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dm 'log;clear;output;clear';
options ps=45 ls=105 nocenter nodate nonumber nolabel
      FORMCHAR="|----|+|----+=|-\<>*" ;
ODS listing;
ods graphics off;
ods html close;
ODS HTML style=minimal body='rsquare demo - high variance.html';

title1 'Monte carlo study to examine the effect of variance on RSquare values
';
data demo;
  seed = 3394903;
  do nobs = 1000, 5000;          * Generate two sample sizes;
  do stddev = 10, 50, 100, 500, 1000, 1500, 2000, 2500, 3000;
                                * Generate an increasing series of
variances;
  do obs = 1 to nobs by 1;      * Loop to generate observations;
    xvar = INT(ranuni(seed)*100)+1; * Randomly generate X variable
(uniform);
    Y = 10000 + 25 * xvar;      *Generate regression b0=10000, b1=25;
    Yvar = Y + rannor(seed)*stddev; *Add randomly distributed variation;
    output;
  end; end; end;
run;

proc sort data=demo; by nobs stddev; run;
proc glm data=demo; by nobs stddev;
  title2 'Fitting the generated data to a regression';
  title3 'GLM was used because the rsquare is easy to output';
  model yvar = xvar;
  ods output ParameterEstimates=est FitStatistics=fits;
run;

* Parmest dataset contains estimates and p values;
* The following gets the values from two lines to one;
data parmest; set est;
  format b0p b1p 7.5;
  if parameter ne 'xvar' then do;
    b0 = (estimate);
    b0p = fuzz(probt);
  end;
  if parameter eq 'xvar' then do;
    b1 = estimate;
    b1p = fuzz(probt);
  end;
  retain b0 b0p;
  if b1 eq . then delete;
  keep nobs stddev b0 b0p b1 b1p;
run;

* combine parmest dataset with fits dataset that contains rsquare values;
data combo; merge parmest fits; by nobs stddev;
  FitTo = 'Raw Data';
  keep nobs stddev b0 b0p b1 b1p rsquare rootMSE FitTo;
run;

*Calculate means and repeat analysis;

proc sort data=demo; by nobs stddev XVAR; run;

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```
proc means data=demo noprint; by nobs stddev XVAR;
  var yvar;
  output out=next02 n=n mean=YMean;
run;

proc sort data=next02; by nobs stddev; run;
proc glm data=next02; by nobs stddev; weight n;
  title2 'Fitting the generated data to a regression';
  title3 'GLM was used because the rsquare is easy to output';
  model YMean = XVAR;
  ods output ParameterEstimates=est2 FitStatistics=fits2;
run;

* Parmest dataset contains estimates and p values;
data parmest2; set est2;
  format b0p blp 7.5;
  if parameter ne 'xvar' then do;
    b0 = (estimate);
    b0p = fuzz(probt);
  end;
  if parameter eq 'xvar' then do;
    b1 = estimate;
    blp = fuzz(probt);
  end;
  retain b0 b0p;
  if b1 eq . then delete;
  keep nobs stddev b0 b0p b1 blp;
run;

* combine parmest dataset with fits dataset that contains rsquare values;
data combo2; merge parmest2 fits2; by nobs stddev;
  FitTo = 'Means @X';
  keep nobs stddev b0 b0p b1 blp rsquare rootMSE FitTo;
run;

data all; set combo combo2; run;
proc sort data=all; BY nobs descending FitTo stddev; run;

proc print data=all;
  title2 'Results to all fits of the generated data';
  var FitTo stddev nobs b0 b0p b1 blp rootMSE rsquare;
run;

