# An Empirical Comparison of within-Subjects and between-Subjects Expectancy Theory Models

CHARLES W. KENNEDY, JOHN A. FOSSUM, AND BERNARD J. WHITE

University of Michigan

One of the most consistent criticisms of expectancy theory research for the prediction of effort is that it has not been tested using the within-subjects choice model that the theory requires. The choice model proposed by the theory generates a motivational force score (MFS) for each of several effort levels for each subject and predicts that each will choose the level of effort which has the highest MFS for that subject. This study operationalizes the choice model and compares it with a difference model (derived from subtracting the MFS for low effort from the MFS for high effort) and a singlealternative model (MFS for high effort only). Subjects were 74 undergraduates who estimated valences and expectancies for the outcomes of six general student activities. The difference model, not the choice model, was generally the best predictor, while the single-alternative model predicted least well. Within-subjects predictions were of greater magnitude than between-subjects predictions. The results indicated that sufficiently designed future studies should gather data for three levels of effort (high, medium, and low) and then compare the models to determine the most effective predictor for that particular situation.

While expectancy theory is the most widely accepted theory of motivation in contemporary industrial/organizational psychology, its empirical support has not been particularly strong. One reason could be that studies have used between- rather than within-subjects designs that are not consistent with the intended theoretical formulation (Mitchell, 1974). A within-subjects approach compares a given subject's attitudes with other attitudes of that same subject, while a between-subjects approach compares one subject's attitudes with attitudes of other subjects for similar outcomes. Unfortunately, there has been no consistent approach to the construction of expectancy theory research. Most of the arguments given for using within-subjects research have referred to how the motivational force scores are measured and computed. Three alternative methods for measuring motivational force have or could be used in both within- and between-subjects studies.

This paper is based on a dissertation submitted to the University of Michigan by the first author. Portions of an earlier version were presented at the Academy of Management Meeting, August 1980. Requests for reprints should be sent to John Fossum, who is now with the Industrial Relations Center, University of Minnesota, 271 19th Ave. So., Minneapolis, MN 55455. Bernard J. White is now with the Cummins Engine Company, Columbus, IN.

# Models for Computing Motivational Force Scores

The *single-alternative* model computes a motivational force score (MFS) for each respondent by multiplying each outcome valence by the expectancy of its attainment given the *single-alternative* "high effort" and then summing the products across outcomes. Persons with higher MFS's are predicted to put forth higher effort.

The difference model obtains the same information as the single-alernative model and, in addition, the respondents are asked their expectancies of receiving those same outcomes if they put forth low effort. The MFS for low effort is then subtracted from the MFS for high effort and persons with larger differences are predicted to put forth higher effort.

The choice model obtains MFS's for several effort levels. Each respondent is predicted to choose that effort level which has the highest MFS. This approach is consistent with Vroom's (1964) original formulation.

### The Choice Model as the Model of Choice

Several reasons support the use of the choice model in expectancy theory research.

- (1) Theoretical accuracy. Vroom (1964) clearly intended a choice model, as shown in his following example. One individual has high-expected payoffs for both high and low effort, the second has high-expected payoffs for only high effort, and the third has high-expected payoffs for neither high nor low effort. The second person is the only one predicted to choose high effort, because only he benefits from putting forth high effort. Choice and single-alternative (high effort) model predictions would be contradictory, since the single-alternative model would have had the first person also choosing high effort.
- (2) Ipsative orientation. A person's score for one level of effort can only be meaningful when compared to that same person's scores for other levels of effort (Dachler & Mobley, 1973; Kopelman, 1977; and Mobley & Meglino, 1977).
- (3) Varied relationships between effort level and motivational force score. Testing for only a single level of effort assumes that there is a similar pattern between effort and MFS for all levels of effort (Mitchell & Pollard, 1973). In fact there could be a nonlinear or nonmonotonic relationship.
- (4) Amount of information. The choice model research method provides more information for changes in employment situations, because it obtains information about several levels of effort (Nebeker & Mitchell, 1974).

The difference model could be considered along with the choice model as fulfilling some of the above needs, but the difference model has po-

tential problems of lower reliability and spurious correlation (Cohen & Cohen, 1975; Cronbach & Furby, 1970).

Between- and within-Subjects Approaches

Motivational force scores (MFS's) can either be compared for effort levels between subjects for a given activity or across activities within a given subject. It is clear from Vroom's (1964) formulation of expectancy theory that the force to perform an act requires a comparison of the probabilities of obtaining certain outcomes from that act and the probabilities of obtaining the outcomes from other acts before a choice is made. The individual is not seen as comparing outcomes to acts with others.

Thus a within-subjects approach is necessary to test adequately the theory as it was formulated. A within-subjects approach also mitigates the following possible difficulties.

- (1) Response bias. Individual response biases are eliminated.
- (2) Between-individual variation. Individual differences on aspects such as ability, wealth, opportunity costs, personality traits, and total energy available which introduce potential measurement error, are eliminated (Kopelman, 1977).
- (3) Lack of ratio scale. Schmidt (1973) and Arnold and Evans (1979) pointed out that expectancy and valence measures lack a rational zero point, which means they are not true ratio scales, giving doubt as to whether it is logically meaningful to multiply them. Mitchell (1974), while recognizing the validity of the argument, noted that to the extent that the variances from the true score are constant, "then a within-subjects analysis will again help to remedy the problem" (p. 1067).

