

VLA SCIENTIFIC MEMORANDUM NO. 136

Robust Solution for Antenna Gains

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VLA visibility data bases frequently are contaminated by "wild" points, or, in statistical jargon, by outliers. The outliers may be due, for example, to transient problems in the receiver/i.f. chains, to sampler problems or correlator problems, or to faults in system error detection. Because that part of the standard calibration procedure which consists in solution for antenna gains is quite sensitive to discordant data points, the calibrator data must be edited carefully before calibration. I will outline a new method for the solution for antenna gains which is much less sensitive to outliers than is the standard method and which, moreover, in the presence of well-behaved random errors, performs nearly as well as the standard least-squares method. Therefore, much of the burden of data editing can be shifted to the post-calibration stage of data reduction. Because the usual map/CLEAN procedure is quite sensitive to wild points, editing is still required -- but it may be accomplished somewhat more easily once the data have been corrected for antenna gains. I shall pay no attention here to the problem of identifying outliers, but only to the task of describing a calibration scheme which is not too severely affected by them.

Both the standard calibration program (ANTSOL) and the self-calibration program - until its most recent implementation - have been based on the method of least-squares. A popular (at least in linear regression), but non-classical, approach to parameter estimation in the presence of outliers is, instead of solving for parameters which minimize the sum of

squared residuals, to solve for parameters which minimize the sum of absolute values of the residuals. The latter approach is known as  $\ell_1$  minimization, and least-squares as  $\ell_2$ , since in both cases one minimizes a discrete  $\ell_p$  norm of the residuals,  $p = 1$  or  $2$ . To gain an intuitive feeling for the difference, consider that in the case of linear regression, if the constant term, or intercept, is included in the regression model, then the  $\ell_2$  solution has the property that the mean residual is equal to zero, whereas an  $\ell_1$  solution (the  $\ell_1$  solution to an overdetermined linear system is not necessarily unique) has the property that the median residual is zero.  $\ell_1$  solutions to linear problems can be computed by standard linear programming methods. Nonlinear  $\ell_1$  minimization, as is required for our problem, has not received much attention in the literature, but effective algorithms for this purpose recently have been published [1,2].

The basic idea of the approach to nonlinear  $\ell_1$  minimization that is outlined in [1] is this:

Replace the problem

find a vector  $x$  which minimizes  $S(x) = \sum_i |f_i(x)|$

by a family of problems

find a vector  $x$  which minimizes  $S_\epsilon(x) = \sum_i [f_i^2(x) + \epsilon]^{\frac{1}{2}}$

and, for a sequence  $\epsilon_1 > \epsilon_2 > \dots$ , tending to 0, apply an extrapolation procedure to the corresponding sequence of solution vectors,  $x_1, x_2, \dots$ .

(The extrapolation procedure converts a convergent sequence to a sequence which converges no less rapidly -- in practice, for a variety of extrapolation methods in common use, the convergence usually is accelerated.)

Such an indirect method is required because, for well-behaved  $f_i$ ,  $S(x)$

may be nondifferentiable at points where some  $f_i$  is equal to 0. Extrapolation usually is required because the problems get more ill-conditioned as  $\epsilon$  tends to 0. The extrapolation method recommended in [1] is a standard one due to Fiacco and McCormick. In the present implementation of the method in the self-calibration program I haven't employed extrapolation because, in a variety of tests, the solutions have tended to stabilize for small  $\epsilon$  before numerical instability has set in. I do, however, solve a sequence of problems (in a typical case, say, with  $\epsilon = 5, .5, \text{ and } .05 \text{ mJy}$ ) since my minimization algorithms tend to require better initial guesses as  $\epsilon$  decreases.

For general purpose  $\ell_1$  minimization the method outlined in [2] ought to be a more reliable method than that in [1]. For our specialized problem, I believe that the above approach is adequate.

For the case of complex antenna gains, we wish, given instantaneous observations  $\tilde{v}_{ij}$  obtained on  $n(n-1)/2$  baselines  $i-j$  ( $i < j$ ), given corresponding model visibilities  $v_{ij}$ , and given weights  $w_{ij}$ , to solve for  $g = (g_i)_{i=1}^n$  by minimizing the summation

$$S_\epsilon(g) = \sum_{i < j} w_{ij} [|\tilde{v}_{ij} - g_i \bar{g}_j v_{ij}|^2 + \epsilon]^{\frac{1}{2}}.$$

A necessary condition for  $S_\epsilon$  to be minimized is that its gradient,  $\nabla S_\epsilon(g)$ , be equal to 0. We can rewrite the equation  $\nabla S_\epsilon(g) = 0$  so that the  $k^{\text{th}}$  antenna gain, in terms of the others, is given by

$$\begin{aligned} g_k &= \frac{\sum_{i < k} w_{ik} [|\mathbf{R}_{ik}|^2 + \epsilon]^{-\frac{1}{2}} g_i \tilde{v}_{ik} + \sum_{k < j} w_{kj} [|\mathbf{R}_{kj}|^2 + \epsilon]^{-\frac{1}{2}} g_j \bar{v}_{kj} \tilde{v}_{kj}}{\sum_{i < k} w_{ik} [|\mathbf{R}_{ik}|^2 + \epsilon]^{-\frac{1}{2}} |g_i \tilde{v}_{ik}|^2 + \sum_{k < j} w_{kj} [|\mathbf{R}_{kj}|^2 + \epsilon]^{-\frac{1}{2}} |g_j \bar{v}_{kj}|^2} \\ &\equiv f_k(g_1, g_2, \dots, g_{k-1}, g_{k+1}, \dots, g_n), \text{ where } \mathbf{R}_{ij} \equiv \tilde{v}_{ij} - g_i \bar{g}_j v_{ij}. \end{aligned}$$

Choosing an initial guess,  $g^{(0)}$ , we can use an iterative relaxation method where, at the  $(m+1)$ st iteration, the  $k^{\text{th}}$  antenna gain (for fixed  $\epsilon$ ) is approximated by

$$g_k^{(m+1)} = g_k^{(m)} + \omega [f_k(g_1^{(m+1)}, g_2^{(m+1)}, \dots, g_{k-1}^{(m+1)}, g_{k+1}^{(m)}, \dots, g_n^{(m)}) - g_k^{(m)}].$$

In practice, I have found successive over-relaxation (with  $\omega = 1.5$ ) to converge reliably and in about the same number of iterations as the corresponding method for  $\ell_2$  solutions.\* Typically, for each  $\epsilon$ , 5-10 iterations are required in order to achieve a relative error of  $10^{-4}$  in the solutions, starting, for  $\epsilon_1$ , with an initial guess  $g_1^{(0)}=1$ .

Phase errors generally have a more deleterious effect on maps than do amplitude errors, thus, especially in self-calibration, one is sometimes driven by signal-to-noise considerations to wish to reduce the number of parameters and so to solve only for antenna-based phase errors,  $\psi = (\psi_i)_{i=1}^n$ . We then have

$$S_\epsilon(\psi) = \sum_{i < j} w_{ij} [|\hat{v}_{ij} - e^{i(\psi_i - \psi_j)} v_{ij}|^2 + \epsilon]^{1/2} .$$

$S_\epsilon$  is periodic of period  $2\pi$  in each of the  $\psi_k$ , and we may use this fact to drive an iterative scheme to a local minimum of  $S_\epsilon$  rather than to a

