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different tasks. tween information-processing components of operations, and systems that support this vital dictive models of the underlying information of the similarities and differences among consider not only how a given memory task scope of memory tasks that are studied, and to single task and for examining the relations bethe information-processing components of a thus requires techniques both for dissecting human capacity. The study of human memory memory tasks that may help us develop prememory tasks are interrelated. It is the study is performed but also how different kinds of memory has driven scholars to broaden the tems. Recognition of the complexity of human ent types of information, operations, and syscesses that undoubtedly rely on many differ-Human memory subsumes a multitude of pro-

researchers as evidence for the operation of tions between memory tasks is taken by some raview). Converging evidence for dissociaanalysis has yielded numerous examples of parallel effects and dissociations among tasks formance in a given task. This kind of task tasks is to look for experimental factors that ponents both within and between memory see Richardson-Klavehn & Bjork, 1988, for a asks, or on different aspects of subjects' perhave different effects on different memory One standard method for separating com-

ving, 1996; Schacter & Tulving, 1994; Tulving multiple memory systems (e.g., Nyberg & Tul-

relatively independent of one another (see Sternberg, 1998, for a review). guish between stages of processing that are curve (e.g., Glanzer & Cunitz, 1966; Murdock, single task. For example, dissociations beberg, 1969) have enabled researchers to distintask analysis with response time data (Sternmemory systems. Techniques for performing tinction between short-term and long-term tween different portions of the serial position [962] have been taken as evidence for the dis-Dissociations can also be observed within a

Correlations derived from contingency tables are measured using a 2×2 contingency table are usually dichotomous variables (recall of an at least in part, the degree to which their memvidual subject-item (e.g., DaPolito, 1967; Estes, item on test 1 and on test 2) these correlations tions, and/or systems. Because the responses ory processes tap common information, operaapplied to the outcomes of successive memory method of task analysis: contingency analyses tion between the observed responses reflects, cessive memory tests at the level of an indithe correlation between the outcomes of suc-1960; Tulving & Wiseman, 1975). The correlaests. This method has been used to examine This chapter is concerned with another Contingency Analyses of Memory

MICHABL J. KAHANA

ferent con ditions. provide information not present in simple measures of memory performance under difwell founded, and that these measures can review is that such avoidance behavior is not to study memory. The position taken in this to shy away from using correlational measures confoundings have caused many investigators tinuous rneasures. Fears of these potential foundings that face correlations between conare susceptible to all of the potential con-

mary and some concluding thoughts. analyses. Finally, the last section offers a sumthe controversy over the use of contingency in memory research. The next section reviews plications of the use of contingency analyses gency analyses. The fourth presents three approvide a context for the discussion of continfunctional and correlational approaches to tions. The first two sections briefly discuss The chapter is organized into six short sec-

Functional Approaches

curacy in human memory, see Kahana & Lofcorrelated pictures of human behavior (for a discussion of the relation between RT and ac-RT and accuracy provide complementary but to be made (response time or RT). In general come) and how long it takes for the response item is remembered (usually a binary outtrieval conditions one can ask whether the experimental condition. Under particular reunit of information is a subject item in a given In laboratory memory experiments, the basic

ables we have manipulated in our experiteaches us how memory depends on the varito the independent variable(s) of interest in with that in another can be made with respect comparison of performance in one condition ther physically or statistically, so that the conditions and taken as estimates of the "ef the experiment. This functional approach item differences are typically controlled, fect" of the manipulated variable. Subject and jects and many items, are compared among terest. Mean values, computed over many subdepends on (1) the subject, (2) the item, and (3) the experimental condition. In the functional approach, the latter effect is of sole in-Whether or not a target item is remembered **e**.

actions. For example, the degree to which a delay between study and test affects retention Comparisons across tasks often reveal inter-

> ferent underlying sources of information. of the nonlinearity in the forgetting function and not because recognition and recall tap difsimple model, the interaction results because the kind of test (recognition vs. recall). In this tion succeeds when the strength is above a interaction between the retention interval and (K_2) . If $K_1 < K_2$, then there will be a statistical when the strength is above a recall threshold recognition threshold (K1); recall tion, a and b are positive constants). Recognimation. Consider a simple model in which of memory tasks tap different kinds of infornot necessarily mean that these different kinds memory as a function of time (in this equa- $S(t) = at^{-b}$ defines the strength of each item in fragment completion, free recall). This does memory test employed (e.g., item recognition, of a given item is mediated by the kind of succeeds

