Standard Deviation and Network/Local Accuracy of Geospatial Data (Note: Paper is self-published and filed with the U.S. Copyright Office) Earl F. Burkholder, PS, PE, F.ASCE – October 2013 President, Global COGO, Inc. – Las Cruces, NM 88003 Email: <u>eburk@globalcogo.com</u> URL: <u>www.globalcogo.com</u>

Abstract:

8 9 The Discussion/Closure items published in the February 2012 issue of the ASCE Journal of Surveying 10 Engineering address a difference of opinion that exists with regard to computing local accuracy of 11 geospatial data. The Discussion focuses on whether one or two rotation matrices should be used when 12 computing local accuracy. After reading the Closure, it appears that a more fundamental question is 13 whether or not a standard deviation can be used as a measure of local accuracy. If it is determined that 14 local accuracy cannot be computed in terms of a standard deviation of the separation between points, 15 then this response is moot. If it can, then a separate question needs to be considered - is local accuracy 16 to be considered as an attribute of a pair of points in 3-D space or is local accuracy an attribute of the 17 distance between two points? That distinction is important because each answer implies a different 18 functional model. It seems that authors in the Discussion/Closure talk past each other on that issue.

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20 This paper provides additional insight by computing the standard deviation of a 3-D slope distance with

the hypothesis that a distance standard deviation is an acceptable measure of accuracy. The advantage

is that the 3-D computation uses no rotation matrices – it is an independent computation. Standard
 deviations of short (1 km), medium (20km) and long (100 km) lines are computed using one rotation

24 matrix, two rotation matrices, and no rotation matrix. Comparison of the results between methods

25 shows very little difference in the computed standard deviations. Additionally, all three methods show

26 the same (significant) improvement of results due to using the off-diagonal cross-correlation portions of

27 the overall covariance matrix – local accuracy is better than network accuracy. The take-away is that

28 use of the stochastic model for local accuracy in Burkholder (2008) is validated and that the relative

29 location of one point with respect to another (network accuracy) can be computed using the covariance

30 matrices of the two points (no statistical correlation between points) but that a better answer (local

31 accuracy) is computed using the full covariance matrix between the two points.32

33 Introduction:

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The goal in writing the book, *The 3-D Global Spatial Data Model* (Burkholder 2008), was to start with the assumption of a single origin (Earth's center of mass) for 3-D geospatial data and to use the Earthcentered, Earth-fixed (ECEF) rectangular geocentric coordinate system as the basis of an efficient model for handling 3-D geospatial data. In the ECEF environment, long-standing rules of solid geometry

39 (functional model equations) are used to compute X/Y/Z positions in three-dimensional space.

40 Equivalent positions in other coordinate systems can be obtained using traditional transformation

41 equations found in Burkholder (2008) and in various texts. Standard deviations of derived quantities are

42 computed using stochastic model equations – in this case, the matrix formulation of error propagation
 43 given by:

44

$$\Sigma_{YY} = J_{YX} \ \Sigma_{XX} \ J_{XY}^T$$

47 Where Σ_{YY} is the covariance matrix of the computed result, Σ_{XX} is the covariance matrix of the variables, 48 and J_{YX} is the Jacobian matrix of partial derivatives of the computed result with respect to the variables.

(1)