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\*for  $\ell_2$  solutions,

$$g_k = \frac{\sum_{i < k} w_{ik} g_i \bar{v}_{ik} \hat{v}_{ik} + \sum_{k < j} w_{kj} g_j \bar{v}_{kj} \hat{v}_{kj}}{\sum_{i < k} w_{ik} |g_i v_{ik}|^2 + \sum_{k < j} w_{kj} |g_j v_{kj}|^2} .$$

local maximum:

$$\psi_k^{(m+1)} = \psi_k^{(m)} - \omega \tan^{-1} \left( \frac{\partial S_\epsilon / \partial \psi_k}{\partial^2 S_\epsilon / \partial \psi_k^2} \right).$$

$$\frac{\partial S_\epsilon}{\partial \psi_k} = - \sum_{i < k} w_{ik} [ |R_{ik}|^2 + \epsilon]^{-1/2} \text{Im}(L) + \sum_{k < j} w_{kj} [ |R_{kj}|^2 + \epsilon]^{-1/2} \text{Im}(M)$$

and

$$\begin{aligned} \frac{\partial^2 S_\epsilon}{\partial \psi_k^2} &= \sum_{i < k} w_{ik} \left\{ [ |R_{ik}|^2 + \epsilon]^{-1/2} \text{Re}(L) - [ |R_{ik}|^2 + \epsilon]^{-3/2} \text{Im}^2(L) \right\} \\ &\quad + \sum_{k < j} w_{kj} \left\{ [ |R_{kj}|^2 + \epsilon]^{-1/2} \text{Re}(M) - [ |R_{kj}|^2 + \epsilon]^{-3/2} \text{Im}^2(M) \right\}, \end{aligned}$$

where  $L \equiv e^{i(\psi_i - \psi_k)}$ ,  $M \equiv e^{i(\psi_k - \psi_j)}$ , and  $R_{ij} \equiv \tilde{V}_{ij} - e^{i(\psi_i - \psi_j)} V_{ij}$ . \*  
 [See erratum sheet preceding  
 this memo.]

In the circumstance in which the real parts of the visibility observations have an error distribution different from that of the imaginary parts, and perhaps in the circumstance in which the noise in the real part of an observation is independent of the noise in the imaginary part of that observation, it might be preferable to minimize the sum of the

\*for  $\ell_2$  solutions,

$$2 \frac{\partial S}{\partial \psi_k} = - \sum_{i < k} w_{ik} \text{Im}(L) + \sum_{k < j} w_{kj} \text{Im}(M) \quad \text{and}$$

$$2 \frac{\partial^2 S}{\partial \psi_k^2} = \sum_{i < k} w_{ik} \text{Re}(L) + \sum_{k < j} w_{kj} \text{Re}(M).$$

absolute values of the real parts of the residuals plus the sum of the absolute values of the imaginary parts of the residuals, rather than the sum of the moduli of the complex residuals. However, I have tested this approach for the case of well-behaved random errors, with identical error distributions for real and imaginary, but independent errors in real part and imaginary part, and I found, in that case, very little difference in the computed solutions. In my implementation, this alternative solution method is much slower because more arithmetic is required. For brevity, the algebraic machinery for the alternative method has been omitted from this report.

The results of a series of tests comparing the  $\ell_1$  gain solution method with the  $\ell_2$  method are presented below. The data used in each test run simulate the observations of a one Jansky point source (located at the visibility phase reference position) by a 27 element interferometer. The visibility model exactly represents the source; i.e.,  $v_{ij} \equiv 1$ . The observations  $\hat{v}_{ij}$  ( $1 \leq i < j \leq 27$ ) for each run are these model visibilities, modified by 27 multiplicative complex gains  $g = (g_i)_{i=1}^{27}$ , possibly with random noise added in. Weights  $w_{ij} \equiv 1$  were used in all cases. In order to simulate wild observations, in some cases a number of the data points have been chosen at random and have no relation to the visibility model or to  $g$ .

Any solution to one of the minimization problems described above is unique only up to a multiplicative constant of unit modulus; that is, if  $\hat{g}$  is one local minimizer, then  $c\hat{g}$  is another, whenever  $|c|=1$ . The solution algorithms are such that, with no special attention to the fact that the problems are underdetermined, they always converge upon some solution<sup>\*</sup> (barring unusual circumstances). In order most fairly to compare a com-

\* Interestingly, if the argument of some gain is held fixed, then the rate of convergence is diminished.

puted solution,  $\hat{g}$ , with the true solution,  $g$ , each computed solution has been multiplied by that factor,  $c$ , of unit modulus, appropriate to minimize  $\sum |g_i - c\hat{g}_i|^2$ . In terms of the true solution,  $c$  is given by  $\sum g_i \bar{\hat{g}}_i / |\sum g_i \bar{\hat{g}}_i|$ .

Behavior when the data are contaminated only by well-behaved random errors

The  $\ell_1$  method clearly would be unattractive if, given data contaminated only by well-behaved random errors, it were to behave much worse than the least-squares method. Two series of tests were performed in which the model observations were contaminated by adding varying levels of zero mean (approximately-) Gaussian distributed random noise (independently) to the real parts and to the imaginary parts of the visibility observations; i.e.,  $\tilde{v}_{ij} = g_i \bar{g}_j v_{ij} + \text{noise} (= g_i \bar{g}_j + \text{noise}, \text{for our 1 Jy point model})$ . To test the two phase solution methods, the antenna gains were chosen to be of unit modulus, with phases drawn at random from the uniform distribution on  $[0^\circ, 360^\circ]$ . To test the complex gain solution methods, the phases of the antenna gains were chosen in the same manner as before, and the moduli were drawn from the uniform distribution on  $[.5, 1.5]$ . Ten trials were performed for each of eleven choices of standard deviation of the random noise. A different choice of gains and a different choice of noise were made for each trial (except that the same data were used for both the  $\ell_1$  and  $\ell_2$  solution tests; hence there will be some correlation in the mean solution errors, between  $\ell_1$  and  $\ell_2$ ). The rms error in each gain solution was computed as  $\sqrt{n^{-1} \sum |g_i - \hat{g}_i|^2}$ . The mean of these errors, for each set of ten trials, is shown in Table 1 and in the plot of Figure 1. For S/N  $\gtrsim 3$  (resp.,  $\gtrsim 2$ ) the rms gain solution error remains  $\approx .1$  for the phase solutions (resp., for the gain

solutions). Excluding the points at 2.0 Jy random noise, the  $\ell_1$  phase solution error (resp., gain solution error) is typically about 13% (resp., 14%) greater than the  $\ell_2$  error. The results of two representative trials (which were not included in the test sample) are shown in Figure 2.

Another test which might be of interest is the case  $\tilde{v}_{ij} = g_i \bar{g}_j (v_{ij} + \text{noise})$ . Trials in which the observations were contaminated by noise of this form gave results which were very similar to those described above.

Sensitivity of the solution methods to observations chosen at random to be outliers

To test the sensitivity of the solution methods to randomly chosen outliers, some probability was assigned according to which an observation was chosen to be made a wild point. The wild points were assigned phases drawn at random from the uniform distribution on  $[0^\circ, 360^\circ]$  and amplitudes drawn at random from the uniform distribution on  $[0, 2]$  Jy. The "tame" observations were contaminated by 0.2 Jy Gaussian noise, as in the random noise tests, above. The results for complex gain solutions and for phase solutions are shown in detail in Figures 3 and 4, respectively. For data made wild with probabilities .05, .10, .20, and .50, the mean observed rms gain solution errors in ten trials were 6.3, 6.6, 8.1, and 23 ( $\times 10^{-2}$ ) (9.7, 12, 19, and 40 ( $\times 10^{-2}$ )) for the  $\ell_1$  (resp., for the  $\ell_2$ ) solution method. In the corresponding test of the phase solution methods, assigning the same probabilities, the mean observed rms errors were 4.1, 4.8, 4.9, and 7.7 ( $\times 10^{-2}$ ) (5.3, 7.3, 10.0, and 27 ( $\times 10^{-2}$ )) for the  $\ell_1$  (resp., for the  $\ell_2$ ) solution method.

To test the sensitivity to the presence of extreme outliers, two

trials were run in which each observation, with probability .10, was assigned a random phase and a random amplitude in the range  $[0, 10^{10}]$  Jy. The results are shown in Figures 5a and 5b. The rms  $\ell_1$  (resp.,  $\ell_2$ ) gain solution errors observed were 7.2 and  $4.5 \times 10^{-2}$  (resp.,  $4.2 \times 10^7$  and  $1.3 \times 10^0$ ) for the complex gain solution and for the phase solution, respectively. This test, with ten trials in each case, was repeated for the case of wild amplitudes in the range  $[0, .01]$  Jy. The results are shown in Figures 5c and 5d. The mean observed rms gain solution errors for  $\ell_1$  (resp., for  $\ell_2$ ) were 7.2 and  $4.6 \times 10^{-2}$  ( $10.9$  and  $4.1 \times 10^{-2}$ ) in the complex gain solutions and in the phase solutions, respectively. In the presence of extreme outliers, the  $\ell_1$  method appears in all cases, except that of phase solutions with very small amplitude outliers, to outperform the  $\ell_2$  method.