This study compares predictions made by the single-alternative, difference, and choice models in both within- and between-subjects modes. It tests Vroom's (1964) formulation of expectancy theory as previously called for by Behling and Starke (1973), Campbell and Pritchard (1976), Connolly (1976), Hollenback (1979), House, Shapiro, and Wahba (1974), Kopelman (1977), Korman, Greenhause, and Badin (1977), Mitchell (1974), Mitchell and Beach (1979), and Muchinsky (1977).

It is hypothesized that (1) correlations between choice model predictions and effort will be of greater magnitude than correlations between difference model predictions and effort, which in turn will be of greater magnitude than correlations between single-alternative model predictions and effort; and (2) within-subjects correlations between the various model formulations and effort will be of greater magnitude than the corresponding between-subjects correlations.

#### **METHOD**

### Sample

This study was part of a time allocation feedback study conducted among dormitory residents at a nondenominational Christian university.

The time allocation feedback study was designed to run for 10 weeks. Of 180 potential participants, 95 volunteered to record each day how much of their discretionary time they had spent in six different activities. At the beginning of each week the records for the previous week were collected and individual computer-generated feedback sheets were given to students summarizing how they had spent their time and how the study group as a whole had spent its time. Of the 95 volunteers, 74 who had been involved in the study for at least 2 weeks completed an expectancy theory questionnaire at the end of the semester, after some experience had been gained about effort effects and outcomes.

#### Data Collection

The questionnaire was administered in small groups. Each subject was given a copy of his/her semester's time allocations along with the questionnaire so that they would not have to rely on memory. Participation in the time allocation feedback study and completion of the questionnaire was entirely voluntary and all results were kept strictly confidential. A small monetary donation was made for participation to each participating student's dormitory unit.

As a result of feedback the students involved in the study knew how they and others had spent their discretionary time in the six activities: studies, athletics, social activities, spiritual activities, service toward others, and leisure. These activities were defined to the students to be mutually exclusive. Athletics was included as an activity because all students were required to take physical education classes each term, earn 30 aerobic athletic points per week, and run 1½ miles under 12 min. to qualify for graduation. Spiritual and service towards others activities were very actively engaged in by students in this Christian university.

Asking the respondents to complete a questionnaire after they knew how they said they had spent their time raises the possibility that the respondents might have attempted to answer the questions in such a way as to be consistent with their actual behavior. However, their experiences may have made them better estimators of the actual effort levels necessary to achieve valent outcomes.

#### Measures

Effort. Effort was operationally defined as the amount of discretionary time spent in that activity. Vroom (1964) first suggested the use of hours to measure effort, and many research studies have since used time spent as an effort measure (Kopelman, 1977; Kopelman & Thompson, 1976; Mitchell & Nebeker, 1973; Mitchell & Pollard, 1973; Peters, 1977). Subjects were asked to observe their behavior and record it, while in most past studies subjects were asked to recollect how much time they spent in certain activities, without consciously observing themselves before-

hand. Subjects also received feedback about how they and others spent their time, meaning that they could more intelligently answer time allocation questions and that the effort measure more closely approximated a "concrete observable" (Campbell & Pritchard, 1976). Reported effort was obtained by asking students, "How many hours per day on the average do you estimate that you actually spend on each of the general activities?" The concept of preferred effort was introduced by Turney (1974; Turney & Cohen, 1976), who referred to it as desired effort. To measure preferred effort, students were asked, "How many hours per day on the average do you estimate that you would reasonably like to spend on each of the six general activities?"

Outcomes. Keeney and Raiffa (1976) suggest that outcomes should meet the following criteria: (a) completeness, (b) operationality, (c) nonredundancy, and (d) parsimony. The outcomes chosen were judged as best meeting those criteria, and were taken from Constantinople (1967) and Mitchell and Nebeker (1973), as well as from careful discussion and personal experience with the students. The 10 outcomes finally selected were (1) Getting accepted into graduate school, (2) Starting a good career, (3) Obtaining a feeling of accomplishment, (4) Enjoying whatever activity you are doing, (5) Having good friends of the same sex, (6) Having good friends of the opposite sex, (7) Having a good relationship with God, (8) Being in good shape physically, (9) Being a significant help to others, and (10) Obtaining a scholarship.

Valence. Valence was measured by asking respondents to state how desirable each of the outcomes were to them, assuming that they could achieve any possibility that they attempted. Anchors ranged from extremely desirable to extremely undesirable and were scored from +3 to -3.

Expectancy. Expectancy was interpreted to be a conditional probability, which is both common and theoretically justified (Campbell & Pritchard, 1976). Expectancies were conditional in the sense that students were asked to state the perceived probability of attaining a certain outcome, given that they put forth a specific amount of effort.

