

Supplement to the paper “Asymptotic expansions for the ability estimator in item response theory”

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This note is to supplement Ogasawara (2012).

1. Expansions of $\hat{h}^{1/2}$, $\hat{g}^{1/2}$ and $\hat{h}\hat{g}^{-1/2}$

For the expansion of $\hat{h}^{1/2}$, recalling

$$\begin{aligned}\hat{\theta} - \theta_0 &= \lambda^{(1)} l_0^{(1)} + \boldsymbol{\lambda}^{(2)'} \mathbf{l}_0^{(2)} + O_p(n^{-3/2}) \\ &= -\lambda^{-1} \frac{\partial \bar{l}}{\partial \theta_0} + \lambda^{-2} m \frac{\partial \bar{l}}{\partial \theta_0} - \frac{\lambda^{-1}}{2} E_T(j_0^{(3)}) \left(\lambda^{-1} \frac{\partial \bar{l}}{\partial \theta_0} \right)^2 + O_p(n^{-3/2}),\end{aligned}$$

we have

$$\begin{aligned}\hat{h}^{1/2} &= h_0^{1/2} + \frac{h_0^{-1/2}}{2} \frac{\partial h_0}{\partial \theta_0} (\hat{\theta} - \theta_0) + \left\{ \frac{h_0^{-1/2}}{4} \frac{\partial^2 h_0}{\partial \theta_0^2} - \frac{h_0^{-3/2}}{8} \left(\frac{\partial h_0}{\partial \theta_0} \right)^2 \right\} (\hat{\theta} - \theta_0)^2 \\ &\quad + O_p(n^{-3/2}) \\ &= (-\lambda)^{1/2} - \frac{(-\lambda)^{-1/2}}{2} m - \frac{(-\lambda)^{-3/2}}{8} m^2 + \frac{1}{2} \left\{ (-\lambda)^{-1/2} + \frac{(-\lambda)^{-3/2}}{2} m \right\} \frac{\partial h_0}{\partial \theta_0} \\ &\quad \times (\lambda^{(1)} l_0^{(1)} + \boldsymbol{\lambda}^{(2)'} \mathbf{l}_0^{(2)}) \\ &\quad + \left[\frac{(-\lambda)^{-1/2}}{4} E_T \left(\frac{\partial^2 h_0}{\partial \theta_0^2} \right) - \frac{(-\lambda)^{-3/2}}{8} \left\{ E_T \left(\frac{\partial h_0}{\partial \theta_0} \right) \right\}^2 \right] (\lambda^{(1)} l_0^{(1)})^2 \\ &\quad + O_p(n^{-3/2})\end{aligned}$$

$$\begin{aligned}
& = (-\lambda)^{1/2} + \left\{ -\frac{(-\lambda)^{-1/2}}{2}, -\frac{(-\lambda)^{-3/2}}{2} E_T(j_0^{(3)}) \right\} (m, l_0^{(1)})' \\
& + \left[-\frac{(-\lambda)^{-3/2}}{8}, -\frac{3}{4}(-\lambda)^{-5/2} E_T(j_0^{(3)}), -\frac{3}{8}(-\lambda)^{-7/2} \{E_T(j_0^{(3)})\}^2 \right. \\
& \left. -\frac{(-\lambda)^{-5/2}}{4} E_T(j_0^{(4)}), -\frac{(-\lambda)^{-3/2}}{2} \right] [m^2, m l_0^{(1)}, (l_0^{(1)})^2, \{j_0^{(3)} - E_T(j_0^{(3)})\} l_0^{(1)}]' \\
& + O_p(n^{-3/2}) \\
& \equiv (-\lambda)^{1/2} + \sum_{i=1}^2 \mathbf{h}^{(i)}' \mathbf{m}_0^{(i)} + O_p(n^{-3/2}).
\end{aligned}$$