Sensitivity of the solution methods to bad i.f.'s

An occasional problem with VLA data is that the data associated with a single i.f. may be bad; sometimes the system error detection facility fails to recognize the problem. In these tests, 0.2 Jy Gaussian distributed random noise was added to the observations, as above, except that all the visibilities associated with antenna 5 were assigned a given amplitude and a random phase uniformly distributed on  $[0^\circ, 360^\circ]$ . Representative results, along with the rms gain solution errors that were observed in ten trials with identical error distributions, are displayed in Figures 6 and 7. For amplitudes associated with antenna 5 of 0.1, 1.0, 2.0, and 5.0 Jy, mean rms gain solution errors (excluding antenna 5) of 5.8, 5.9, 6.6, and  $8.2 \times 10^{-2}$  ( $5.2, 5.3, 6.6$ , and  $80 \times 10^{-2}$ ) were observed for the  $\ell_1$  (resp., the  $\ell_2$ ) solutions when solving for full complex gains. In

ten trials with 26 antennas and 0.2 Jy random noise, with no outliers, the mean observed rms gain solution error was  $5.9 \pm .5$  and  $5.3 \pm .4$  ( $\times 10^{-2}$ ) for the  $\ell_1$  and  $\ell_2$  solutions, respectively.

When solving only for phase errors, for visibility amplitudes associated with antenna 5 of 0.5, 1.0, 2.0, and 100.0 Jy, mean rms gain solution errors (excluding antenna 5) of 5.1, 4.6, 5.3, and 5.4 ( $\times 10^{-2}$ ) (4.8, 5.2, 7.3, and 126 ( $\times 10^{-2}$ )) were observed with the  $\ell_1$  (resp., the  $\ell_2$ ) solution method. In ten trials with 26 antennas, 0.2 Jy random noise, and no outliers, the mean observed rms gain solution error was  $4.6 \pm .6$  and  $4.1 \pm .5$  ( $\times 10^{-2}$ ) for the  $\ell_1$  and  $\ell_2$  solutions, respectively.

Clearly it is desirable to detect the presence of a bad i.f. before solving for antenna gains, no matter which solution method is used. In the case of full complex gain solutions, if there is a bad i.f., and if the associated amplitudes are very large, then the  $\ell_1$  method is preferable. For phase solutions, the  $\ell_1$  method is extremely insensitive to a bad i.f. In both cases, the  $\ell_1$  method is strikingly superior to the least-squares method only when the errors associated with the bad i.f. are very large.

A few additional tests were run: If the observations associated with antenna 5 are assigned not a random phase, but each is assigned, instead, a phase of 45 degrees, then the conclusions above remain unaltered. In the case of not a single bad i.f., but of five bad i.f.'s, it is still preferable to detect the presence of the bad i.f.'s before solving for the gains (the solutions are quite inferior to those for 22 good i.f.'s), and the  $\ell_2$  solutions still are not much inferior to the  $\ell_1$  solutions except when the amplitudes of the bad points are very large.

### Conclusions

The  $\ell_1$  method for gain solutions appears to be an attractive alternative to the least-squares method, both in self-calibration and, perhaps more so, in standard calibration. In the case of standard calibration in the pipeline data reduction system, the  $\ell_1$  method has appeal because in the pipeline it might not be possible to correct data editing errors before calibration. The appeal in self-calibration lies in the fact that, in combination with a scheme for the identification of outliers, one would be able first to solve for the gains and then to identify and flag the outliers, resting assured that the gain solutions probably were not unduly influenced by the outliers (otherwise it would be necessary to recompute the gains after flagging data). Since the  $\ell_1$  method performs nearly as well as  $\ell_2$  in the presence of well-behaved errors, any decision not to use  $\ell_1$  in preference to  $\ell_2$  probably would be based either upon faith that there were not many harmful outliers or upon considerations of computational speed. In a careful implementation, especially if the gains are computed in an array processor, the penalty due to the added computational complexity ought not to be too great.

Cornwell's recent error analysis of calibration [3] is a useful complement to the test results reported here. I plan to do a related empirical study of the sensitivity of simultaneous solution for antenna gains and for constant or slowly time-varying additive or multiplicative correlator based errors, to errors in the observations and to perturbations in the source model.

The problem of identifying outliers merits careful attention. [4] ought to be a useful reference on this subject.

TABLE 1.

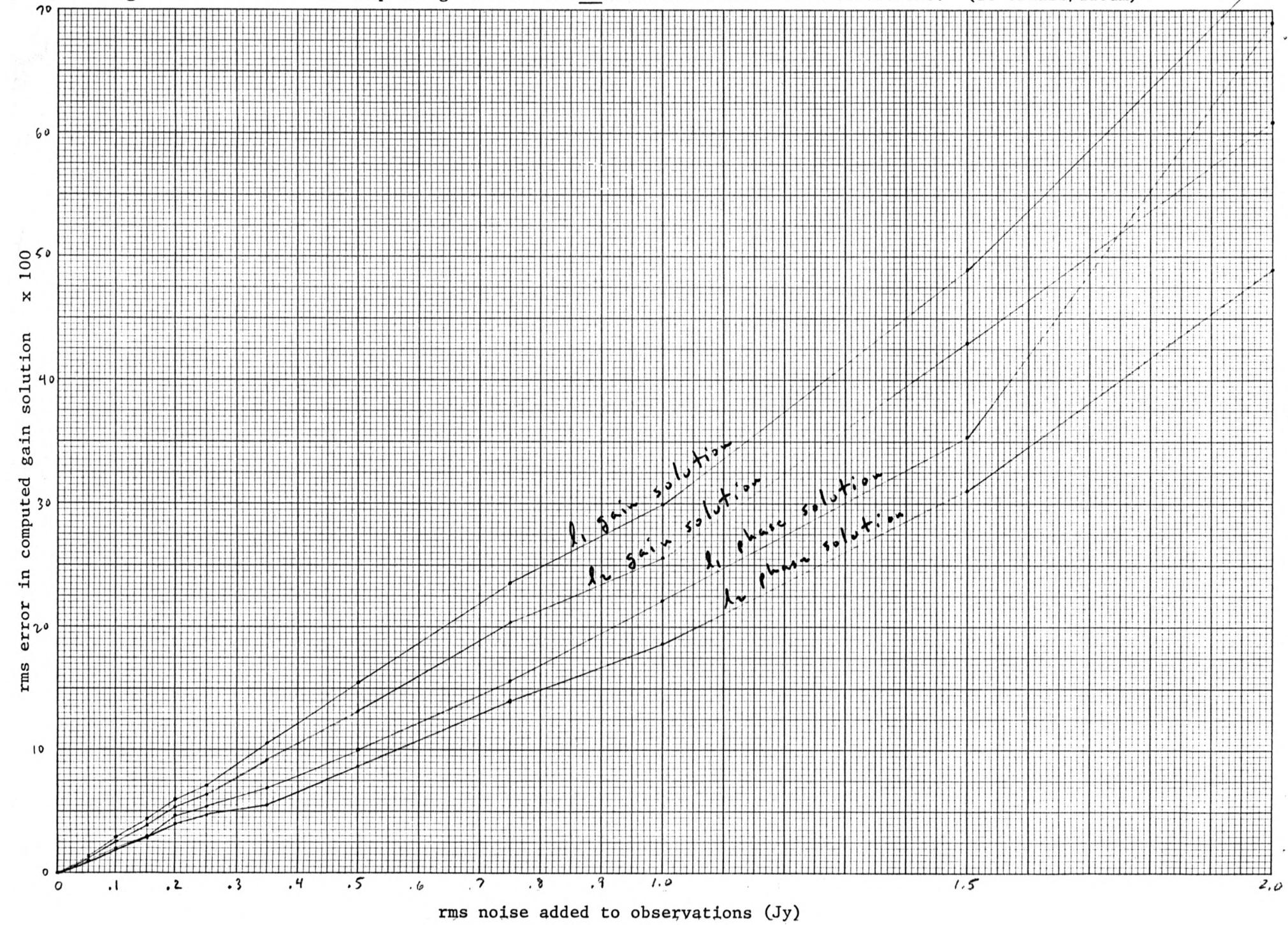
rms error in computed gain solution x100