ods html close;
quit;
```

Monte Carlo generated data to examine the effect of variance on RSquare values

Results to all fits of the generated data for the model “ $Y = 10000 + 25 * \text{xvar}$ ” where xvar values are randomly generated integers between 1 and 100. In most Monte Carlo studies numerous replicates would have been generated to observe patterns and consistency. Here I included only a single sample for demonstration purposes.

Obs	FitTo	stddev	nobs	b ₀	b ₀ P=	b ₁	b ₁ P=	RootMSE	RSquare
1	Raw Data	10	1000	10000.19	0.00000	24.9973	0.00000	10.057	0.999809
2	Raw Data	50	1000	10000.87	0.00000	25.0383	0.00000	50.759	0.994977
3	Raw Data	100	1000	10010.90	0.00000	24.7832	0.00000	99.096	0.981575
4	Raw Data	500	1000	10046.82	0.00000	24.8381	0.00000	478.202	0.704850
5	Raw Data	1000	1000	9998.72	0.00000	24.5740	0.00000	989.588	0.341024
6	Raw Data	1500	1000	9986.95	0.00000	24.8846	0.00000	1481.111	0.192970
7	Raw Data	2000	1000	9796.84	0.00000	27.5388	0.00000	2026.934	0.130349
8	Raw Data	2500	1000	9812.89	0.00000	29.3624	0.00000	2515.937	0.100578
9	Raw Data	3000	1000	10074.13	0.00000	26.2960	0.00000	2916.948	0.064187
10	Means @X	10	1000	10000.19	0.00000	24.9973	0.00000	9.897	0.999982
11	Means @X	50	1000	10000.87	0.00000	25.0383	0.00000	52.199	0.999476
12	Means @X	100	1000	10010.90	0.00000	24.7832	0.00000	95.688	0.998284
13	Means @X	500	1000	10046.82	0.00000	24.8381	0.00000	550.288	0.948361
14	Means @X	1000	1000	9998.72	0.00000	24.5740	0.00000	968.722	0.846144
15	Means @X	1500	1000	9986.95	0.00000	24.8846	0.00000	1519.915	0.698094
16	Means @X	2000	1000	9796.84	0.00000	27.5388	0.00000	2095.414	0.588182
17	Means @X	2500	1000	9812.89	0.00000	29.3624	0.00000	2372.036	0.561625
18	Means @X	3000	1000	10074.13	0.00000	26.2960	0.00000	2735.743	0.442616
19	Raw Data	10	5000	10000.24	0.00000	24.9970	0.00000	9.822	0.999814
20	Raw Data	50	5000	9999.77	0.00000	25.0096	0.00000	50.013	0.995213
21	Raw Data	100	5000	9997.99	0.00000	25.0384	0.00000	99.637	0.981405
22	Raw Data	500	5000	10005.02	0.00000	24.9416	0.00000	506.177	0.673171
23	Raw Data	1000	5000	9995.89	0.00000	24.6553	0.00000	976.386	0.352368
24	Raw Data	1500	5000	9995.71	0.00000	25.6092	0.00000	1508.451	0.194871
25	Raw Data	2000	5000	9966.97	0.00000	26.4152	0.00000	1987.082	0.127592
26	Raw Data	2500	5000	10043.72	0.00000	25.1749	0.00000	2523.095	0.076964
27	Raw Data	3000	5000	9997.24	0.00000	25.0297	0.00000	2950.834	0.055928
28	Means @X	10	5000	10000.24	0.00000	24.9970	0.00000	9.729	0.999996
29	Means @X	50	5000	9999.77	0.00000	25.0096	0.00000	53.157	0.999893
30	Means @X	100	5000	9997.99	0.00000	25.0384	0.00000	96.387	0.999652
31	Means @X	500	5000	10005.02	0.00000	24.9416	0.00000	478.122	0.991578
32	Means @X	1000	5000	9995.89	0.00000	24.6553	0.00000	962.175	0.966187
33	Means @X	1500	5000	9995.71	0.00000	25.6092	0.00000	1696.120	0.907092
34	Means @X	2000	5000	9966.97	0.00000	26.4152	0.00000	2326.593	0.844741
35	Means @X	2500	5000	10043.72	0.00000	25.1749	0.00000	2614.196	0.798438
36	Means @X	3000	5000	9997.24	0.00000	25.0297	0.00000	2959.723	0.750198