Instrumentality and expectancy were combined, because most of the activities had no clearly defined separation between first- and second-level outcomes. Omitting the instrumentality measure is common (e.g., Lawler & Porter, 1967; Lawler & Suttle, 1973) as well as theoretically justified in situations where clear intermediate outcomes are lacking (Mitchell & Pollard, 1973), such as with the athletic, social, spiritual, service, and leisure activities. An example of the expectancies requested from subjects for just one level of effort (1 hr) for one activity (studying) follows. Since there were five levels of effort for six activities, this question format was asked of each student 30 times.

If you spend an average of *one* hour studying per day, what is the probability that this will result in . . . Getting accepted into graduate school—

```
0. .1 .2 .3 .4 .5 .6 .7 .8 .9 1.

Starting a good career—

0. .1 .2 .3 .4 .5 .6 .7 .8 .9 1.

Obtaining a feeling of accomplishment—

0. .1 .2 .3 .4 .5 .6 .7 .8 .9 1.

Enjoying what you are doing—

0. .1 .2 .3 .4 .5 .6 .7 .8 .9 1.

Obtaining a scholarship—

0. .1 .2 .3 .4 .5 .6 .7 .8 .9 1.
```

The expectancy question used specific hour levels instead of vague terms, such as "high effort," meeting the need for a "specific referent" (Dachler & Mobley, 1973). Subjects were asked to rate expectancies for only some of the 10 outcomes for every activity (e.g., subjects were not asked to rate the likelihood that 1 hr of studying would result in being in good shape physically). What might be lost in terms of accuracy is compensated for by the significant reduction in the length of the questionnaire. In addition, Connolly and Vines (1977), Leon (1979), and Matsui and Ikeda (1976) have found that relatively few outcomes explain most of the variance.

## Computations

For the choice model, a motivational force score was computed for each of the five hour levels by multiplying the expectancies for each hour level with the valences of their respective outcomes and then summing the products. The numerical value of the choice model predictor variable was that hour level (1–5) which had the highest computed motivational force score, with a tie going to the lower hour level, because it was assumed that, other things being equal, people seek to avoid the expenditure of effort (Vroom, 1964, p. 195). For the difference model, the numerical value was obtained by subtracting the computed motivational force score for low effort from the computed motivational force score for high effort. (Low effort was assumed to be 1 hr and high effort was assumed to be 5 hr.) For the single-alternative model, the numerical value was simply the motivational force score for high effort (assumed to be 5 hr).

Once numerical values were obtained a within-subjects correlation and a between-subjects correlation could be computed for each model. To obtain a within-subjects correlation, for each subject the computed force scores for all six activities were correlated with their respective effort measures. This resulted in 74 correlations (one correlation for each subject), each correlation having an n of 6. These 74 correlations were av-

eraged to obtain a within-subjects average correlation. To obtain a between-subjects correlation, for each activity the computed force scores for all 74 subjects were correlated with their respective effort measures. This resulted in six correlations (one correlation for each activity), each with an n of 74. These 74 correlations were averaged to obtain a between-subjects average correlation.

# Data Analysis

The hypotheses were tested by comparing average correlations. The test for Hypothesis One compared all four average correlations for the choice model to predict preferred effort and reported effort (two withinsubjects correlations and two between-subjects correlations) against their four respective counterparts using the difference model and the single-alternative model. The test for Hypothesis Two compared all six withinsubjects average correlations using the three models to predict the two dependent measures of effort against their respective between-subjects correlations.

Average correlations were tested to determine if the differences between the correlations were statistically significant. Since the average correlations were taken from the same sample, the Hotelling test for comparing dependent correlations was used (Cohen & Cohen, 1975). For the comparisons between the within-subjects and the between-subjects correlation methods (Hypothesis 2) the more common comparison test using r to z transformations was used, because some of the intercorrelations required for the Hotelling test could not be computed.

### **RESULTS**

### Independent and Dependent Measures

The motivational force scores listed in Table 1 and graphed in Fig. 1 show the differences between the MFS-effort relationships for the various activities. As an activity, studying demonstrates a highly positive and nearly linear relationship between effort and motivational force scores, while spiritual, social, and service activities have monotonically de-

<sup>&</sup>lt;sup>1</sup> The process of computing the average of a set of correlations is clearly described by Snedecor and Cochran (1967, p. 186). The first step is to run a special chi-square test of the hypothesis that the set of correlations are from the same population. If the result of the chi-square test permits the assumption of homogeneity, then the correlations can be averaged. This averaging is an iterative process and uses the r to z transformation as well as specified changes in the degrees of freedom. The resulting average correlation has fewer df than the sum of the n's of the correlations.

<sup>&</sup>lt;sup>2</sup> This test is not to be confused with the Hotelling T test.