49 Following-the-evidence and developing efficient procedures for handling spatial data were the intent 50 when writing the 3-D book. Computational results (e.g., geodetic forward/inverse computations and 51 network adjustments) can often be achieved more easily using the 3-D formulation rather than 52 traditional geodesy equations. In the case of spatial data accuracy, standard deviations were used by 53 Burkholder (1999) as the basis for computing what is called network accuracy and local accuracy. The 54 extension from computing a standard deviation to calling it network or local accuracy seems reasonable 55 and provides a concise mathematical basis for those accuracy definitions. However Soler & Smith (2010) 56 provide an alternative rigorous formulation for computing local accuracy that is more general and uses 57 the positional errors at each point as the basis of the functional model instead of the inverse distance. 58 Their implication is that local accuracy as presented in Chapter 11 of Burkholder (2008) lacks 59 appropriate rigor. That implication is made specific following equation (13) in Soler/Han/Smith (2012) 60 when they state that the approximation used by Burkholder "is not a general local accuracy estimate 61 and may be applied only when the two points are located very close to one another." Later in the same 62 paragraph they continue "...it is not an estimate of local accuracy as currently defined." Ironically, the 63 comparisons tabulated herein show very little difference in the accuracies computed using one or two 64 rotation matrices, even for points separated by 100 km. While there may be a technical difference in 65 the definitions of "standard deviation" and "local accuracy," computation of the standard deviation of a 66 3-D slope distance (that is nearly horizontal) and comparing those results with results obtained using 67 either one or two rotation matrices validates use of the stochastic model in Burkholder (2008). 68 69 This paper provides additional detail by computing the standard deviation of a 3-D mark-to-mark (slope) 70 distance between points. The reason for using this approach is that the standard deviation of the 3-D 71 slope distance can be computed from the equation of the 3-D inverse distance between points without 72 using a rotation matrix. But, this point must also be clear – the standard deviation of the slope distance

- is not the same as the standard deviation of the <u>horizontal distance</u> between points. Functional model equations for horizontal distance are different from the functional model equations for slope distance.
- 75 However, in cases where the horizontal distance agrees closely with the 3-D slope distance, the results
- 76 of the methods being compared should be nearly identical. Specifically, each of the three methods
- compared herein uses a different functional model equation but the same covariance matrix betweenpoints is used in all cases.
- 79
- The short (0.968 km) line cited herein is the example given in Chapter 11 of Burkholder (2008). The
 medium (21.87 km) and long (99.78 km) lines are a portion of the GPS CORS (WISCORS) network in
 Southeastern Wisconsin see Figure 1. RINEX data were downloaded from 9 CORS stations and 16 non trivial baselines were computed (ΔX, ΔY, ΔZ along with baseline covariance matrices) using off-the-shelf
- 84 software. Standard matrix manipulation software was used to compute the least squares adjustment of 85 the network from which the covariance matrices of the computed positions were obtained. Note that 86 although the computed coordinates of the network agreed your closely with the adopted COPS values
- although the computed coordinates of the network agreed very closely with the adopted CORS values,
 duplicating those ECEF coordinate values was not the objective. The (successful) objective was
- obtaining the covariances of the computed X/Y/Z coordinates of each point in the example network
 along with the correlation sub-matrices between points.
- 90

91 Functional Model:92

93 The functional model equation used to compute the slope distance between points is obtained from 94 ECEF coordinates as:

95

$$D = \sqrt{\Delta X^2 + \Delta Y^2 + \Delta Z^2}$$

(2)

97

98 Where D = the 3-D slope distance, $\Delta X = X_2 - X_1$, $\Delta Y = Y_2 - Y_1$, and $\Delta Z = Z_2 - Z1$. Note that the distance in 99 equation (2) is the same whether going from Point 1 to Point 2 or from Point 2 to Point 1, insuring the 90 existence of the commutative property for the standard deviation of the slope distance between points

100 existence of the commutative property for the standard deviation of the slope distance between points.

102 Solution:

103

This derivation uses the ECEF coordinates for each point, the covariance matrix at each point, the cross covariances between points, and the error propagation procedure as defined by equation (1). In this case, the answer will be a 1 x 1 matrix representing the variance of the 3-D spatial distance. Standard deviation is the square root of the variance. In order to use equation (1), the partial derivatives of the inverse distance with respect to each coordinate variable need to be computed and the matrix multiplications need to be performed. The Jacobian matrix (in this case a vector) of partial derivatives of the 3-D spatial distance with respect to each ECEF coordinate value is given by equation (3) as:

111

 $\boldsymbol{J} = \begin{bmatrix} \frac{\partial D}{\partial X_1} & \frac{\partial D}{\partial Y_1} & \frac{\partial D}{\partial Z_1} & \frac{\partial D}{\partial X_2} & \frac{\partial D}{\partial Y_2} & \frac{\partial D}{\partial Z_2} \end{bmatrix} = \begin{bmatrix} \frac{-\Delta X}{D} & \frac{-\Delta Y}{D} & \frac{-\Delta Z}{D} & \frac{\Delta X}{D} & \frac{\Delta Y}{D} & \frac{\Delta Z}{D} \end{bmatrix}.$ (3)