rms noise (Jy)	phase solutions		complex gain solutions	
	$\ell_1$	$\ell_2$	$\ell_1$	$\ell_2$
.05	1.0 ± .1	1.0 ± .1	1.4 ± .2	1.3 ± .2
.10	2.0 ± .3	1.9 ± .3	2.9 ± .4	2.6 ± .3
.15	3.0 ± .3	2.9 ± .3	4.4 ± .5	3.8 ± .4
.20	4.7 ± .6	4.0 ± .5	6.0 ± .6	5.3 ± .5
.25	5.4 ± .6	4.7 ± .7	7.1 ± .8	6.4 ± .7
.35	6.9 ± .8	6.0 ± .4	11 ± 1	9.2 ± 1
.50	10 ± 1	8.7 ± 1	16 ± 2	13 ± 1
.75	16 ± 3	14 ± 2	24 ± 3	20 ± 2
1.00	22 ± 3	19 ± 3	30 ± 4	26 ± 4
1.50	35 ± 6	31 ± 4	49 ± 7	43 ± 5
2.00	69 ± 23	49 ± 12	74 ± 17	61 ± 10

10 trials/datum

simulated observations of 1 Jy point source

Figure 1. Rms error in computed gain solution vs. rms noise added to observations. (10 trials/datum)



REFERENCES

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2. Murray, W. and M. L. Overton, "A projected Lagrangian algorithm for nonlinear  $\ell_1$  approximation", SIAM J. Sci. Stat. Comp., 2 (1981), 207-224.
3. Cornwell, T., "An error analysis of calibration", VLA Scientific Memo. No. 135, September 1981.
4. Barnett, V., and T. Lewis, Outliers in Statistical Data, Wiley, Chichester, 1978.

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TRUE SOLUTION			L1 SOLUTION		L2 SOLUTION	
ANT	AMP	PHASE	AMP	PHASE	AMP	PHASE
1	0.74019	18.96	0.68422	16.14	0.69212	16.55
2	0.76571	-142.61	0.76475	-147.41	0.76112	-146.78
3	1.46513	135.12	1.43573	134.16	1.42796	134.83
4	0.69369	139.70	0.70972	140.11	0.78609	137.89
5	1.20463	-112.25	1.19658	-112.99	1.17391	-113.42
6	0.99270	114.12	0.98797	114.17	0.99179	112.62
7	1.25279	73.59	1.29354	71.34	1.29637	72.74
8	1.10175	-132.58	1.14348	-134.57	1.12624	-133.99
9	0.63656	-166.74	0.63576	-161.29	0.68157	-161.84
10	1.30140	64.92	1.29727	65.87	1.29738	66.12
11	0.59614	86.63	0.56424	84.91	0.57076	85.48
12	1.31672	161.50	1.34419	160.76	1.35684	159.88
13	1.21291	33.59	1.14781	33.99	1.13274	33.16
14	1.21615	115.15	1.26326	116.42	1.27399	117.71
15	0.66908	-77.36	0.73036	-79.32	0.69313	-78.14
16	0.71566	40.06	0.70078	42.26	0.68445	43.15
17	0.87360	155.69	0.85458	155.59	0.85356	156.47
18	0.74752	66.06	0.82046	64.84	0.78710	65.27
19	0.73871	41.38	0.72071	44.42	0.73614	43.78
20	0.90312	61.31	0.92573	58.27	0.94986	56.03
21	0.74135	111.30	0.73756	113.17	0.74264	111.79
22	1.10917	-134.93	1.02993	-137.61	1.03807	-137.41
23	0.55408	-21.38	0.45843	-25.68	0.48913	-24.61
24	0.64669	-132.17	0.53754	-129.21	0.62701	-129.51
25	1.22916	160.26	1.23900	167.84	1.23783	166.14
26	1.37211	-116.55	1.40111	-115.99	1.40668	-116.18
27	1.17198	8.24	1.20896	7.63	1.21978	7.94

RMS SOLUTION ERROR:

6.45E-02

5.76E-02

complex gain solution  
0.2 Jy random noise

BADIF=F  
NOISEAMP= 0.20 JY  
WILDPCT= 0.08  
WILDAMP= 0.000E+00 JY

TRUE SOLUTION			L1 SOLUTION		L2 SOLUTION	
ANT	AMP	PHASE	AMP	PHASE	AMP	PHASE
1	1.00000	21.48	1.00000	19.56	1.00000	20.72
2	1.00000	-150.77	1.00000	-154.83	1.00000	-150.31
3	1.00000	119.24	1.00000	119.42	1.00000	118.35
4	1.00000	86.25	1.00000	91.98	1.00000	87.84
5	1.00000	129.27	1.00000	136.87	1.00000	135.93
6	1.00000	156.29	1.00000	161.88	1.00000	159.96
7	1.00000	-58.31	1.00000	-64.59	1.00000	-65.39
8	1.00000	121.55	1.00000	113.52	1.00000	117.62
9	1.00000	97.42	1.00000	102.99	1.00000	98.68
10	1.00000	157.98	1.00000	165.80	1.00000	168.37
11	1.00000	104.85	1.00000	109.47	1.00000	103.14
12	1.00000	-93.02	1.00000	-91.38	1.00000	-93.28
13	1.00000	81.85	1.00000	75.48	1.00000	76.88
14	1.00000	-78.73	1.00000	-77.62	1.00000	-77.98
15	1.00000	-154.98	1.00000	-150.37	1.00000	-151.88
16	1.00000	-11.08	1.00000	-16.54	1.00000	-15.28
17	1.00000	-81.36	1.00000	-81.81	1.00000	-76.75
18	1.00000	-149.89	1.00000	-152.95	1.00000	-151.72
19	1.00000	-156.57	1.00000	-153.85	1.00000	-155.18
20	1.00000	32.21	1.00000	32.30	1.00000	29.34
21	1.00000	-128.13	1.00000	-125.81	1.00000	-124.38
22	1.00000	98.61	1.00000	96.11	1.00000	95.71
23	1.00000	-134.91	1.00000	-131.66	1.00000	-129.81
24	1.00000	-28.66	1.00000	-27.86	1.00000	-25.22
25	1.00000	56.97	1.00000	58.29	1.00000	56.89
26	1.00000	71.38	1.00000	82.98	1.00000	77.83
27	1.00000	66.68	1.00000	64.44	1.00000	65.79

RMS SOLUTION ERROR:

8.35E-02

6.03E-02

phase solution  
0.35 Jy random noise

BADIF=F  
NOISEAMP= 0.35 JY  
WILDPCT= 0.08  
WILDAMP= 0.000E+00 JY

Figure 2. Two representative trials illustrating behavior of the solution methods in the presence of random noise.

a)