Effort	Activity						
level	Study	Athletic	Social	Spiritual	Service	Leisure	
One hour							
Mean	2.13	5.17	3.92	8.14	7.10	5.03	
SD	1.85	2.01	1.96	3.36	2.56	2.20	
Two hours							
Mean	3.59	5.84	5.08	9.89	8.18	5.32	
SD	2.30	1.93	2.06	3.21	2.23	2.17	
Three hrs.							
Mean	6.10	5.83	5.76	10.62	8.73	5.21	
SD	2.76	1.94	2.01	3.02	2.02	2.09	
Four hours							
Mean	7.85	5.87	5.96	10.65	8.95	5.11	
SD	2.82	2.05	2.13	3.02	1.96	2.26	
Five hours							
Mean	8.75	5.84	6.15	10.56	9.16	4.95	
SD	3.04	2.25	2.33	3.43	2.29	2.53	

TABLE 1 MOTIVATIONAL FORCE SCORES FOR EACH EFFORT LEVEL BY ACTIVITY<sup>a</sup>

creasing relationships. Athletic and leisure activities have essentially flat relationships over the effort levels measured. The shape of the MFS-effort relationship is important, since the single-alternative model would be inadequate if the MFS-effort relationship is not positive and linear.

Examining the within-subjects analysis, Fig. 1 shows that marginal benefits seem potentially important, since for studying there appears to be substantial additional expected gains during the 2nd, 3rd, 4th, and 5th hr. For service, spiritual, and social activities there appears to be additional expected gain only for the 2nd and 3rd hr, for athletic activities only for the 2nd hr, and for leisure activities no additional expected gain past the 1st hr. Since the difference model takes into account additional (or marginal) gains, the difference model would predict that studying would receive the most effort and that leisure would receive the least. In contrast, the single-alternative model looks only at the relative rankings for the single-effort level of 5 hr, and therefore would predict that the most effort will go toward spiritual activities.

Table 2 presents the means and standard deviations for the three model predictions (choice, difference, and single-alternative models) and the two effort measures (preferred effort, and reported effort). Preferred effort was how much time the subject would reasonably like to spend in each of the activities. Reported effort was the time students said they actually spent in the various activities.

a n = 74.

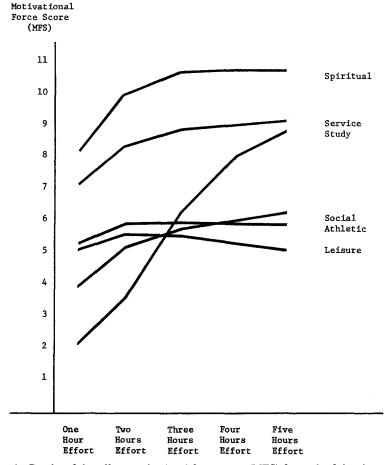


FIG. 1. Graphs of the effort-motivational force score (MFS) for each of the six general activities.

The choice model seemed to overpredict. For example, the choice model predicted nearly  $4\frac{1}{2}$  hr (4.46) studying per day, while reported effort is only a little over 3 hr (3.15) per day. Even more disparate, the choice model predicted 3 hr (3.01) in athletics per day, while reported effort is only  $2\frac{1}{3}$  hr (.67) per day.

It appears that the single-alternative model using the within-subjects analysis is inadequate, since it predicts that each individual will spend the most time in the activity with the highest MFS. This has clearly not happened, in that spiritual activity has the highest MFS for 5 hr (MFS = 10.56), but studying has the highest reported effort (3.16 hr). The primary problem in prediction is between-activity variance. Some activ-

TABLE 2
MEANS AND STANDARD DEVIATIONS FOR MODEL PREDICTIONS AND EFFORT MEASURES <sup>a</sup>

	Activity						
	Study	Athletic	Social	Spiritual	Service	Leisure	
		Mo	odel predictio	ons			
Choice model			-				
Mean	4.61	3.01	3.86	3.47	3.68	2.76	
SD	0.68	1.47	1.29	1.30	1.40	1.59	
Difference model							
Mean	6.62	0.67	2.23	2.42	2.06	08	
SD	3.31	2.55	2.15	4.15	3.01	2.73	
Single-alt. Model							
Mean	8.75	5.84	6.15	10.56	9.16	4.95	
SD	3.04	2.25	2.33	3.43	2.29	2.53	
		E	ffort measure	es			
Preferred effort							
Mean	3.90	1.10	2.29	1.61	1.46	1.40	
SD	1.00	0.55	0.95	0.73	0.70	0.87	
Reported effort							
Mean	3.16	0.67	2.27	1.05	0.86	1.46	
SD	1.34	0.39	1.01	0.76	0.66	1.30	

a n = 74.

ities simply result in a greater number of outcomes than others. However, the choice and difference model are not as affected by between-activity variance, because their measures are determined by referring to other levels of the same activity.

# Average Correlations

Between-subjects correlations. The average between-subjects correlations and the individual correlations are shown in Table 3. For studying, the individual correlation between the choice model measure and preferred effort was .36 (p < .01). The lowest correlation was -.03, for the single-alternative model's correlation with reported effort spent studying.

Comparing the correlations in more detail, it is apparent that, when limited to studying, the single-alternative model was a poor predictor of preferred and reported effort (r = -.01 and -.03), and yet was a reasonable predictor for activities such as athletics (r = .30 and .18). This is unexpected, because, as discussed earlier, studying had a positive linear MFS-effort relationship, while athletics had an essentially flat MFS-effort relationship over the effort levels measured. Previous studies (Mitchell & Pollard, 1973; Nebeker & Mitchell, 1974) would have suggested that the positive, linear MFS-effort relationship for studying would have yielded higher correlations for the single-alternative model.