For $\hat{g}^{1/2}$,

$$\begin{aligned}
\hat{g}^{1/2} & = g_0^{1/2} + \frac{g_0^{-1/2}}{2} \frac{\partial g_0}{\partial \theta_0} (\hat{\theta} - \theta_0) + \left\{ \frac{g_0^{-1/2}}{4} \frac{\partial^2 g_0}{\partial \theta_0^2} - \frac{g_0^{-3/2}}{8} \left(\frac{\partial g_0}{\partial \theta_0} \right)^2 \right\} (\hat{\theta} - \theta_0)^2 \\
& + O_p(n^{-3/2}).
\end{aligned}$$

Recalling $\gamma = E_T(g_0)$ and $g_0 = \gamma + m_g$ with $m_g = O_p(n^{-1/2})$, the above result becomes

$$\begin{aligned}
& = \gamma^{1/2} + \frac{\gamma^{-1/2}}{2} m_g - \frac{\gamma^{-3/2}}{8} m_g^2 + \frac{1}{2} \left\{ \gamma^{-1/2} - \frac{\gamma^{-3/2}}{2} m_g \right\} \frac{\partial g_0}{\partial \theta_0} (\lambda^{(1)} l_0^{(1)} + \boldsymbol{\lambda}^{(2)}' \mathbf{l}_0^{(2)}) \\
& + \left[\frac{\gamma^{-1/2}}{4} E_T \left(\frac{\partial^2 g_0}{\partial \theta_0^2} \right) - \frac{\gamma^{-3/2}}{8} \left\{ E_T \left(\frac{\partial g_0}{\partial \theta_0} \right) \right\}^2 \right] (\lambda^{(1)} l_0^{(1)})^2 + O_p(n^{-3/2}) \\
& = \gamma^{1/2} + \left\{ \frac{\gamma^{-1/2}}{2}, -\frac{\gamma^{-1/2}}{2} \lambda^{-1} E_T \left(\frac{\partial g_0}{\partial \theta_0} \right) \right\} (m_g, l_0^{(1)})'
\end{aligned}$$

$$\begin{aligned}
& + \left[-\frac{\gamma^{-3/2}}{8}, \frac{\gamma^{-3/2}}{4} \lambda^{-1} E_T \left(\frac{\partial g_0}{\partial \theta_0} \right), \frac{\gamma^{-1/2}}{2} \lambda^{-2} E_T \left(\frac{\partial g_0}{\partial \theta_0} \right), \right. \\
& \quad \left. - \frac{\gamma^{-1/2}}{4} \lambda^{-3} E_T \left(\frac{\partial g_0}{\partial \theta_0} \right) E_T(j_0^{(3)}) \right. \\
& \quad \left. + \frac{\gamma^{-1/2}}{4} \lambda^{-2} E_T \left(\frac{\partial^2 g_0}{\partial \theta_0^2} \right) - \frac{\gamma^{-3/2}}{8} \lambda^{-2} \left\{ E_T \left(\frac{\partial g_0}{\partial \theta_0} \right) \right\}^2, - \frac{\gamma^{-1/2}}{2} \lambda^{-1} \right] \\
& \times \left[m_g^2, m_g l_0^{(1)}, m l_0^{(1)}, (l_0^{(1)})^2, \left\{ \frac{\partial g_0}{\partial \theta_0} - E_T \left(\frac{\partial g_0}{\partial \theta_0} \right) \right\} l_0^{(1)} \right]' + O_p(n^{-3/2}) \\
& \equiv \gamma^{1/2} + \sum_{i=1}^2 \mathbf{g}^{(i)} \cdot \mathbf{m}_{g0}^{(i)} + O_p(n^{-3/2}).
\end{aligned}$$