Correlational Approaches

contemporary terms, the strength of the con-"strength" of the underlying memory (or, in relation is consistent with the view that RT text-item association), low-confidence judgments. This positive corconfidence judgments are made faster than RT and confidence judgments covary-highfully documented the well-known finding that to make confidence judgments. Murdock and sponse in a recognition task that asks subjects even a single response. Consider a single reobtained from different tasks, a single task, or Anderson (1975; see also Koppell, 1977) carestudy the relations between different measures manipulating any variables. Rather, one can The correlational approach need not involve confidence are both affected by the

duces correlated effects on both dependent subjective organization. Rather, the indepenvariables. dent variable, number of study trials, prodirectly manipulate either recall probability or tional relationship, the experimenter does not tive organization). Although plotted as a funcingly stereotyped (the phenomenon of subjec-Tulving (1962, 1966) showed that as subjects of performance are correlated. For example, learn a list, their recall order becomes increas-Within the free-recall task, different aspects

call tasks. The correlation between item recogognition, associative-recognition, and cued-remeasured subjects' performance on item-recent memory tasks, Kahana and Rizzuto (1999) Examining the correlation between differ

> 0.80). These subject correlations involve comwhereas the correlation between associative correlations across subjects can be positive ten yield very different results. In some cases, correlations across subjects and items can ofgator to compute a Pearson product-moment ther subjects or items), they allow the investicall both tap "strength" but recall requires a recognition and cued recall was high (r'= nition and cued recall was moderate (r' = 0.33) tive. The use of subject and item based correwhile correlations across items can be negacorrelation. Mandler (1959) demonstrated that these techniques involve averaging (across eipositive correlation across items. Because higher threshold, we would expect a strong is positive. For instance, if recognition and rewould expect that the correlation across items tion and recall tap common information, we across items in one task or situation is correasks whether the ordering of performance puting an average for each subject across Underwood, Boruch, & Malmi, 1978). search (for exceptions, see Rubin, lations is fairly uncommon in memory reitems in another task or situation. If recognilated with the ordering of performance across many items. Another correlational technique

Successive Tests Contingency Analyses of

ognition over recall. However, examining the thetical data show the usual advantage for recgency table from the pairs of outcomes on test menter presents B items from the studied of items (A-B) and are then given two successive tests: item-recognition followed by cued gives hypothetical data for successive itemover subjects or items we compute a continusing a recall test. Because we cannot average as cues. The same B items are tested twiceattempt to recall the B items given the A items a lure. In the second, cued-recall test, subjects recall. In the item recognition test, the experirelation between recognition and recall at the relation one step further, we can examine the recognition and cued-recall tests. These hypofirst using a recognition test and then, later, the canonical procedure, subjects study pairs we employ the method of successive tests. In level of individual subject items. To do this, laking this analysis of the recognition-recall (lures). Subjects judge each item as a target or and test 2 for each subject item. Table 4.1 intermixed with nonstudied items

Table 4.1 Hypothetical data comparing the same subject-items tested successively. recognition and recall performance for

																											1		
Mean	26	25	24	23	22	21	20	19	18	17	16	15	14	13	12	11	10	9	œ	7	6	51	4	ယ	2	1	Subject-Item		no oumo ouojo
0.69	1	0	12	0	1	0	-	0	L	_	در	Ь	μ.	Ъ	_	0	_	1	0	_	μ.	0	_	,_	0	1	Subject-Item (recognition)	Test 1	or reason on
0.42	_	0	_	0		_	0			0	0	,_	0	<u>,,</u>	0	0	0	1	_	_	0	0	0	_	0	0	(recall)	Test 2	. 6 2.0 4 10000

speaking, this means that an item that is rerecall requires a higher threshold. Strictly recall tap exactly the same information, there is recognition failure of recallable items of Tulving and Thomson (1973), indicate that pothetical data in table 4.1, and the real data the level of subject items, exactly one. The hycorrelation between recognition and recall, at called will always be recognized, making the strength theory of memory, recognition and stronger. For example, according to a simple contingency table allows us to say something -contradicting the predictions of strength but

pendency, as it is calculated from the cells in tion is often referred to as contingency or dethe data in a contingency table. This correlabetween recognition and recall by tabulating ables, we compute the subject-item correlation Because the test outcomes are binary vari-

a 2 × 2 contingency table. Yule's Q is a popular measure of correlation for 2 × 2 contingency tables (Bishop, Feinberg, & Holland, 1975). Like a standard Pearson correlation, Yule's Q varies from -1.0 (perfect negative correlation) to +1.0 (perfect positive correlation). For a discussion of other correlational measures that are applicable to contingency tables, see Poldrack (1996).