BETWEEN-SUBJECTS CORRELATIONS BETWEEN MODEL PREDICTIONS AND EFFORT MEASURES TABLE 3

Predictor			Act	Activity			Simple
model	Study	Athletic	Social	Spiritual	Service	Leisure	avg r
			Preferred effort	ırt			
Choice	.36**	**54.	.32**		.28*	.21	.31***
Difference	.14	.41**	.33**	*67.	.25*	.26*	.28***
Single-alternative	01	.30**	.22	.31**	.27*	.13	.20***
			Reported effort	ort			
Choice	.21	.27*	.17		.18	.17	.21***
Difference	.14	.26*	.21	.31**	.14	.17	.21***
Single-alternative	03	.18	.05	.22	90:	.02	80.

Note. All chi-squares testing if correlations were from the same population were not significant (p < .25, df = 5). \* p < .05 (df = 71). \*\* p < .01 (df = 71). \*\* p < .01 (df = 71). \*\*\* p < .001 (df = 426).

The average between-subjects correlations are also listed in Table 3 and were derived by averaging the correlations of the six activities.<sup>3</sup> All the average between-subjects correlations were significant at the .01 level, except for the single-alternative model's prediction of reported effort (avg r = .08).

Within-subjects correlations. Turning now to within-subjects correlations, Table 4 shows the six average within-subjects correlations.<sup>4</sup> The method used to average correlations was to transform each individual correlation to a corresponding Fisher's z, average the z's, and then transform the average z to obtain an average r.<sup>5</sup> The z transformation had the effect of weighting the extreme correlations more heavily, and this explains why the average z's in the fourth column of Table 4 were signifi-

<sup>3</sup> The statistically correct method of averaging correlations (Snedecor & Cochran, 1967) was not used for the between-subjects data, because the requirement that the correlations be from different samples of subjects was not met. However, because the individual correlations had large sample sizes (n = 74), the simple average correlations listed in Table 3 are virtually identical to what would have been the statistically correct average correlations. Because the average correlations were a pooling of individual correlations, their degrees of freedom were larger (df = 426), though it must be remembered that the assumption of independence was not met.

 $^4$  The standard deviations of the sample average correlations were large because the n of each individual correlation was only 6, there being only six activities to correlate for each individual. Not all of the average correlations had an n of 74, since for some individuals the predictor value was the same for all six activities and it was not possible to compute a correlation when one of the variables was a constant. This was only a problem with the choice model, since for 5 respondents the choice model predicted that they would spend five hours in all six activities.

 $^{5}$  This transformation was necessary because repeated correlations taken from the same sample do not form a normal distribution, unless the population correlation is zero. For example, if a population correlation were .8, repeated sample correlations would form a distribution that is skewed toward 0.0. But if the correlations were transformed to Fisher's z's, then the distribution would become normal. The impact that this had when averaging the correlations was that the positive correlations were made more positive (an r of .99 becomes a z of 2.647), the correlations near zero remained unchanged, and the negative correlations were made more negative.

Snedecor and Cochran (1967) point out that there is a small bias in each z. They recommend that if the individual correlations have small n's a correlation factor be subtracted from each z before it is averaged with the others. With the within-subjects correlations listed in Table 4, subtracting the correction factor had the effect of reducing somewhat the increase that resulted from the z transformations, but even the corrected average correlations (listed in the last column) were somewhat higher than their respective simple average correlation. For example, the simple average correlation between the choice model and preferred effort was .53 (column 1), the z transformation then increased the average correlation to .75, and subtracting the correction factor reduced the average correlation to .60 (last column). So when dealing with a large number of correlations, with each correlation having a small sample size, both the z transformation and the correction factor can have a noticeable effect on the size of the average correlation.

Predictor model	Simple avg r	SD	n	Avg z	Chi square	Corrected avg r
		Preferr	ed effor	t		
Choice	.53	.35	69	.75	84.08	.60***
Difference	.68	.29	74	1.02	68.33	.74***
Single-alternative	.26	.41	74	.34	70.60	.30***
		Reporte	ed effor	t		
Choice	.39	.43	69	.52	107.63**	.44***
Difference	.48	.41	74	.68	86.81	.55***
Single-alternative	.05	.43	74	.07	62.10	.07

TABLE 4
WITHIN-SUBJECTS CORRELATIONS BETWEEN MODEL PREDICTIONS AND EFFORT MEASURES

cantly higher than their respective simple average r's in column 1. For example, the average z of the choice model's predictions of preferred effort (row 1) was .75, and was larger than the simple average correlation of .53.

The first of two tests of statistical significance to be run on the withinsubjects average correlations was the chi-square test of the assumption that the individual correlations came from the same population. Only the chi-square test of the choice model's prediction of reported effort is significant at the .01 level, which means averaging that correlation is not statistically appropriate. However, for purposes of comparison, the choice model–reported effort average correlation is still used in the hypothesis tests, but it is with the understanding that the homogeneity assumption has not been met. The second statistical significance test concerned the probability that the average correlation was significantly different from zero, and the last column of Table 4 shows that all but one of the correlations (the single-alternative model prediction of reported effort) were significant at the .001 level.<sup>6</sup>

# Testing the Hypotheses

Hypothesis 1 predicted that correlations between the choice model formulations and effort would be stronger than correlations between the difference model formulations and effort, which in turn would be stronger

<sup>\*\*</sup> p < .01 (df = 73).