113

114 The covariance matrix of the variables, Σ_{XX} , equation (4), adopts the labeling convention used by 115 Soler/Smith (2010) and includes both the covariance matrix for each point and the cross covariance

- 116 matrices between the points.
- 117

118
$$\Sigma_{XX} = \begin{bmatrix} \Sigma_{11} = \begin{bmatrix} \sigma_{X_1}^2 & \sigma_{X_1Y_1} & \sigma_{X_1Z_1} \\ \sigma_{X_1Y_1} & \sigma_{Y_1}^2 & \sigma_{Y_1Z_1} \\ \sigma_{X_1Z_1} & \sigma_{Y_1Z_1} & \sigma_{Z_1}^2 \end{bmatrix} & \Sigma_{12} = \begin{bmatrix} \sigma_{X_1X_2} & \sigma_{X_1Y_2} & \sigma_{X_1Z_2} \\ \sigma_{Y_1X_2} & \sigma_{Y_1Y_2} & \sigma_{Y_1Z_2} \\ \sigma_{Z_1X_2} & \sigma_{Z_1Y_2} & \sigma_{Z_1Z_2} \end{bmatrix} \\ \Sigma_{21} = \begin{bmatrix} \sigma_{X_1X_2} & \sigma_{Y_1X_2} & \sigma_{Z_1X_2} \\ \sigma_{X_1Y_2} & \sigma_{Y_1Y_2} & \sigma_{Z_1Y_2} \\ \sigma_{X_1Z_2} & \sigma_{Y_1Z_2} & \sigma_{Z_1Z_2} \end{bmatrix} & \Sigma_{22} = \begin{bmatrix} \sigma_{X_2}^2 & \sigma_{X_2Y_2} & \sigma_{X_2Z_2} \\ \sigma_{X_2Y_2} & \sigma_{Y_2Z_2}^2 & \sigma_{Y_2Z_2} \\ \sigma_{X_2Z_2} & \sigma_{Y_2Z_2} & \sigma_{Z_2}^2 \end{bmatrix} \end{bmatrix}$$
(4)

119

120 Two equivalent methods can be used to compute the variance and standard deviation of the 3-D slope 121 distance. First, matrix multiplication will provide the variance for the 3-D slope distance by using values 122 of the partial derivatives from equation (3) and the covariance values from equation (4).

Second, the variance can be obtained by first substituting equations (3) and (4) into equation (1) and
performing the algebraic operations. The result is:

126

$$127 \qquad \sigma^{2} = \frac{\Delta X^{2}}{D^{2}} \left(\sigma_{X_{1}}^{2} + \sigma_{X_{2}}^{2} \right) + \frac{\Delta Y^{2}}{D^{2}} \left(\sigma_{Y_{1}}^{2} + \sigma_{Y_{2}}^{2} \right) + \frac{\Delta Z^{2}}{D^{2}} \left(\sigma_{Z_{1}}^{2} + \sigma_{Z_{2}}^{2} \right) \\ + \frac{2}{D^{2}} \left[\Delta X \Delta Y \left(\sigma_{X_{1}Y_{1}} + \sigma_{X_{2}Y_{2}} \right) + \Delta X \Delta Z \left(\sigma_{X_{1}Z_{1}} + \sigma_{X_{2}Z_{2}} \right) + \Delta Y \Delta Z \left(\sigma_{Y_{1}Z_{1}} + \sigma_{Y_{2}Z_{2}} \right) \right] \\ 130 \\ 131 \qquad - \frac{2}{D^{2}} \left[\Delta X \Delta Y \left(\sigma_{X_{1}Y_{2}} + \sigma_{Y_{1}X_{2}} \right) + \Delta X \Delta Z \left(\sigma_{X_{1}Z_{2}} + \sigma_{Z_{1}X_{2}} \right) + \Delta Y \Delta Z \left(\sigma_{Y_{1}Z_{2}} + \sigma_{Z_{1}Y_{2}} \right) \right] \\ 132 \\ 133 \qquad - \frac{2}{D^{2}} \left[\Delta X^{2} \sigma_{X_{1}X_{2}} + \Delta Y^{2} \sigma_{Y_{1}Y_{2}} + \Delta Z^{2} \sigma_{Z_{1}Z_{2}} \right].$$
(5)