For $\hat{h} \hat{g}^{-1/2}$,

$$\begin{aligned}
\hat{h} \hat{g}^{-1/2} & = h_0 g_0^{-1/2} + \left(\frac{\partial h_0}{\partial \theta_0} g_0^{-1/2} - \frac{h_0}{2} g_0^{-3/2} \frac{\partial g_0}{\partial \theta_0} \right) (\hat{\theta} - \theta_0) + \frac{1}{2} \left\{ -g_0^{-3/2} \frac{\partial h_0}{\partial \theta_0} \frac{\partial g_0}{\partial \theta_0} \right. \\
& \quad \left. + g_0^{-1/2} \frac{\partial^2 h_0}{\partial \theta_0^2} - \frac{h_0}{2} g_0^{-3/2} \frac{\partial^2 g_0}{\partial \theta_0^2} + \frac{3}{4} h_0 g_0^{-5/2} \left(\frac{\partial g_0}{\partial \theta_0} \right)^2 \right\} (\hat{\theta} - \theta_0)^2 \\
& \quad + O_p(n^{-3/2}) \\
& = -(\lambda + m) \left(\gamma^{-1/2} - \frac{\gamma^{-3/2}}{2} m_g + \frac{3}{8} \gamma^{-5/2} m_g^2 \right) + \left\{ -j_0^{(3)} \left(\gamma^{-1/2} - \frac{\gamma^{-3/2}}{2} m_g \right) \right. \\
& \quad \left. + \frac{1}{2} (\lambda + m) \left(\gamma^{-3/2} - \frac{3}{2} \gamma^{-5/2} m_g \right) \frac{\partial g_0}{\partial \theta_0} \right\} (\lambda^{(1)} l_0^{(1)} + \boldsymbol{\lambda}^{(2)} \cdot \mathbf{l}_0^{(2)}) \\
& \quad + \frac{1}{2} \left[\gamma^{-3/2} E_T(j_0^{(3)}) E_T \left(\frac{\partial g_0}{\partial \theta_0} \right) - \gamma^{-1/2} E_T(j_0^{(4)}) \right. \\
& \quad \left. + \frac{\lambda}{2} \gamma^{-3/2} E_T \left(\frac{\partial^2 g_0}{\partial \theta_0^2} \right) - \frac{3}{4} \lambda \gamma^{-5/2} \left\{ E_T \left(\frac{\partial g_0}{\partial \theta_0} \right) \right\}^2 \right] (\lambda^{(1)} l_0^{(1)})^2 \\
& \quad + O_p(n^{-3/2})
\end{aligned}$$

$$\begin{aligned}
&= -\lambda \gamma^{-1/2} + \left\{ -\gamma^{-1/2}, \frac{\gamma^{-3/2}}{2} \lambda, E_T(j_0^{(3)}) \gamma^{-1/2} \lambda^{-1} - \frac{\gamma^{-3/2}}{2} E_T \left(\frac{\partial g_0}{\partial \theta_0} \right) \right\} \\
&\quad \times (m, m_g, l_0^{(1)})' \\
&+ \left[-\frac{3}{8} \gamma^{-5/2} \lambda, \frac{\gamma^{-3/2}}{2}, -E_T(j_0^{(3)}) \gamma^{-1/2} \lambda^{-2}, -\frac{\gamma^{-3/2}}{2} \lambda^{-1} E_T(j_0^{(3)}) \right. \\
&\quad \left. + \frac{3}{4} \gamma^{-5/2} E_T \left(\frac{\partial g_0}{\partial \theta_0} \right), \gamma^{-1/2} \lambda^{-1}, \right. \\
&\quad \left. -\frac{\gamma^{-3/2}}{2}, \frac{\gamma^{-1/2}}{2} \lambda^{-3} \{E_T(j_0^{(3)})\}^2 + \frac{\gamma^{-3/2}}{4} \lambda^{-2} E_T(j_0^{(3)}) E_T \left(\frac{\partial g_0}{\partial \theta_0} \right) \right. \\
&\quad \left. - \frac{\gamma^{-1/2}}{2} \lambda^{-2} \{E_T(j_0^{(4)})\} + \frac{\gamma^{-3/2}}{4} \lambda^{-1} E_T \left(\frac{\partial^2 g_0}{\partial \theta_0^2} \right) - \frac{3}{8} \gamma^{-5/2} \lambda^{-1} \left\{ E_T \left(\frac{\partial g_0}{\partial \theta_0} \right) \right\}^2 \right] \\
&\quad \times \left[m_g^2, mm_g, ml_0^{(1)}, m_g l_0^{(1)}, E_T \{j_0^{(3)} - E_T(j_0^{(3)})\} l_0^{(1)}, \left\{ \frac{\partial g_0}{\partial \theta_0} - E_T \left(\frac{\partial g_0}{\partial \theta_0} \right) \right\} l_0^{(1)}, (l_0^{(1)})^2 \right]' \\
&+ O_p(n^{-3/2}) \\
&\equiv -\lambda \gamma^{-1/2} + \sum_{i=1}^2 \mathbf{r}^{(i)}' \mathbf{m}_{r0}^{(i)} + O_p(n^{-3/2}).
\end{aligned}$$