<sup>\*\*\*</sup> p < .001 (df = 222).

<sup>&</sup>lt;sup>6</sup> The df of the average correlation was obtained by subtracting 3 from the n of each individual correlation and then adding the remainders. Thus, 6 minus 3 equals 3, which is added 74 times to obtain a degrees of freedom for the average within-subjects correlation of 222 (Snedecor & Cochran, 1967).

than correlations between single-alternative model formulations and effort. The results of the tests are presented in Table 5, where the average correlations produced by the three models are listed in columns 1, 3, and 5. The t values of the Hotelling test that the correlation of one model is higher than the correlation of another model are presented in columns 2 and 4.

Column 2 of Table 5 shows that the first part of Hypothesis 1 was not supported. Choice model correlations were not higher than difference model correlations. In fact, for the within-subjects correlations (top of column 2) the difference model correlations were actually higher than the choice model correlations (avg r: .74 > .60; .55 > .44). (The t values were such that a hypothesis predicting the difference model to be stronger than the choice model would have been supported.) Tests comparing the average between-subjects correlations (bottom half of column 2) show there was essentially no difference between the choice model and the difference model (avg r: .31, .28; .21, .21).

Column 4 of Table 5 shows that the second part of Hypothesis 1 was supported. Difference model correlations were significantly higher than single-alternative model correlations. This was particularly true for the within-subjects correlations (avg r: .74 > .30; .55 > .07), but the between-subjects correlation differences were also statistically significant (avg r: .28 > .20; .21 > .08).

Hypothesis 2 predicted that within-subjects correlations would be higher than between-subjects correlations. Table 6 shows that Hypothesis 2 was supported for two of the three models. Within-subjects correlations were substantially higher than between-subjects correlations when using either

Dependent measure	Choice model	t value	Difference model	t value	Single-alternative model
	N	/ithin-subjec	ts correlations	(df = 222)	
Preferred effort	.60	-3.37	.74	12.23**	.30
Reported effort	.44	-2.02	.55	11.46**	.07
	Be	tween-subje	cts correlations	(df = 426)	
Preferred effort Reported	.31	.57	.28	2.02*	.20
effort	.21	01	.21	3.12**	.08

TABLE 5
CORRELATION COMPARISON TESTS FOR HYPOTHESIS 1

<sup>\*</sup> p < .05.

<sup>\*\*</sup> p < .01 (df = 222 or 426) one-tailed test.

Dependent measure	Within-subject correlation $(df = 222)$	z value	Between-subject correlation $(df = 426)$
Preferred	Choice	model	
effort	.60	4.40***	.31
Reported			
effort	.44	3.14**	.21
	Differen	ce model	
Preferred			
effort	.74	7.92***	.28
Reported			
effort	.55	4.93***	.21
	Single-alter	native model	
Preferred	-		
effort	.30	1.22	.20
Reported			
effort	.07	02	.08

TABLE 6
CORRELATION COMPARISON TESTS FOR HYPOTHESIS 2

the choice model (avg r: .60 > .31; .44 > .21) or the difference model (avg r: .74 > .28; .55 > .21), but they were not substantially higher when using the single-alternative model. Using the same data set for both the between- and within-subjects analyses results in the sums of squares being equal for both methods. However, the higher correlations for the within-subjects analysis indicates that more variance in effort is explained using this method than using the between-subjects analysis.

#### DISCUSSION AND CONCLUSIONS

Seven reasons were listed for using the choice model, the primary reason being that the choice model duplicates the original formulation of expectancy theory. Despite these reasons, the choice model proved no better than the difference model and was actually less effective than the difference model for the within-subjects correlation analysis. However, both the choice model and the difference model were clearly superior to the single-alternative model.

<sup>\*\*</sup> p < .01.

<sup>\*\*\*</sup> p < .001.

<sup>&</sup>lt;sup>7</sup> The test comparing the between- and within-subjects correlations assumed a normal distribution, a valid assumption because before the difference between the average r's was tested the r's were transformed to z's. (The z-value computed for the test is not to be confused with the z's resulting from the Fisher r to z transformation.)

A closer look at the data showed that the choice model was not as effective as expected because it continually and significantly overpredicted the amount of effort. The problem of overprediction might have occurred because the choice model does not fully take into account opportunity costs or limits. It also overpredicted whenever an additional hour of effort brought only a very small gain in MFS. Thus a situation of varied relationships (i.e., not positively linear relationships) between effort level and MFS turned out to be a disadvantage for the choice model.

Mitchell (1974) wrote that "before we reject Vroom's original formulation, we should correctly test it" (p. 1075). When the theory was tested as formulated (as a choice model), the results with the between-subjects analysis were not significantly better than for the difference model, and the results for the within-subjects analysis show the choice model to predict effort less well than the difference model. This lack of empirical support, combined with the logical consideration that the theory does not fully address the issues of opportunity costs and limited resources, indicates a possible need for reformulation.