TRUE SOLUTION					
	ANT	AMP	PHASE	AMP	PHASE
	1	1.89987	175.55	1.10575	176.47
	2	1.29383	-159.41	1.27422	-154.79
	3	0.84909	-180.12	0.51694	-164.38
	4	0.70445	-41.25	#.59571	-38.39
	5	1.30524	-30.68	1.38917	-38.73
	6	1.16762	-29.93	1.16249	-38.43
	7	0.86445	-127.58	0.87108	-127.68
	8	0.52583	-172.03	0.94576	-171.35
	9	1.10568	-130.3	1.39208	-130.97
	10	1.05573	-152.29	1.51442	-162.21
	11	0.73202	72.65	#.74676	71.39
	12	1.11833	17.02	1.09703	16.85
	13	0.79723	128.42	#.72391	128.06
	14	1.14247	-115.59	1.17068	-114.35
	15	1.09569	-80.94	1.04371	-81.24
	16	1.27259	-50.11	1.18478	-56.87
	17	0.90325	-55.64	#.93718	-56.92
	18	1.40229	-78.05	1.47415	-72.69
	19	1.42598	-117.88	1.39516	-116.83
	20	1.15649	-164.13	1.08549	-163.81
	21	1.00573	70.37	1.42876	75.54
	22	0.65594	-78.74	#.69129	-75.02
	23	1.07673	89.69	1.00044	89.27
	24	0.84003	3.49	#.68582	3.11
	25	1.19393	135.01	1.25216	134.58
	26	1.62587	118.11	1.00143	117.55
	27	0.71339	63.24	#.76277	64.83

RMS SOLUTION ERROR: 4.87E-82      Error (x100), 10 trials: 7.98E-82

P = .05

BADIF=5  
NOISEAMP= 8.29 JY  
WILDCT= 4.27  
WILDAMP= 2.000E+00 JY

4.9    7.9  
6.0    7.3  
7.2    11.5  
6.9    12.3  
7.1    3.3  
6.1    2.5  
6.1    9.0  
5.3    11.2  
6.5    11.1  
6.4    6.6  

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6.3 ± .3    9.7 ± 1.5

b)

TRUE SOLUTION					
	ANT	AMP	PHASE	AMP	PHASE
	1	1.83603	-167.97	#.98723	-166.39
	2	1.87474	129.58	#.96317	127.37
	3	1.23897	146.39	1.12397	148.53
	4	1.18522	-171.53	1.15354	-173.71
	5	0.95301	75.06	#.95236	0.99
	6	1.31045	126.89	1.27214	126.89
	7	0.74670	77.82	#.70197	80.23
	8	1.26754	51.68	1.32887	53.83
	9	1.11677	49.27	1.09359	49.50
	10	1.05765	-75.50	#.61631	-74.69
	11	1.13125	-170.97	1.12292	-175.42
	12	0.93848	173.66	#.01178	178.18
	13	0.65374	132.69	#.60178	132.61
	14	0.74572	-39.58	#.75758	-44.47
	15	0.96338	116.11	1.00164	121.65
	16	0.80555	132.39	#.62377	129.34
	17	1.24466	20.98	1.12271	17.91
	18	1.09485	177.85	#.99267	-179.18
	19	1.15035	-114.04	1.17818	-115.62
	20	1.48091	-15.10	1.46719	-14.88
	21	#.51893	-26.13	#.49165	-34.64
	22	1.04209	-128.21	1.45165	-132.58
	23	0.81563	173.64	0.79644	174.47
	24	#.84111	145.62	#.04492	149.38
	25	0.95564	36.44	1.02493	34.33
	26	1.30801	-51.49	1.29051	-54.53
	27	#.62514	-123.87	#.66384	-117.88

RMS SOLUTION ERROR: 7.88E-82      Error (x100), 10 trials: 1.11E-81

P = .10

BADIF=5  
NOISEAMP= 8.28 JY  
WILDCT= 7.69  
WILDAMP= 2.000E+00 JY

7.1    11.1  
5.7    12.2  
5.7    13.0  
7.6    15.0  
6.0    10.3  
6.6    14.3  
2.5    11.7  
6.0    11.0  
6.6    12.1  

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6.6 ± .6    12.1 ± 1.4

c)

TRUE SOLUTION					
	ANT	AMP	PHASE	AMP	PHASE
	1	0.95525	35.54	#.95548	33.88
	2	1.10518	-164.96	1.23329	-164.28
	3	0.51311	-69.98	#.94556	-78.78
	4	0.87468	-141.64	#.77247	-138.59
	5	1.15979	-142.17	1.15837	-143.78
	6	1.13292	12.43	1.03599	12.43
	7	0.77551	172.47	#.64348	171.51
	8	0.89095	85.87	#.59596	85.44
	9	1.22228	-23.64	1.23271	-25.64
	10	0.91648	-38.53	#.92929	-39.98
	11	1.11762	5.28	1.12212	4.91
	12	1.15024	-178.61	1.18251	-174.54
	13	1.33274	113.77	1.24598	-112.64
	14	1.16511	165.71	1.21163	126.71
	15	0.77755	87.49	#.71653	75.66
	16	1.45331	79.65	1.48563	88.89
	17	1.31629	78.46	1.19946	71.54
	18	1.36141	-137.95	1.34840	-130.37
	19	1.19244	-86.47	1.04827	-86.52
	20	0.69509	-57.94	0.57807	-55.56
	21	0.96324	-33.98	0.74537	-33.63
	22	0.96328	141.18	#.97319	143.51
	23	1.29798	-132.72	1.38671	-131.66
	24	0.55052	-49.38	#.59908	-41.41
	25	1.29528	-43.37	1.24077	-44.79
	26	1.37174	-185.18	1.38121	-184.11
	27	0.59817	95.18	#.68469	94.96

RMS SOLUTION ERROR: 9.88E-82      Error (x100), 10 trials: 2.10E-81

P = .20

BADIF=5  
NOISEAMP= 8.28 JY  
WILDCT= 22.65  
WILDAMP= 2.000E+00 JY

9.0    21.8  
8.1    16.4  
2.6    19.7  
7.7    21.0  
2.7    21.9  
2.0    16.7  
7.8    21.5  
2.0    15.7  
1.1    17.2  

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8.1 ± .8    19.1 ± 2.7

d)

TRUE SOLUTION					
	ANT	AMP	PHASE	AMP	PHASE
	1	1.45545	-15.37	1.16775	-21.31
	2	0.65221	93.25	#.76659	92.51
	3	1.10179	-29.48	1.17332	-27.83
	4	1.19879	21.59	#.50927	30.31
	5	0.75231	-36.82	#.38514	-39.89
	6	1.00271	-64.64	#.07724	-67.23
	7	1.17020	-161.66	1.12589	-161.53
	8	1.41176	-18.85	1.27163	-15.38
	9	0.71407	84.14	#.68284	92.45
	10	0.55778	-34.67	#.30543	-29.32
	11	1.15709	51.78	1.19098	52.25
	12	1.47797	176.78	1.26771	179.24
	13	1.07110	-17.79	#.97805	-111.12
	14	0.97334	-16.01	#.87909	-156.35
	15	1.20262	-81.10	1.14625	-84.18
	16	1.30072	95.95	1.14981	97.44
	17	0.95544	148.73	#.90668	143.26
	18	1.11079	-110.79	1.07004	-113.96
	19	1.23092	-55.90	1.21799	-55.95
	20	0.55951	93.14	#.08980	95.83
	21	0.39571	21.55	0.36921	20.16
	22	#.75306	107.63	#.60746	110.68
	23	0.92375	-171.68	0.74162	-166.98
	24	0.03562	-2.48	#.57799	-15.98
	25	0.89958	58.38	0.91446	60.88
	26	1.25293	38.95	1.15981	29.39
	27	0.55964	137.15	#.62153	146.77

RMS SOLUTION ERROR: 2.30E-81      Error (x100), 10 trials: 4.84E-81

P = .50

BADIF=5  
NOISEAMP= 8.28 JY  
WILDCT=49.06  
WILDAMP= 2.000E+00 JY

23.5    40.4  
19.1    37.4  
19.7    59.1  
24.8    39.0  
14.5    24.2  
25.2    41.3  
25.5    47.3  
26.1    42.6  
24.7    46.4  
18.5    38.0  

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21.6 ± 4.4    40.0 ± 4.8

Figure 3. Sensitivity of complex gain solutions to outliers. The "tame" observations had random noise of 0.2 Jy added. Each observation, with a given probability, P, was made a wild point by assigning it a random phase and an amplitude in the range 0-2 Jy.

