

**Supplement C: Data Generation with the 5<sup>th</sup> Order Power Polynomial**

The 5th-order power polynomial method (Headrick, 2002) was used to generate  $X$  and  $Y$  variables with specified correlations, skewness, and kurtosis.  $X$  and  $Y$  each consisted of polynomials constructed from standard normal variates ( $Z$  and  $Z'$ ):

$$X = c_0 + c_1Z + c_2Z^2 + \dots + c_5Z^5 \quad (S1)$$

$$Y = c'_0 + c'_1Z' + c'_2Z'^2 + \dots + c'_5Z'^5 . \quad (S2)$$

In order to assure a correlation between  $Z$  and  $Z'$ , and therefore a correlation between  $X$  and  $Y$ , the normal variate  $Z'$  was constructed as a linear combination of  $Z$  and an independently generated standard normal variate,  $Z_0$ .

$$Z' = tZ + \sqrt{1 - t^2}Z_0 , \quad (S3)$$

where  $t$  is an intermediate correlation coefficient which typically differs from the target population correlation,  $\rho$ .

To simulate bivariate normal data, the constants are  $c_i=c'_i=0$  for all  $i \neq 1$ ,  $c_1=c'_1=1$ , and  $t=\rho$ . These constants assure that both  $X$  and  $Y$  are normal, and more generally, that all linear combinations of them are also normal. To simulate other situations, constants must be estimated by solving a series of polynomial equations and unknowns based on the 3<sup>rd</sup> through 6<sup>th</sup> standardized moments and the desirable population correlation,  $\rho$ . First, as recommended in Headrick (2010), values for 5<sup>th</sup> and 6<sup>th</sup> standardized moments ( $\gamma_3$  and  $\gamma_4$ ) were estimated through 3<sup>rd</sup> order method Mathematica code. For situations with no skew ( $\gamma_1=0$ ),  $\gamma_3$  was already known to be 0 and required no estimation. In one case, a  $\gamma_4$  value had to be adjusted through trial and error to find the smallest approximate value with a stable solution (specifically, for  $\gamma_1=0$ ,  $\gamma_2=-1$ , a stable solution was found with  $\gamma_4=130$ ). Next, the target  $\gamma_1$  and  $\gamma_2$  and estimated  $\gamma_3$  and  $\gamma_4$  were used as input in the 5<sup>th</sup> order Mathematica code (Headrick, Sheng, & Hodis, 2007), which was used to estimate the constants shown in Table SC1.

Table SC1

*Constants For the 5th-Order Power Polynomial Data Generation Method.*

$\gamma_1$	$\gamma_2$	$\gamma_3$	$\gamma_4$	$c_0$	$c_1$	$c_2$	$c_3$	$c_4$	$c_5$	Intermediate Correlation ( $t$ ) for $\rho=.5$	
										When $X$ & $Y$ have shown moments	When $X$ has shown moments, and $Y \sim$ normal
0	-1	0	130.000	0	1.351811	0	-.171467	0	.00922413	.51919820	.51241520
0	0	0	0	0	1	0	0	0	0	.50000000	.50000000
0	2	0	24.290	0	.835664	0	.052057	0	-.00000001	.50612141	.50411522
0	4	0	98.219	0	.737382	0	.080925	0	.00000001	.51486784	.51012261
0	6	0	219.863	0	.662682	0	.101891	0	.00000000	.52367439	.51634007
0	8	0	386.838	0	.600653	0	.118712	0	.00000000	.53225522	.52258136
0	10	0	597.122	0	.546682	0	.132938	0	.00000000	.54056987	.52882240
0	20	0	2249.117	0	.338718	0	.184461	0	.00000000	.57857895	.56047454
0	30	0	4823.820	0	.179290	0	.220676	0	.00000000	.61184825	.59430653
0	40	0	8265.937	0	.041453	0	.249820	0	.00000000	.64137253	.63218184
1	2	5.584	20.677	-.147206	.904758	.147200	.023863	.00000182	-.00000041	.51211961	.51211603
1	4	15.109	99.796	-.116977	.776585	.116976	.065500	.00000017	-.00000003	.51658316	.51383046
1	6	23.821	224.238	-.102889	.690336	.102888	.091164	.00000010	-.00000002	.52423746	.51876519
1	8	31.952	392.944	-.094035	.622114	.094036	.110454	-.00000013	.00000002	.53232730	.52439761
1	10	39.652	604.397	-.087718	.564261	.087718	.126209	.00000002	.00000000	.54038671	.53028549
1	20	74.162	2258.952	-.070726	.348061	.070726	.180925	.00000001	.00000000	.57803596	.56126952
1	30	104.930	4834.381	-.062431	.185712	.062431	.218254	-.00000003	.00000000	.61127901	.59490252
1	40	133.764	8276.673	-.057246	.046374	.057246	.247965	.00000002	.00000000	.64083621	.63269574

2	8	44.468	333.629	-.233358	.710436	.233353	.072266	.00000154	-.00000036	.53904439	.53924128
2	10	62.223	565.094	-.203447	.630443	.203445	.098743	.00000044	-.00000011	.54294195	.53956621
2	20	136.665	2256.827	-.149826	.378622	.149826	.168920	.00000003	.00000000	.57685505	.56472832
2	30	200.482	4843.997	-.129343	.205931	.129343	.210419	.00000002	.00000000	.60972212	.59723732
2	40	259.463	8291.717	-.117429	.061593	.117429	.242096	-.00000003	.00000000	.63929429	.63461319
3	20	170.420	2095.789	-.260482	.442217	.260482	.140478	-.00000009	.00000002	.57865945	.57893735
3	30	274.943	4767.762	-.208933	.243540	.208933	.194735	.00000005	-.00000001	.60800770	.60405018
3	40	367.694	8249.858	-.185083	.088675	.185083	.231041	-.00000001	.00000000	.63706984	.63955121
4	30	306.683	4318.326	-.334667	.307703	.334668	.160288	-.00000022	.00000005	.61218548	.63405982
4	40	444.793	7995.129	-.271450	.130891	.271450	.211277	-.00000002	.00000000	.63559382	.65383172

*Notes.*  $\gamma_1$ =population skewness,  $\gamma_2$ =population kurtosis. For distributions with negative skewness (not shown here), constants for positive skewness were used except that the following values were multiplied by -1:  $c_1$ ,  $c_3$ , and  $c_5$ . Intermediate correlations ( $t$ ) for  $\rho=0$  were all 0. Constants for  $Y$  variables were equal to those of  $X$  when they had the same distribution shape ( $c'_i = c_i$ ). When  $Y$  had a normal distribution,  $c'_1=1$ , and  $c'_i=0$  for all  $i \neq 1$ .