135

136	In either case, the standard deviation of the mark-to-mark (point-to-point) distance between any pair of					
13/	points defined with geocentric X/Y/Z coordinates (in the same datum) and associated covariance					
138	matrices is commutative a	nd computed as:				
139						
140	$\sigma = \sqrt{\sigma^2}$				(6)	
141						
142	Note here that the first tw	o terms in equation (5) will p	rovide the varia	nce of the 3-D sl	ope distance	
143	under the assumption of r	o correlation, i.e., the cross c	ovariance value	s are all zero in t	he third and	
144	fourth terms of equation (5).				
145						
146	Data:					
147						
148	Short line data – Stations	Pseudo and USPA from Chapte	er 11, Burkholde	er (2008):		
149				. ,		
150	Standpoint (Point 1): Psei	ıdo				
151	Geocentric coordinates	ECEF station co	wariance matrix	, m ² (Σ ₁₁ in eq. (4	+))	
152	X = -1.556.206.615 m	1.24245E-06	1.34309E-06	-9.21239E-07		
153	Y = -5.169.400.740 m	1.34309E-06	4.50646E-06	-2.77505E-06		
154	Z = 3.387.285.987 m	-9.21239F-07	-2.77505E-06	4.18236F-06		
155		0.22002.07				
156	Forepoint (Point 2): USPA					
157	Geocentric coordinates	ECEF station co	ovariance matrix	x_{1} m ² (Σ_{22} in eq. (4)	4))	
158	X = -1.555.678.579 m	2.16088F-06	2.34746F-06	-1.49586F-06	-11	
159	Y = -5.169.961.396 m	2.34746F-06	8.47410F-06	-5.01722F-06		
160	Z = 3.386.700.089 m	-1.49586E-06	-5.01722E-06	6.81155E-06		
161						
162	Correlatio	n		Correlation		
163	Σ_{21} in equation	$\frac{1}{2}$ (4). m ²	Σ12 Ϊ	n equation (4), r	n ²	
164	1.1405F-06 1.2934F-0	6 -8.8294F-07	1.1405F-06	1.2848F-06	-8.9269F-07	
165	1.2848F-06 4.2401F-0	6 -2.5676F-06	1.2934F-06	4.2401F-06	-2.5759F-06	
166	-8.9269E-07 -2.5759E-	06 3.6030E-06	-8.8294E-07	-2.5676E-06	3.6030E-06	
167						
168	Medium line data – Statio	ns FSRI and RASN in Southeas	tern Wisconsin	– see Figure 1:		
169				0.1		
170	Standpoint (Point 1): FRS					
171	Geocentric coordinates	ECEF station co	wariance matrix	. m² (Σ ₁₁ in eq. (4	L))	
172	X = 164.796.251 m	2.137380E-04	- 1.347237E-06	9.504950E-07		
173	Y = -4.679.186.506 m	-1.347237E-06	2.7009126E-04	4 -4.442558E-05		
174	Z = 4.316.889.723 m	9.504950F-07	-4.442558F-05	2.544635F-04		
175						
176	Forepoint (Point 2): RASN					
177	Geocentric coordinates	ECEF station co	wariance matrix	$m^{2} (\Sigma_{22} in eq. (4))$	L))	
178	X = 153.059.630 m	2.136655F-04	-9.939590F-07	3.374190F-07		
179	Y = -4.666.858.811 m	-9.939590F-07	2.702019F-04	-4.501482F-05		
180	Z = 4.330.615.801 m	3.374190F-07	-4.501482F-05	2.551067F-04		
181	,	3.37 11902 07				
182						
183						