2. Supplementary numerical results

Table A gives the values of bias, skewness and kurtosis for t under model misspecification. Figures 1 and 2 plot the empirical and asymptotic distributions of t for $\hat{\theta}$ when $n=50$ and $\theta=0, 1, 2$ and t_h for $\exp(\hat{\theta})$ when $n=50, 100, 300, 800$ and $\theta=2$, respectively. In the plot of $\theta=2$ in Figure 1, some discrepancy between the normal approximation (dashed lines) and the empirical distribution (histogram) is observed. The discrepancy is more pronounced in Figure 2 when n is relatively small.

Table B shows the errors of the distribution functions approximated by the usual normal (N^*), the single- (E1) and two-term (E2) Edgeworth expansions (see Theorems 1 and 2), and Hall's (1992b) method of variable transformation, which is asymptotically as accurate as E1, under selected conditions with or without model misspecification. The true or empirical distributions were given

from the same simulations shown earlier. The values of t and t_r are taken at $-3.8(0.2)4.0$. It is found that E1, E2 and Hall have reduced the errors of N^* and that E2 on average has reduced the errors of E1 and Hall.

Reference

Ogasawara, H. (2012). Asymptotic expansions for the ability estimator in item response theory. *Computational Statistics*, 27, 661-683.

Table A. Simulated and asymptotic cumulants when the IRT model is false

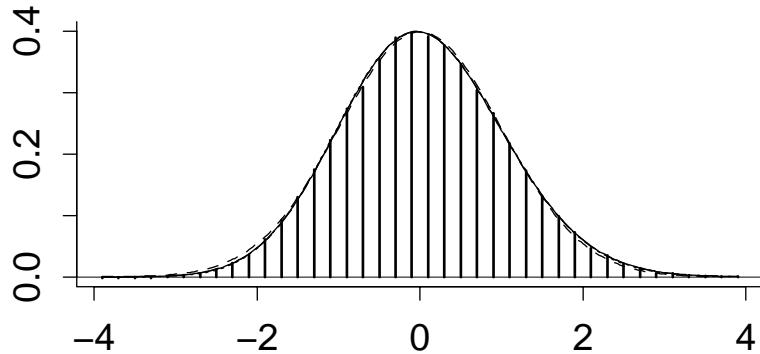
$n=300$		Slight misspecification				Gross misspecification			
		$\theta = -1$		$\theta = 2$		$\theta = -1$		$\theta = 2$	
t	$\hat{\theta}$	Sim.	Th.	Sim.	Th.	Sim.	Th.	Sim.	Th.
α_1	$\hat{\theta}$.58	.54	.02	.08	.53	.49	.03	.04
	P.C.	-.27	-.28	.94	.96	-.19	-.22	.83	.81
	logistic	-.02	-.05	.95	.98	.02	-.01	.84	.83
	$\exp(\hat{\theta})$	-.70	-.73	-1.17	-1.10	-.56	-.60	-1.01	-.98
α_3	$\hat{\theta}$	3.11	3.23	-1.71	-1.66	2.84	2.83	-1.47	-1.38
	P.C.	-1.73	-1.39	4.13	3.57	-.63	-.56	2.86	2.59
	logistic	-.21	-.07	4.06	3.69	.44	.41	2.83	2.68
	$\exp(\hat{\theta})$	-4.15	-3.90	-8.92	-8.67	-2.43	-2.40	-6.91	-6.72
α_4	$\hat{\theta}$	-3	-2	-9	-10	7	5	-11	-6
	P.C.	25	19	55	43	12	10	28	27
	logistic	9	8	41	34	6	6	19	21
	$\exp(\hat{\theta})$	35	30	109	99	13	13	64	70

Note. n =the number of items, Sim.=simulated value, Th.=theoretical or asymptotic value, P.C.=the estimator of the proportion-correct true score, logistic= $1/\{1+\exp(-\hat{\theta})\}$.