I  
L

TRUE SOLUTION			L1 SOLUTION		L2 SOLUTION		TRUE SOLUTION			L1 SOLUTION		L2 SOLUTION	
ANT	AMP	PHASE	AMP	PHASE	AMP	PHASE	ANT	AMP	PHASE	AMP	PHASE	AMP	PHASE
1	1.00000	69.31	1.00000	78.29	1.00000	66.15	1	1.00000	155.79	1.00000	158.03	1.00000	157.87
2	1.00000	120.88	1.00000	122.57	1.00000	123.31	2	1.00000	53.73	1.00000	57.19	1.00000	63.19
3	1.00000	-34.63	1.00000	-33.83	1.00000	-33.44	3	1.00000	-136.79	1.00000	-134.91	1.00000	-137.61
4	1.00000	-37.59	1.00000	-37.78	1.00000	-30.76	4	1.00000	-111.17	1.00000	-110.18	1.00000	-126.39
5	1.00000	89.77	1.00000	88.21	1.00000	85.15	5	1.00000	-108.74	1.00000	-108.37	1.00000	-103.49
6	1.00000	-83.87	1.00000	-83.98	1.00000	-79.68	6	1.00000	51.77	1.00000	52.76	1.00000	52.55
7	1.00000	46.27	1.00000	48.72	1.00000	49.84	7	1.00000	102.88	1.00000	102.78	1.00000	102.24
8	1.00000	-104.97	1.00000	-103.89	1.00000	-100.24	8	1.00000	-12.29	1.00000	-13.53	1.00000	-9.58
9	1.00000	-51.13	1.00000	-53.49	1.00000	-56.27	9	1.00000	-65.26	1.00000	-64.53	1.00000	-63.81
10	1.00000	-153.46	1.00000	-163.66	1.00000	-162.93	10	1.00000	30.17	1.00000	30.54	1.00000	30.37
11	1.00000	127.66	1.00000	128.08	1.00000	127.98	11	1.00000	163.79	1.00000	163.84	1.00000	163.83
12	1.00000	-36.63	1.00000	-38.23	1.00000	-38.39	12	1.00000	37.82	1.00000	35.65	1.00000	40.23
13	1.00000	-119.26	1.00000	-120.21	1.00000	-120.84	13	1.00000	-169.98	1.00000	-168.42	1.00000	-178.26
14	1.00000	-70.06	1.00000	-73.29	1.00000	-73.67	14	1.00000	8.38	1.00000	7.37	1.00000	3.37
15	1.00000	161.59	1.00000	163.90	1.00000	165.31	15	1.00000	-81.29	1.00000	-78.19	1.00000	-77.92
16	1.00000	-56.73	1.00000	-54.62	1.00000	-59.08	16	1.00000	-111.51	1.00000	-108.50	1.00000	-115.46
17	1.00000	-62.78	1.00000	-63.34	1.00000	-63.08	17	1.00000	128.89	1.00000	127.76	1.00000	128.99
18	1.00000	143.22	1.00000	145.71	1.00000	147.23	18	1.00000	-114.53	1.00000	-116.15	1.00000	-117.49
19	1.00000	61.93	1.00000	63.63	1.00000	64.15	19	1.00000	49.26	1.00000	53.37	1.00000	52.74
20	1.00000	132.99	1.00000	137.50	1.00000	136.55	20	1.00000	48.93	1.00000	47.89	1.00000	40.43
21	1.00000	118.13	1.00000	113.81	1.00000	113.16	21	1.00000	-38.75	1.00000	-35.89	1.00000	-35.37
22	1.00000	-35.26	1.00000	-38.81	1.00000	-37.03	22	1.00000	-10.54	1.00000	-9.32	1.00000	-2.95
23	1.00000	-69.17	1.00000	-68.32	1.00000	-67.70	23	1.00000	125.98	1.00000	123.12	1.00000	123.51
24	1.00000	77.18	1.00000	88.03	1.00000	79.74	24	1.00000	65.74	1.00000	59.17	1.00000	46.44
25	1.00000	116.68	1.00000	119.12	1.00000	119.78	25	1.00000	107.22	1.00000	109.17	1.00000	108.68
26	1.00000	-7.98	1.00000	-11.58	1.00000	-11.72	26	1.00000	-65.95	1.00000	-61.87	1.00000	-62.25
27	1.00000	38.29	1.00000	31.44	1.00000	31.46	27	1.00000	14.81	1.00000	15.85	1.00000	16.33
RMS SOLUTION ERROR:			3.92E-82		5.51E-82		RMS SOLUTION ERROR:			4.91E-82		9.83E-82	
BADIF=F			NOISEAMP= 8.28 JY		WILDPCT= 4.27		BADIF=F			NOISEAMP= 8.28 JY		WILDPCT= 9.57	
WILDAAMP= 2.00E+08 JY													
$P = .05$			Error ( $\times 10^6$ ), 10 trials:		1.2		$P = .10$			Error ( $\times 10^6$ ), 10 trials:		1.2	
BADIF=F			NOISEAMP= 8.28 JY		WILDPCT= 4.27		BADIF=F			NOISEAMP= 8.28 JY		WILDPCT= 9.57	
WILDAAMP= 2.00E+08 JY													
$P = .41 \pm .42$			5.3 ± 1.2		4.1 ± 1.2		$P = .50$			4.8 ± 1.7		7.3 ± 7.4	

Figure 4. Sensitivity of phase solutions to outliers. The "tame" observations had random noise of 0.2 Jy added. Each observation, with probability P, was made a wild point by assigning it a random phase and an amplitude in the range 0-2 Jy.

TRUE SOLUTION			L1 SOLUTION			L2 SOLUTION			TRUE SOLUTION			L1 SOLUTION			L2 SOLUTION		
ANT	AMP	PHASE	AMP	PHASE	AMP	PHASE	AMP	PHASE	ANT	AMP	PHASE	AMP	PHASE	AMP	PHASE		
1	1.06404	-176.62	1.09113	-177.29	2.17852E+00	-176.62	1.08000	155.36	1.	1.00000	151.24	1.00000	93.95				
2	1.42777	-16.99	1.46943	-17.87	5.67394E-09	-19.00	1.00000	129.72	1.	1.00000	133.79	1.00000	-102.88				
3	0.56165	-8.97	0.59236	-3.89	4.31308E-09	-7.41	1.	1.00000	7.28	1.	1.00000	4.64	1.00000	149.98			
4	1.11222	-22.82	1.17535	-18.44	5.14358E-09	-19.52	4.	1.00000	93.67	1.	1.00000	94.87	1.00000	9.31			
5	0.57124	-120.79	0.57232	-129.55	4.20181E-09	-122.27	5.	1.00000	-40.85	1.	1.00000	-42.44	1.00000	-161.58			
6	0.66483	92.62	0.67166	94.46	5.76939E-09	-4.55	6.	1.00000	169.74	1.	1.00000	161.77	1.00000	107.74			
7	1.15405	-71.38	1.17841	-76.62	5.78044E-09	-04.55	7.	1.00000	172.12	1.	1.00000	169.99	1.00000	-66.89			
8	1.02302	-120.49	1.09079	-19.49	5.94423E-09	-109.42	8.	1.00000	-55.23	1.	1.00000	-56.23	1.00000	-76.87			
9	0.60387	-4.34	0.80398	-38.09	2.37950E-09	-35.50	9.	1.00000	-72.43	1.	1.00000	-71.81	1.00000	-129.07			
10	1.10392	141.36	1.04621	137.81	4.02642E-09	152.18	10.	1.00000	-113.37	1.	1.00000	-113.25	1.00000	0.38			
11	1.26570	-59.67	1.25026	-63.70	2.03181E-07	143.41	11.	1.00000	-05.55	1.	1.00000	-05.85	1.00000	-31.01			
12	0.62248	187.11	0.60065	181.31	2.56449E-06	-99.98	12.	1.00000	-6.22	1.	1.00000	-24.49	1.00000	-111.38			
13	1.05118	122.48	1.06193	128.25	3.10128E-06	-149.89	13.	1.00000	-40.50	1.	1.00000	-39.30	1.00000	51.85			
14	1.59518	139.4	1.16112	139.3	5.23479E-09	141.05	14.	1.00000	63.38	1.	1.00000	59.58	1.00000	09.53			
15	1.18574	169.46	1.17918	169.29	6.7945CE-29	-170.36	15.	1.00000	-8.49	1.	1.00000	-4.27	1.00000	-166.46			
16	0.87957	46.91	0.96176	44.27	2.62833E-06	52.42	16.	1.00000	-67.17	1.	1.00000	-67.57	1.00000	-68.49			
17	0.87204	45.82	0.92053	58.50	6.13256E-09	51.37	17.	1.00000	143.73	1.	1.00000	146.61	1.00000	-131.20			
18	0.98262	-168.33	0.89531	-165.58	4.49944E-29	-164.07	18.	1.00000	136.82	1.	1.00000	137.51	1.00000	-164.59			
19	0.90399	-177.80	0.95698	-179.94	1.00000	177.97	19.	1.00000	69.45	1.	1.00000	70.99	1.00000	-119.88			
20	1.07059	-4.56	1.19218	-5.95	5.91873E-09	-49.74	20.	1.00000	113.35	1.	1.00000	133.63	1.00000	42.66			
21	0.81568	174.05	0.85079	173.81	2.92069E-09	165.75	21.	1.00000	-164.39	1.	1.00000	-163.24	1.00000	-111.04			
22	0.77071	165.71	0.74168	173.58	1.94705E-06	162.85	22.	1.00000	163.17	1.	1.00000	165.38	1.00000	-166.82			
23	0.87968	-144.75	0.87257	-149.36	2.20327E-07	50.38	23.	1.00000	78.39	1.	1.00000	75.76	1.00000	-54.47			
24	0.87156	128.19	0.87079	120.48	1.56973E-06	-09.38	24.	1.00000	36.13	1.	1.00000	34.78	1.00000	-26.12			
25	1.37938	18.23	1.35452	19.16	2.78925E-06	59.58	25.	1.00000	-24.81	1.	1.00000	-24.08	1.00000	-12.82			
26	1.33882	-82.03	1.29344	-24.56	6.70175E-09	-39.38	26.	1.00000	66.45	1.	1.00000	98.63	1.00000	-92.58			
27	1.16795	-37.18	1.06358	-34.81	1.052539	22.92	27.	1.00000	116.52	1.	1.00000	122.43	1.00000	-78.91			