184	Correlation	Correlation			
185	Σ_{21} in equation (4), m ²			Σ_{12} in equation	(4), m ²
186	2.128074E-04 -5.581210E-07	9.322200E-08	2.128074E-04	-4.365860E-07	-7.018000E-09
187	-4.365860E-07 2.522824E-04	-3.039416E-05	-5.581210E-07	2.522824E-04	-3.054197E-05
188	-7.018000E-09 -3.054197E-06	2.401769E-04	9.322200E-08	-3.039416E-05	2.401769E-04
189					
190	The long line data – Stations FS	RI and SHAN in SE Wisco	nsin – see Figure	e 1:	
191	5		0		
192	Standpoint (Point 1): FRSI				
193	Geocentric coordinates	ECEF station co	variance matrix,	$m^{2} (\Sigma_{11} in eq. (4$))
194	X = 164,796.251 m	2.137380E-04	-1.347237E-06	9.504950E-07	,,
195	Y = -4.679.186.506 m	-1.347237E-06	2.700913E-04	-4.442558E-05	
196	Z = 4.316.889.723 m	9.504950E-07	-4.442558E-05	2.544635E-04	
197	,,				
198	Forepoint (Point 2): SHAN				
199	Geocentric coordinates	FCFF station co	variance matrix.	$m^{2}(\Sigma_{22})$ in eq. (4	.))
200	X = 182.409.481 m	2.130396F-04	-1.320212F-06	6.137860F-07	//
201	Y = -4.611.414.938 m	-1 320212F-06	2 498616F-06	-2 975560F-05	
202	7 = 4387983545 m	6 137860F-07	-2 975560E-05	2.37330002 03	
202	2 - +,307,303.3+3 m	0.1370002 07	2.5755002 05	2.5502572.04	
203	Correlation			Correlation	
205	Σ_{24} in equation (4) m	2	Σ _{sa} i	n equation (4)	m^2
205	2 123000F-04 -3 206330F-07	-2 744470E-07	2 123000E-04	-6 70/10F-08	-5 27/200E-07
200	-6 70/10E-08 2 353/120E-0/	-2.744470E-07	-3 206330E-04	2 353/20F-0/	-3.274200E-07
207	-5.774200E-07 -1.911484E-05	2 288600E-03	-3.200330E-07	-1 02/826E-05	-1.511404E-05
208	-5.2742002-07 -1.9114842-05	2.20000E-04	-2.744470E-07	-1.924020E-03	2.200000E-04
207	Computation of 2 D Standard [) oviations:			
210	Computation of 5-D Standard E				
211	In each case, equation (5) is	used to compute the s	tandard deviati	ion of the 3-D r	mark-to-mark
212	distance between points for t	wo senarate circumsta	ances.		naik-to-maik
213	distance between points for t				
215	 Using the full covarian 	ce matrix in equation (4) and		
215	Assuming no correlation	on hetween points by e	liminating the t	third and fourth	terms in
210	equation (5)				
218					
219	For comparison, the slope of	each vector is listed for	or verifvina how	close the slop	e distance is to
220	being horizontal. The accura	icv values shown belo	w were comput	ed as standard	deviations (an
221	unambiguous well-defined m	athematical process) b	out are listed as	s network accu	racv and local
222	accuracy.	1 /			,
223					
224			Network Accu	racy Loca	al Accuracy
225	Short line Pseudo to USPA - <u>(</u>	<u>).968 km</u>			
226	Slope = 89° 16' 18"		0.0018	0 m	0.00105 m
227	Medium line FRSI to RASN – 2	21.87 km			
228	Slope = 89° 59′ 07″		0.0208	1 m	0.00179 m
229	Long line FRSI to SHAN – <u>99.7</u>	<u>8 km</u>			
230	Slope = 89° 31' 44"		0.0207	7 m	0.00251 m
231					
232	Two important points to be m	ade in reference to the	e slope distanc	e standard dev	riations are:

- 233
- These results are consistent with the conclusion in Soler/Han/Smith (2012) that significant improvement in the uncertainty between points is realized by using the full covariance matrix between points as opposed to using only the covariance matrices of the two endpoints. If the cross covariance matrices are not used (their values are zero) the two endpoints are statistically independent of each other.
- 239
- 240
 24. The 3-D accuracy values above are readily computed and, because the lines are nearly
 24. horizontal, those values compare very favorably with the accuracy values in the following
 24. sections as computed using both Burkholder's (1999) one-rotation-matrix procedure and
 24. the rigorous two-rotation-matrix procedure defined by Soler/Smith (2010).