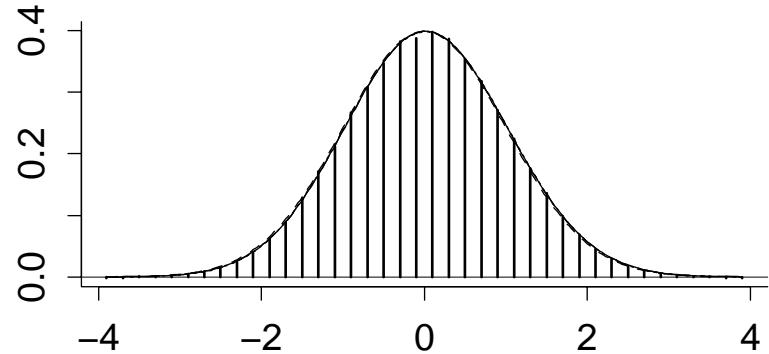
Table B. $10^5 \times$ root mean square errors of the asymptotic distribution functions for the pivots

		True model				Gross misspecification			
		$n=50, \theta=0$		$n=300, \theta=2$		$n=300, \theta=2$			
t	$\hat{\theta}$	N*	E1	E2	Hall	N*	E1	E2	Hall
t	$\hat{\theta}$	477	88	89	103	837	167	104	156
P.C.		250	222	97	221	820	179	103	166
logistic		553	332	111	323	826	149	104	142
$\exp(\hat{\theta})$		1901	366	145	297	1241	201	127	239
t_r	$\hat{\theta}$	530	384	161	385	259	109	103	109
P.C.		610	601	169	600	830	214	107	202
logistic		781	703	183	697	836	182	106	175
$\exp(\hat{\theta})$		2024	611	214	513	1230	182	114	221
						1143	138	105	199
						316	80	87	80
						760	172	89	171
						768	143	88	147
						869	204	89	203
						877	174	85	177
						978	113	95	166

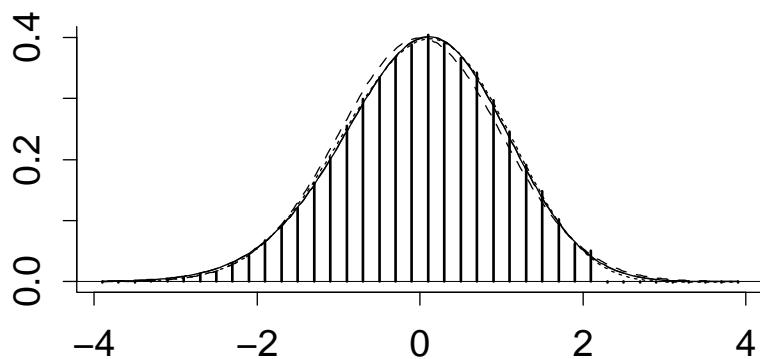
Note. n =the number of items, N*=normal approximation, E1=single-term Edgeworth expansion, E2=two-term Edgeworth expansion, Hall=Hall's method involving variable transformation, P.C.=the estimator of the proportion-correct true score, logistic= $1/\{1+\exp(-\hat{\theta})\}$.