One possible alternative would be to choose effort *budgets* rather than single effort levels. Another alternative would be to approach the allocation of effort in the same manner that microeconomic theory addresses the allocation of money (Ferguson, 1972). This could be done by taking the data for all the hour levels, assuming a time limit, and then filling up that time limit by allocating an hour at a time to whichever activity gave the highest marginal utility for the next additional hour. More sophisticated microeconomic approaches might determine a utility function for effort for each activity (usually using either a log or power function) and then mathematically solve for the optimal time allocation through various optimization techniques.8 Yet another alternative would be to formally incorporate a return-on-effort (ROE) difference model (Kopelman, 1977) into the generally accepted expectancy theory framework. A final alternative would be to use multiobjective utility theory, as described by Keeney and Raiffa (1976). These alternatives still assume a rational motivational model consistent with the principles of expectancy theory. Other alternatives which question the premise of rationality, such as probability matching (Herrnstein, 1974), could be contrasted to this general, rational approach.

One of the primary purposes of this study was to determine which of the three models would likely be the most effective in predicting effort.

<sup>8</sup> A variant of this procedure (Samuelson, 1976) uses the log utility function to determine what percentage of the total time spent will be allocated to each activity. The advantage of this last technique is that no assumption needs to be made concerning how much total time is available.

The analysis indicated that the difference model would probably be most effective. However, there are two reasons for recommending a different strategy for future research than the sole use of the difference model. The first reason is that results were not always consistent and that sometimes the choice model predicted better than the difference model. With the between-subjects analysis (Table 3), for studying as an activity the choice model predicted preferred effort better than the difference model (r: .36 > .14).

A second reason is that data analysis not displayed in the tables showed that when the single-alternative model was tested using low effort as the single alternative it was found to predict sometimes more effectively than the difference model. For example, with studying as an activity, the single-alternative model using low effort correlated -.27 (p>.01) with preferred effort while the difference model correlated only .14. Low effort correlated negatively because the more a person perceived they obtained from low effort the less they intended to put forth high effort. A general observation is that the difference model predicts better than the single-alternative models of high or low effort only when both high and low effort are at least marginally effective predictors (say, |r| greater than .08).

A research strategy that would be both effective and efficient would be to gather expectancies of respondents toward high, medium, and low effort. These data would then be used for testing the choice model, the difference model (both high minus low effort and high minus medium effort), and three single-alternative models (high, medium, and low effort).

A practitioner using expectancy theory to diagnose the motivational system of an organization would find data on three effort levels more useful than data on just high effort alone. In fact, such information would allow a practitioner to emphasize lowering expectancies for medium and low effort, rather than just raising the expectancies for high effort. Such an emphasis on decreasing the expectancies for positive outcomes resulting from medium and low effort is very rarely encountered in the literature, but it might be as effective and maybe more achievable than increasing the expectancies for high effort. Of course making low and medium effort "nonrewarding" has many practical and ethical implications which would have to be seriously considered.

Another future research issue concerns the effective use of the withinsubjects and between-subjects correlation methods. Hypothesis 2 compared the effectiveness of the two correlation methods, and the tests (see Table 6) showed the within-subjects correlation method to be clearly superior to the between-subjects method, though not as much for the singlealternative model as for the other two models. The implication for research would be to use the within-subjects correlation method whenever possible. In addition, practitioners should remember that the ipsative orientation toward worker attitudes is important: workers must choose effort budgets for attaining personal outcomes, and all will not respond similarly to a given outcome, since valences and expectancies will likely vary between persons.

However, the interpretability of within-subjects correlations and the applicability of such results to a real work situation are not clear. If a practitioner knew through within-subjects analysis how a worker valued different options, it may not be practical to attempt to custom tailor a motivational package based on that analysis for each worker. The implication seems to be that for research purposes the within-subjects correlation method is more effective, while for applied purposes the between-subjects correlation method is more useful.

# Summary

This study compared the choice, difference, and single-alternative models of expectancy theory, using both a between- and within-subjects analysis. The subjects were 74 university students who had recorded for several weeks how they spent their time in six general activities. Using expectancy measures that referred to specific hour levels (instead of general effort levels), MFS's were generated for the choice, difference, and single-alternative models in order to predict reported and preferred effort. Contrary to the hypothesis, the difference model was generally the best predictor, but as expected the single-alternative model was the worst. Within-subjects analysis was superior to between-subjects analysis. The choice model was not as effective as expected, because it tended to overpredict, which leaves expectancy theory open to criticism, because the choice model is how the theory was originally formulated. The results indicated that future research should gather data for three levels of effort (high, medium, and low) and then compare the models to determine the most effective predictor for that particular situation.

### **REFERENCES**

- Arnold, H. J., & Evans, M. G. Testing multiplicative models does *not* require ratio scales. *Organizational Behavior and Human Performance*, 1979, 24, 41-59.
- Behling, O., & Starke, F. S. The postulates of expectancy theory. *Academy of Management Journal*, 1973, 16, 373-388.
- Campbell, J. P., & Pritchard, R. D. Motivation theory in industrial and organizational psychology. In M. D. Dunnette (Ed.), *Handbook of industrial and organizational psychology*. Chicago: Rand McNally, 1976.
- Cohen, J., & Cohen, P. Applied multiple regression/correlation analysis for the behavioral sciences. New York: Wiley, 1975.
- Connolly, T. Some conceptual and methodological issues in expectancy models of work performance motivation. *Academy of Management Review*, 1976, 1, 37-46.