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Computation of Local Accuracies:

This section shows computation of both network and local accuracies by Soler/Smith (2010) and Burkholder (1999) using the same data for short, medium, and long lines as was used for computing the standard deviation of the slope distance. First, a comparison is made between the values of the 3x3 covariance matrix in each case (short, medium, long) used in equation (14) by Soler/Smith (2010) and equation (11) by Burkholder (Discussion 2012) which is shown to be the same as the procedure used in Burkholder (1999). Following that, a comparison is made of the computed network and local accuracies of the distances between points.

For convenient reference, the equations for the local perspective covariance matrix of the separation between points are:

258 Soler/Smith (2010), eq. (14)
$$\Sigma_{\Delta e, \Delta n, \Delta u} = R_1 \Sigma_{11} R_1^T - R_1 \Sigma_{12} R_2^T - R_2 \Sigma_{21} R_1^T + R_2 \Sigma_{22} R_2^T$$
 (7)

259 260

261 Burkholder (2012), eq. (11)
$$\Sigma_{\Delta e, \Delta n, \Delta u} = R \Sigma_{11} R^T - R \Sigma_{12} R^T - R \Sigma_{21} R^T + R \Sigma_{22} R^T$$
 (8)

262

263 The 3x3 covariance matrix of the local p264 Short line on the NMSU campus is:

265

266

$$\Sigma_{\Delta e,\Delta n,\Delta u} = \begin{bmatrix} 7.87895E - 7 & 1.22998E - 7 & 8.36203E - 8 \\ 1.22998E - 7 & 1.62944E - 6 & 6.97400E - 7 \\ 8.36203E - 8 & 6.97400E - 7 & 6.99166E - 6 \end{bmatrix} & \left\{ \begin{matrix} \sigma_{\Delta e} = 0.0009 \ m \\ \sigma_{\Delta n} = 0.0013 \ m \\ \sigma_{\Delta u} = 0.0026 \ m \end{matrix} \right\}$$
(9)

The 3x3 covariance matrix of the local perspective computed by equation (8) above for the Short line on the NMSU campus is:

270

271
$$\Sigma_{\Delta e,\Delta n,\Delta u} = \begin{bmatrix} 7.87895E - 7 & 1.22991E - 7 & 8.35829E - 8 \\ 1.22991E - 7 & 1.62948E - 6 & 6.97427E - 7 \\ 8.35829E - 8 & 6.97427E - 7 & 6.99162E - 6 \end{bmatrix} & \left\{ \begin{array}{c} \sigma_{\Delta e} = 0.0009 \, m \\ \sigma_{\Delta n} = 0.0013 \, m \\ \sigma_{\Delta u} = 0.0026 \, m \end{array} \right\}$$
(10)

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The 3x3 covariance matrix of the local perspective computed by equation (7) above for the
 Medium line in the GPS network in Wisconsin is:

276
$$\Sigma_{\Delta e,\Delta n,\Delta u} = \begin{bmatrix} 1.73686E - 6 & 4.01300E - 8 & 2.96738E - 7 \\ 4.01300E - 8 & 3.81055E - 6 & -1.23848E - 6 \\ 2.96738E - 7 & -1.23848E - 6 & 6.11934E - 5 \end{bmatrix} & \left\{ \begin{matrix} \sigma_{\Delta e} = 0.0013 \ m \\ \sigma_{\Delta n} = 0.0020 \ m \\ \sigma_{\Delta u} = 0.0078 \ m \end{matrix} \right\}$$
(11)

The 3x3 covariance matrix of the local perspective computed by equation (8) above for the Medium line in the GPS network in Wisconsin is:

- 280
- 281

 $\Sigma_{\Delta e,\Delta n,\Delta u} = \begin{bmatrix} 1.73701E - 6 & 3.68349E - 8 & 3.51352E - 7 \\ 3.68349E - 8 & 3.81215E - 6 & -1.32017E - 6 \\ 3.51352E - 7 & -1.32017E - 6 & 6.11844E - 5 \end{bmatrix} & \left\{ \begin{matrix} \sigma_{\Delta e} = 0.0013 \ m \\ \sigma_{\Delta n} = 0.0020 \ m \\ \sigma_{\Delta u} = 0.0078 \ m \end{matrix} \right\}$ (12)