population $\theta=0$, true model, $n=50$, t , θ hat

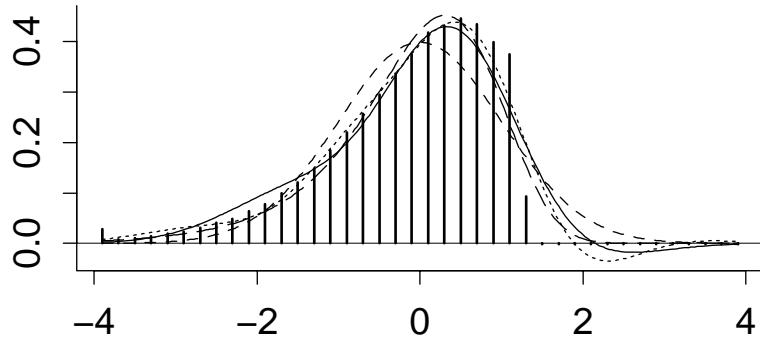


population $\theta=1$, true model, $n=50$, t , θ hat

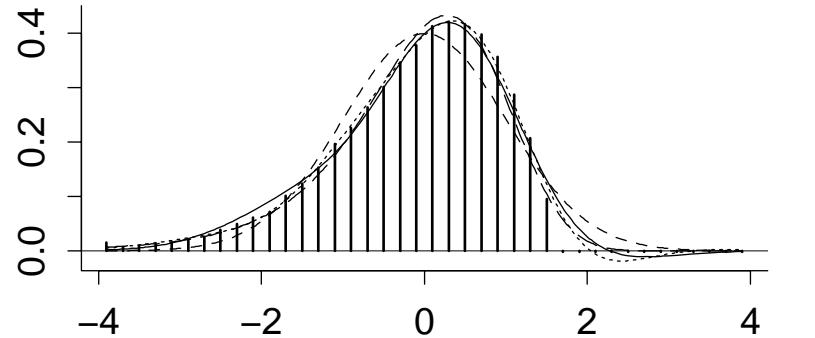


population $\theta=2$, true model, $n=50$, t , θ hat

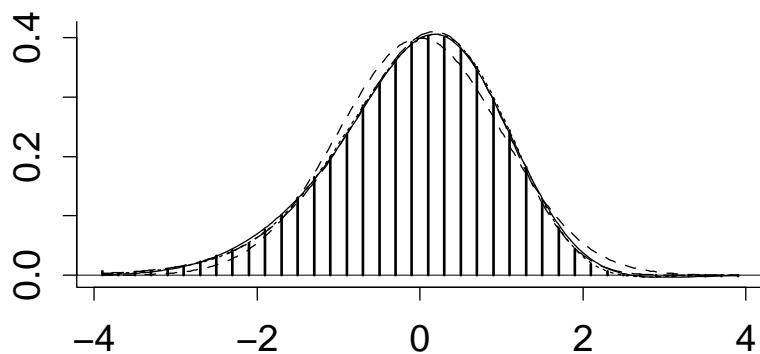
Figure 1. Theoretical (curved lines) and simulated (histograms) distributions of the pivot t (dashed lines=the normal distribution, solid lines=the single-term Edgeworth expansion, dotted lines=the two-term Edgeworth expansion, long dashed lines=Hall's variable transformation).



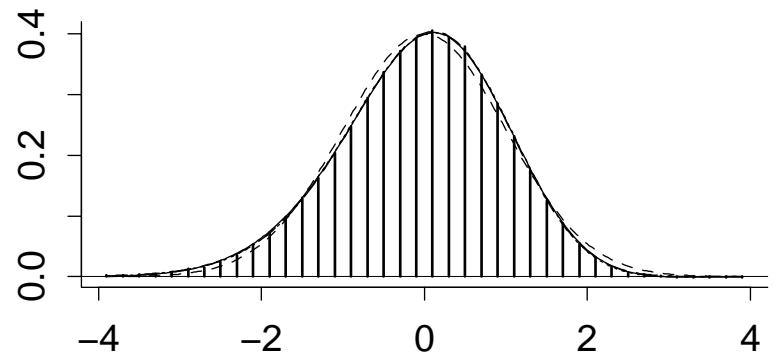
pop. theta=2, true model, n=50, t_h , $\exp(\theta)$



pop. theta=2, true model, n=100, t_h , $\exp(\theta)$



pop. theta=2, true model, n=300, t_h , $\exp(\theta)$



pop. theta=2, true model, n=800, t_h , $\exp(\theta)$

Figure 2. Theoretical (curved lines) and simulated (histograms) distributions of the pivot t_h (dashed lines=the normal distribution, solid lines=the single-term Edgeworth expansion, dotted lines=the two-term Edgeworth expansion, long dashed lines=Hall's variable transformation).