- Connolly, T., & Vines, C. V. Some instrumentality-valence models of undergraduate college choice. *Decision Sciences*, 1977, 8, 311-317.
- Constantinople, A. Perceived instrumentality of the college as a measure of attitudes toward college. *Journal of Personality and Social Psychology*, 1967, 5, 197–201.
- Cronbach, L. J., & Furby, L. How should we measure "change"—Or should we? Psychological Bulletin, 1970, 74, 68-80.
- Dachler, H. P., & Mobley, W. H. Construct validation of an instrumentality-expectancy-task-goal model of work motivation: Some theoretical boundary conditions. *Journal of Applied Psychology*, 1973, 58, 397-418.
- Ferguson, C. E. Microeconomic theory. Homewood, Ill.: Irwin, 1972. 3rd ed.
- Herrnstein, R. J. Formal properties of the matching law. *Journal of the Experimental Analysis of Behavior*, 1974, 21, 159-164.
- Hollenback, J. A matrix method for expectancy research. Academy of Management Review, 1979, 4, 579-587.
- House, R. J., Shapiro, H. J., & Wahba, M. A. Expectancy theory as a predictor of work behavior and attitude: A re-evaluation of empirical evidence. *Decision Sciences*, 1974, 5, 481-506.
- Keeney, R. L., & Raiffa, H. Decisions with multiple objectives: Preferences and value tradeoffs. New York: Wiley, 1976.
- Kopelman, R. E. Across-individual and within-individual return on effort versions of expectancy theory. *Decision Sciences*, 1977, 18, 651-662.
- Kopelman, R. E., & Thompson, P. H. Boundary conditions for expectancy theory predictions of work motivation and job performance. *Academy of Management Journal*, 1976, 19, 237–258.
- Korman, A. K., Greenhause, J. H., & Badin, I. J. Personnel attitudes and motivation. *Annual Review of Psychology*, 1977, 28, 75-96.
- Lawler, E. E. III, & Porter, L. W. Antecedent attitudes of effective managerial performance. Organizational Behavior and Human Performance, 1967, 2, 122-142.
- Lawler, E. E. III, & Suttle, J. L. Expectancy theory and job behavior. Organizational Behavior and Human Performance, 1973, 9, 482-503.
- Leon, F. R. Number of outcomes and accuracy of prediction in expectancy research. Organizational Behavior and Human Performance, 1979, 23, 251-267.
- Matsui, T., & Ikeda, H. Effectiveness of self-generated outcomes for improving prediction in expectancy theory research. Organizational Behavior and Human Performance, 1976, 17, 289-298.
- Mitchell, T. R. Expectancy models of job satisfaction, occupational preference and effort: A theoretical, methodological, and empirical appraisal. *Psychological Bulletin*, 1974, 81, 1053-1077.
- Mitchell, T. R., & Beach, L. R. A review of occupational preference and choice research using expectancy theory and decision theory. *Journal of Occupational Psychology*, 1979, 49, 231-248.
- Mitchell, T. R., & Nebeker, D. M. Expectancy theory predictions of academic effort and performance. *Journal of Applied Psychology*, 1973, 57, 61-67.
- Mitchell, T. R., & Pollard, W. E. Instrumentality theory predictions of academic behavior. Journal of Social Psychology, 1973, 89, 34-45.
- Mobley, W. H., & Meglino, B. M. A behavioral choice model analysis of the budget allocation behavior of academic deans. *Academy of Management Journal*, 1977, 20, 564-572.
- Muchinsky, P. M. A comparison of within- and across-subjects analyses of the expectancy-valence model for predicting effort. *Academy of Management Journal*, 1977, 20, 154-158.

- Nebeker, D. M., & Mitchell, T. E. Leader behavior: An expectancy theory approach. Organizational Behavior and Human Performance, 1974, 11, 355-367.
- Peters, L. H. Cognitive models of motivation, expectancy theory and effort: An analysis and empirical test. *Organizational Behavior and Human Performance*, 1977, 20, 129-148.
- Samuelson, P. A. Economics. New York: McGraw-Hill, 1976.
- Schmidt, F. L. Implications of a measurement problem for expectancy theory research. Organizational Behavior and Human Performance, 1973, 10, 243-251.
- Snedecor, G. W., & Cochran, W. G. Statistical methods. Ames: Iowa State Univ. Press, 1967.
- Turney, J. R. Activity outcome expectancies and intrinsic activity values as predictors of several motivation indexes for technical-professionals. *Organizational Behavior and Human Performance*, 1974, 11, 65-82.
- Turney, J. R., & Cohen, S. L. Influence of work content on extrinsic outcome expectancy and intrinsic pleasure predictions of work effort. *Organizational Behavior and Human Performance*, 1976, 17, 311–327.
- Vroom, V. H. Work and motivation. New York: Wiley, 1964.

RECEIVED: December 28, 1981