282

The 3x3 covariance matrix of the local perspective computed by equation (7) above for the Long line in the GPS network in Wisconsin is:

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286 $\Sigma_{\Delta e,\Delta n,\Delta u} = \begin{bmatrix} 2.07569E - 6 & 3.98260E - 7 & 1.13848E - 6 \\ 3.98260E - 7 & 6.39059E - 6 & -4.44871E - 6 \\ 1.13848E - 6 & -4.44871E - 6 & 7.88627E - 5 \end{bmatrix} & \left\{ \begin{array}{c} \sigma_{\Delta e} = 0.0014 \ m \\ \sigma_{\Delta n} = 0.0025 \ m \\ \sigma_{\Delta u} = 0.0089 \ m \end{array} \right\}$ (13)

287

The 3x3 covariance matrix of the local perspective computed by equation (8) above for the Long
 line in the GPS network in Wisconsin is:

- 291 $\Sigma_{\Delta e,\Delta n,\Delta u} = \begin{bmatrix} 2.07098E 6 & 3.99797E 7 & 9.54143E 7 \\ 3.997967E 7 & 6.42459E 6 & -5.22940E 6 \\ 9.54143E 7 & -5.22940E 6 & 7.87179E 5 \end{bmatrix} & \left\{ \begin{array}{c} \sigma_{\Delta e} = 0.0014 \ m \\ \sigma_{\Delta n} = 0.0025 \ m \\ \sigma_{\Delta u} = 0.0089 \ m \end{array} \right\}$ (14)
- 292

293 Comparison of the covariance matrices computed using equation (7) and equation (8) for all 294 three (short, medium, long) lines shows that they are, in fact, slightly different. However, when 295 taking the square root of the diagonal elements (down to the tenth of a millimeter), there is no 296 difference between the results obtained from equation (7), called rigorous local accuracy by 297 Soler/Smith, and equation (8), computed as a standard deviation of the horizontal distance by 298 Burkholder (2008) and called local accuracy.

In order to facilitate a comparison of similar quantities, the covariance values in equations (9) &
 (10), (11) & (12), and (13) & (14) were used to compute the standard deviation of the horizontal
 distance - the assumption in Burkholder (1999) - and the standard deviation of the separation
 between points - called rigorous local accuracy by Soler/Smith (2010).

As listed in Burkholder (1999) in equations (12) to (15), the standard deviation of the horizontal
 distance is computed using the partial derivatives of equation (15) below and the local 3x3
 covariance matrix computed above.

$$HD = \sqrt{\Delta e^2 + \Delta n^2} \tag{15}$$

311 The partial derivatives in the Jacobian matrix as obtained from equation (15) are:

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 $J = \begin{bmatrix} \frac{\partial HD}{\partial \Delta e} & \frac{\partial HD}{\partial \Delta u} & \frac{\partial HD}{\partial \Delta u} \end{bmatrix} = \begin{bmatrix} \frac{\Delta e}{HD} & \frac{\Delta n}{HD} & 0 \end{bmatrix}$ (16)

The matrix product defined in equation (1) was used to multiply the Jacobian values of equation (16) and the local 3x3 covariance matrices in equations (10), (12), and (14) to compute the variances (and standard deviations) of the horizontal distance for the short, medium, and long lines as follows:

Network Accuracy	Local Accuracy
<u>968 km</u> 0.00180 m	0.00105 m
<u>1.866 km</u> 0.02081 m	0.00179 m
1	<u>Network Accuracy</u> 0.00180 m 0.02081 m

