begin{aligned} y_1 &= \sqrt{x_1^2 - 60x_1 + 900 + x_2^2} - \sqrt{x_1^2 + y_1^2} \\ y_2 &= \sqrt{x_1^2 + 52x_1 + 901 + x_2^2} - 30x_2 - \sqrt{x_1^2 + x_2^2} \end{aligned} \quad (11)$$

Implicitising these formulae we get a system of algebraic equations  $G$ :

$$\begin{aligned} y_1^4 - 4y_1^2 x_1^2 - 4y_1^2 x_2^2 + 120y_1^2 x_1 - 1800y_1^2 + 3600x_1^2 - \\ - 108000x_1 + 810000 &= 0 \\ y_2^4 - 4y_2^2 x_1^2 - 4y_2^2 x_2^2 - 104y_2^2 x_1 + 60y_2^2 x_2 - 1802y_2^2 + \\ + 2704x_1^2 - 3120x_1 x_2 + 900x_2^2 + 93704x_1 - 54060x_2 + 811801 &= 0 \end{aligned} \quad (12)$$

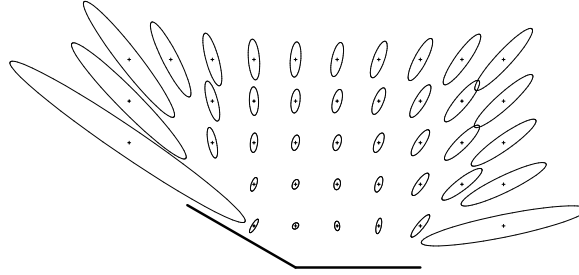
By implicit partial differentiation we were able, using the program Maple 8, to symbolically compute the partial derivatives of  $F^{-1}$  up to the degree 3. As the formulae become quickly very complicated, let us give just one example. The first component of the first partial derivative  $\frac{\partial F^{-1}}{\partial y_1}$  at the point  $F([x_1, x_2])$  is equal to

$$\begin{aligned} &\left( 780x_1 + 13515 - 15y_2^2 - 450x_2 + 2y_2^2 x_2 \right) y_1 \left( 2x_1^2 - 60x_1 + 900 - y_1^2 + 2x_2^2 \right) / \\ &(-780x_1^2 y_1^2 + 702000x_1^2 + 1633500x_1 - 1815x_1 y_1^2 - 182452500 + 202725y_1^2 + \\ &+ 15y_2^2 x_1 y_1^2 - 13500x_1 y_2^2 + 202500y_2^2 - 225y_2^2 y_1^2 - 902x_2 x_1 y_1^2 - 405000x_1 x_2 + \\ &+ 6075000x_2 - 30176x_2 y_1^2 + 1800y_2^2 x_2 x_1 - 27000y_2^2 x_2 + 56y_2^2 x_2 y_1^2 + 780x_2^2 y_1^2) \end{aligned} \quad (13)$$



Substituting (11) into this expression we get  $\frac{\partial F^{-1}}{\partial y_1}$  depending on the target point  $[x_1, x_2]$ . Doing the same for all partial derivatives up to the degree 3, we get a general Taylor expansion of the third order, in all points, depending on  $[x_1, x_2]$ . This general expression can be now used for the simultaneous description of the system of the confidence sets representing the error of  $x$ .

Let us suppose, for simplicity, that the time measurement error is the same at the three sites and is independent on the measured values. For a given probability the system of the confidence sets representing the error of  $[y_1, y_2]$  consists simply of the circles of the same radius. If we take a parametrisation of these circles and compose it with general Taylor expansion, we get the parametrisation of the system of confidence sets representing the error of  $[x_1, x_2]$ .



**Fig. 1.** System of the confidence sets representing the error of  $x$ , scaled by 100.

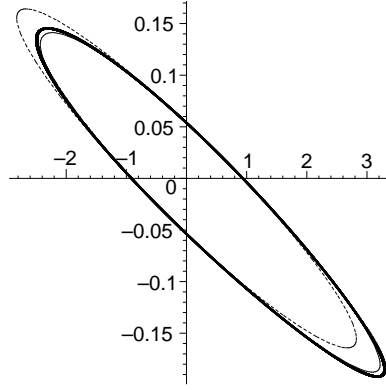
The figure 1 shows the position of the three sites  $O$  (central),  $A$  (right) and  $B$  (left) and the 3rd order approximation of several confidence sets (scaled by 100) corresponding to the error of  $x$  at the points marked by small crosses. These confidence sets correspond to the probability  $\alpha = 0.99$  in the case of a standard time measurement error.

The practical interpretation of this figure is as follows: If an object is situated at a point marked by a small crosses, then the POS will with the probability 99% detect its position within the corresponding set. The obtained general description of the system of the confidence sets is clearly very usefull for the visualisation and the analysis of the precision of the POS and of its range of operation.

The figure 2 shows more in detail the approximate parametrisation of the confidence set at the point  $x = [60, 2]$ , using the Taylor expansion of the 1st, the 2nd and the 3rd order. The first order approximation gives an ellipse. The third order approximation is indiscernible from the numerically computed confidence set.

## 4 Conclusion

The application of the described methods is not limited to the POS. The result presented in the paragraph 2.1 can be very useful in the construction of any



**Fig. 2.** Approximation of the confidence set at the point  $x = [60, 2]$ , using the Taylor expansion of order 1 (dotted line), 2 (thin solid line) and 3 (thick solid line). Note the different scaling of both axes.

devices using the quadric surfaces of revolution. This is for example the case of various observation systems based on the sum of distances, in which the ellipsoids occur.

The method described in the paragraph 3.4 can be applied in all situations satisfying the general setting 3.1. It is particularly interesting in the cases, in which we are interested by the analysis and the visualisation of the error depending not on the measured values  $y$ , but on the resulting values  $x$ . Let us mention for example the case of parallel robots, for which we want to know which positions can be reached with a prescribed precision.

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