RMS SOLUTION ERROR:  $7.24 \times 10^{-8}$   $4.19 \times 10^{-8}$

Wild amplitudes  $0-10^8$  Jy

BADIF=F  
NOISEAMP= 8.28 JY  
WILDPCP= 12.54  
WILDAMP= 1.0000E+18 JY

RMS SOLUTION ERROR:  $4.46 \times 10^{-8}$   $1.29 \times 10^{-8}$

Wild amplitudes  $0-10^8$  Jy

BADIF=F  
NOISEAMP= 8.28 JY  
WILDPCP= 9.69  
WILDAMP= 1.0000E+18 JY

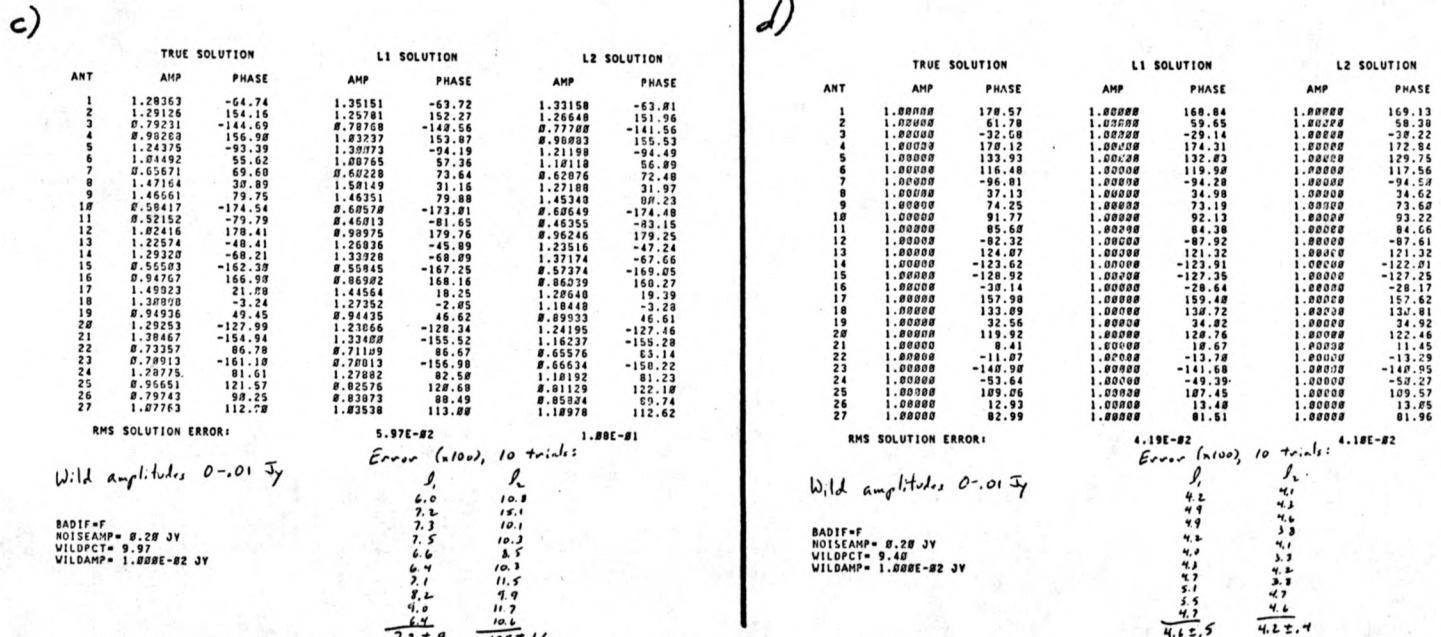


Figure 5. Sensitivity to the presence of extreme outliers. "Tame" observations have 0.2 Jy random noise added. Observations were made wild with probability .10. a) and b) illustrate the sensitivity to extremely high amplitude outliers, c) and d) to low amplitude.