324 325 326	Long line FI	RSI to SHAN -	- <u>99.784 km</u>		0.0207	77 m	0.00250 m	
327 328 329 330	Since the derivation of local accuracy by Soler/Smith (2010) does not specify horizontal distance, the local accuracy computation here uses all three components of the separation between points							
331 332		$SD = \sqrt{\Delta e^2}$	$+ \Delta n^2 + \Delta u^2$				(17)	
333 334	The partial de	e partial derivatives in the Jacobian matrix as obtained from equation (17) are:						
335 336		$J = \begin{bmatrix} \frac{\partial SD}{\partial \Delta e} & \frac{\partial SD}{\partial \Delta e} \end{bmatrix}$	$\left[\frac{D}{n} \frac{\partial SD}{\partial \Delta u}\right] = \left[\frac{\Delta e}{SD}\right]$	$\frac{\Delta n}{SD} \; \frac{\Delta u}{SD} \; \bigg]$			(18)	
337 338 339 340	The matrix product defined in equation (1) was used to multiply the Jacobian values of equation (18) and the local 3x3 covariance matrices in equations (9), (11), and (13) to compute the variances (and standard deviations) of for the short, medium, and longs lines as follows (Soler/Smith 2010):							
341		2010).			Network Accu	uracy Loc	Local Accuracy	
342 343 344	Short line F	Pseudo to USF	PA - <u>0.968 km</u>		0.0018	80 m	0.00105 m	
345 346	Medium lin	Medium line FRSI to RASN – <u>21.866 km</u>		<u>n</u>	0.0208	31 m	0.00179 m	
347 348 349	Long line FRSI to SHAN – <u>99.784 km</u>				0.02078 m 0.00254 m			
350 351	Summary of	results:						
352	A summary of	of results, bot	h network an	d local, for a	ll three lines (s	short, medium,	long) and three	
353	methods (3-	D standard de	eviation, Burk	cholder 1 rota	ation matrix, ar	nd Soler/Smith	2 matrix	
354 355 356	rotations) is i meters canno	rotations) is included in the following table. All units are meters. Showing five decimal places of meters cannot be justified except to show where differences in the answers begin to occur.						
357 358 359	Line	3-D Slope I <u>Network</u>	Distance Local	Burkholde <u>Network</u>	r, Hor. Dist. <u>Local</u>	Soler/S <u>Network</u>	mith Local	
360 361	Short	0.00180 m	0.00105 m	0.00180 m	0.00105 m	0.00180 m	0.00105 m	
362 363	Medium	0.02081 m	0.00179 m	0.02081 m	0.00179 m	0.02081 m	0.00179 m	
364 365	Long	0.02077 m	0.00251 m	0.02077 m	0.00250 m	0.02078 m	0.00254 m	
366 367 368	Observations/Conclusions:							
369 370 371	1. All three merit.	e methods list	ed herein giv	e essentially	the same ans	swer. Each ap	proach has	
372 373 374 375	2. The mat computa that the the com	terial in Burkl ation of stanc definition hol putation of its	nolder (2008) lard deviation rizontal distar s standard de	remains val of the horizonce over a lo eviation suffe	id. That deriva ontal distance ng distance be rs. See Burkh	ation is rigorou between point comes meani older (1991).	s for s. It appears ngless before	

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 3. The method of Soler/Smith is not restricted to horizontal distance and there is no limit on
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- 4. The computation of standard deviation for a 3-D spatial distance also has no limit on the distance between points and enjoys the commutative property. The attractive feature of this method is that the answer is computed directly in terms of the 3-D inverse distance between points and the full covariance matrix of that pair of points. No rotation matrices are required and there are no approximations in the algorithm.
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388 Further Study:389

390 Considerations in this paper focus on computation of standard deviation (or network/local 391 accuracy) based upon a given pair of points and their covariance matrices. The computed 392 relative position of one point with respect to another can be different if different covariance 393 values are used. The covariance values for points in the network may be guite different 394 depending upon which control points are held in an adjustment (how the adjustment was 395 constrained). For example, the Wisconsin GPS network cited in this example was minimally 396 constrained with station "FOLA" being assigned 0.010 m standard deviation in all three 397 components. Other scenarios were also computed and the results were found to support and 398 be consistent with the "dam monitoring" example described in Burkholder (2004). The study 399 and comparison of those "local accuracies" will be reported in a separate paper. 400

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Network Overlay – Burkholder, 2013

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Figure 1 - GPS Network in SE Wisconsin