TRUE SOLUTION			L1 SOLUTION			L2 SOLUTION			TRUE SOLUTION			L1 SOLUTION			L2 SOLUTION		
ANT	AMP	PHASE	AMP	PHASE		AMP	PHASE		ANT	AMP	PHASE	AMP	PHASE		AMP	PHASE	
1	1.49261	125.25	1.50945	126.49		1.51124	125.75		1	1.43904	177.39	1.44957	178.87		1.43397	179.93	
2	1.00722	130.45	0.90822	136.70		0.90021	136.43		2	1.02349	-104.74	0.86908	-104.54		0.87293	-104.20	
3	1.01440	116.41	1.01534	107.04		0.00093	125.07		3	1.24262	-107.87	1.27982	-107.72		1.20337	-107.69	
4	0.63325	-170.95	0.81095	-113.93		0.86512	-112.82		4	1.14319	142.52	1.14378	141.34		1.17417	141.14	
5	0.63127	-167.89	0.81094	-117.83		0.86514	-103.45		5	1.11540	-137.38	1.07171	-137.16		0.83264	-138.29	
6	1.42569	151.38	1.46252	152.27		1.45946	151.08		6	1.11540	8.78	1.08379	7.90		0.81956	6.65	
7	0.91442	-120.91	0.85093	-120.73		0.84603	-121.68		7	1.43107	165.00	1.45008	166.64		1.46218	167.00	
8	0.82261	-151.92	0.05772	-157.93		0.02466	-151.43		8	0.59960	64.07	1.06286	61.29		0.93301	61.73	
9	1.27472	-53.24	1.22659	-53.17		1.01900	-54.69		9	0.59960	-25.75	0.95123	-36.43		0.97977	-36.22	
10	0.76709	-0.15	0.75110	-05.48		0.71312	-05.03		10	0.03407	105.5	0.70289	-157.76		0.80982	-108.58	
11	1.15247	20.47	1.10510	22.77		1.02000	22.09		11	1.29460	70.73	1.03446	70.77		1.25725	77.33	
12	0.70171	161.63	0.69704	156.75		0.71526	150.96		12	0.15390	20.10	1.01111	28.71		0.92620	20.71	
13	0.81157	-131.06	0.82269	-131.90		0.79351	-132.26		13	0.92111	168.24	0.51075	-161.06		0.84844	-168.19	
14	1.06494	-39.55	1.11247	-38.31		1.09066	-30.67		14	1.19746	-107.00	1.20421	-108.26		1.20319	-106.17	
15	1.39381	51.14	1.44730	51.52		1.42700	50.66		15	1.02085	107.97	0.70475	-158.05		0.70374	-159.34	
16	1.17030	-53.32	1.17220	-55.79		1.40559	-52.84		16	1.20205	97.97	1.05638	93.03		1.10245	93.91	
17	0.54201	116.35	0.55000	120.01		0.57000	110.50		17	1.14913	180.02	1.02467	182.66		1.19619	171.11	
18	0.73123	-61.11	0.71993	-62.51		0.73427	-61.11		18	1.24647	-43.00	1.14439	-44.00		1.15857	-44.09	
19	1.46332	32.37	1.46991	33.59		1.45961	33.08		19	0.09917	53.23	1.10338	55.72		1.15655	56.35	
20	0.77432	55.17	0.72132	53.50		0.71290	51.85		20	0.94912	-80.70	0.70303	-96.32		1.01943	-76.31	
21	1.23373	-121.49	1.21943	-121.51		1.20554	-121.47		21	1.27900	105.00	0.74442	115.06		0.77451	117.19	
22	1.15276	-41.76	1.15276	-41.76		1.09302	-103.04		22	1.31967	119.76	1.27569	119.09		1.25032	121.72	
23	0.79703	113.76	0.74991	110.70		0.77155	112.76		23	1.29213	-62.93	1.27569	-73.55		1.28015	-7.62	
24	0.95745	-65.05	0.95765	-66.04		0.94493	114.14		24	0.07247	111.75	0.87475	111.92		0.85370	111.54	
25	1.07103	127.22	1.01295	129.02		1.01659	124.05		25	0.84909	121.01	0.05393	122.74		0.82059	122.78	
26	1.14591	-24.35	1.12014	-24.57		1.20247	-21.96		26	1.41296	-115.08	1.44376	-120.33		1.44160	-119.95	
27	0.57726	138.32	0.67140	134.22		0.64750	134.05		27	1.43104	-36.31	1.42295	-34.81		1.40458	-35.72	

RMS SOLUTION ERROR:  
(EXCLUDING ANTENNA 5) 5.79E-02 4.94E-02

0.2 Jy random noise  
0.1 Jy wild points on if 5

BADIF=T  
NOISEAMP= 0.28 JY  
WILDPCT= 0.00  
WILDAHP= 1.00E-01 JY

RMS SOLUTION ERROR:  
(EXCLUDING ANTENNA 5) 7.05E-02 6.68E-02

0.2 Jy random noise  
2.0 Jy wild points on if 5

BADIF=T  
NOISEAMP= 0.28 JY  
WILDPCT= 0.00  
WILDAHP= 2.00E+00 JY

RMS SOLUTION ERROR:  
(EXCLUDING ANTENNA 5) 7.05E-02 6.68E-02

0.2 Jy random noise  
5.0 Jy wild points on if 5

BADIF=T  
NOISEAMP= 0.28 JY  
WILDPCT= 0.00  
WILDAHP= 5.00E+00 JY

RMS SOLUTION ERROR:  
(EXCLUDING ANTENNA 5) 6.62E-02 9.74E-01

0.2 Jy random noise  
5.0 Jy wild points on if 5

BADIF=T  
NOISEAMP= 0.28 JY  
WILDPCT= 0.00  
WILDAHP= 5.00E+00 JY

Figure 6. Sensitivity of complex gain solutions to a bad i.f.

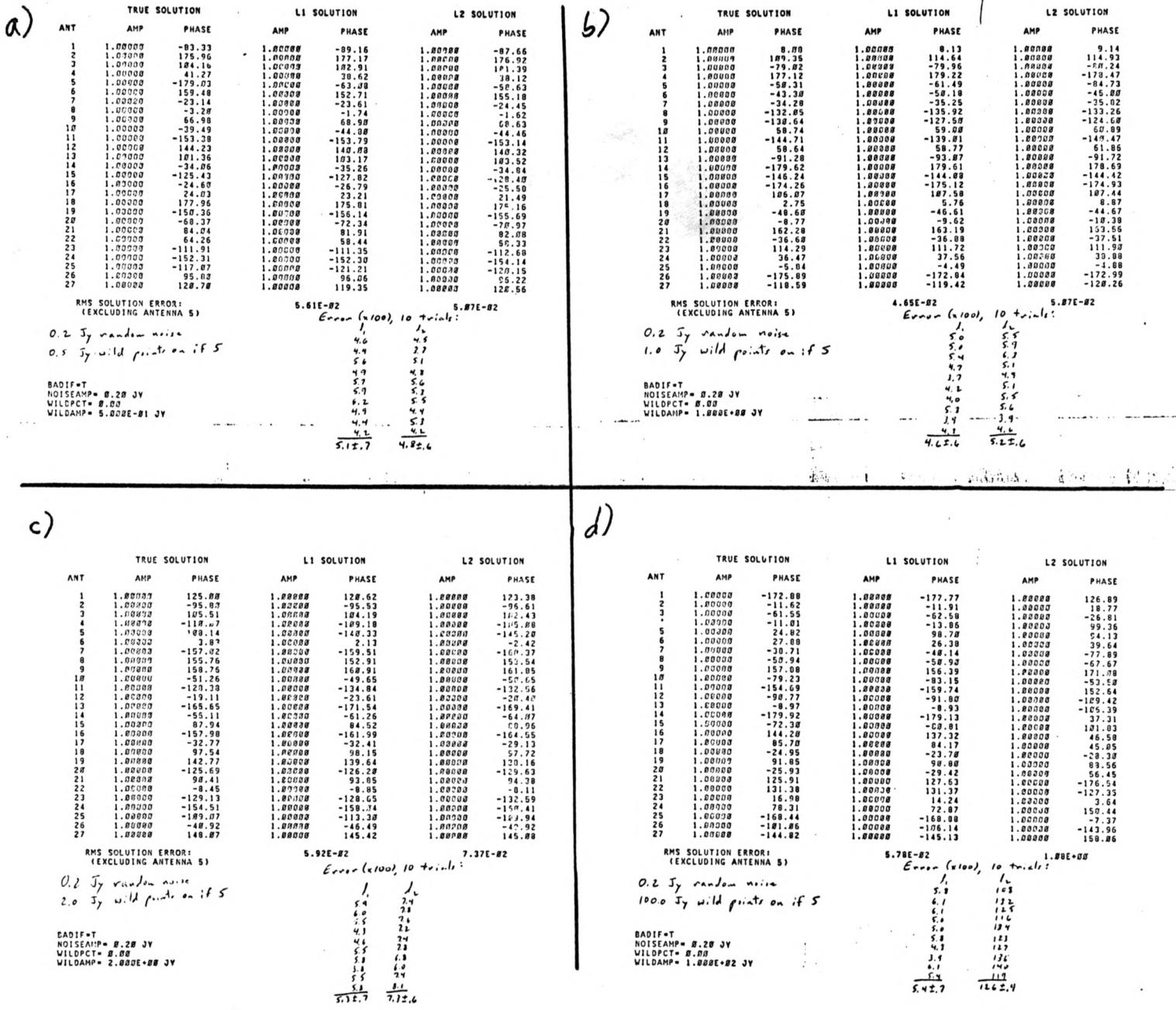


Figure 7. Sensitivity of phase solutions to a bad i.f.