Table 4.2 shows the results of a contingency analysis applied to the hypothetical data in table 4.1. For each subject item, the combination of outcomes on test 1 and test 2 determine the values of the four cells in the contingency table (A, B, C, & D). Yule's Q is given by the equation:

$$Q = \frac{A \times D - B \times C}{A \times D + B \times C}.$$

For the hypothetical data shown in table 4.1, the dependency between the item recognition and cued recall, as measured by Yule's Q, is 0.5. This dependency of 0.5, measured at the level of subject items, does not mean that the correlation across subjects is 0.5 or that the correlation across items is 0.5. As pointed out by Tulving (1985), correlations at the level of subjects, items, and subject items can yield very different values.

This example was chosen because there is a vast literature examining the relation between recognition and recall using the method of successive tests. This literature, first surveyed by Tulving and Wiseman (1975) and more recently reviewed by Nilsson and Gardiner (1993), reveals an invariance: successive-item recognition and cued-recall tasks almost always yield moderate correlations, with Yule's Q rarely deviating from the range 0.3 to 0.75. This makes the successive-testing data inconsistent with a simple strength-threshold theory (Tulving, 1983), and also with certain dis-

Table 4.2 Contingency table for the hypothetical data shown in table 1.

Test 1

Test 2	
0	
B = 9	-
b = 2 d ≈ 6	0

tributed memory models (Kahana & Rizzuto, 1999).

Together with Yule's Q, the percent correct for tests 1 and 2 fully characterize the data in our 2 × 2 contingency table. Without the correlational information, the accuracy data would do little to constrain theory. Because theories of memory make claims about the processes acting on a given subject item, it is important to consider not just accuracy data but also the correlation between tests.

The foregoing example illustrates the comparison of tasks using contingencies of outcomes on successive tests. In comparing performance across the two tests, it is important to recognize that the successive tests do not necessarily measure the "same thing." The first measure is of the consequences of study—subject to all the input/output interference effects that may operate on retrieval of individual list items (e.g., Tulving & Arbuckle, 1966). The second measure is affected not only by study and the interpolated conditions but also by the earlier test and its outcome.

One can also analyze contingencies across successive trials of the same tesk. This approach was fruitfully employed by Estes (1960) in his studies of one-trial learning and by Tulving (1964) in his analysis of inter-vs. intra-trial forgetting in multitrial free recall.

Empirical Regularities in Successive Testing Experiments

ciations, the recognition failure paradigm, and onstrations of one-trial learning (e.g., Estes, Gardiner, 1993; Tulving & Wiseman, 1975) There are many interesting examples of the some recent results pertaining to the classic discussed: independence of A-B and A-C assodence of A-B and A-C associations in the repoon & Moscovitch, 1989). Other important plicit memory tasks (e.g., Hayman & Tulving, successive explicit and implicit memory tasks successive episodic memory tasks (Nilsson & well-studied problems include comparisons of use of contingency analyses, but the most applications of successive tests include dem-1989b; Tulving & Hayman, 1995; Withers-Schacter, & Stark, 1982), and successive im-(Hayman & Tulving, 1989a, 1989b; Tulving, 1971). In this section, three applications are 1960) and demonstrations of the indepen-(e.g., Greeno, James, & Dapolito,

question of associative symmetry versus independent associations. These three examples were chosen because they yield correlations of approximately 0.1, 0.5, and 0.9, respectively.

Analysis of Competing A-B and A-C Associations

to some degree of mastery. Briggs (1954) showed that with increasing trials of A-C sociative interference is the A-B/A-C a review). A major technique used to study asguiding a generation of research in verbal The classic associative interference theory of verbal learning played a prominent role in competition, Barnes and Underwood (1959) der conditions designed to minimize response digm. Subjects first master a list of A-B pairs. verbal learning played a prominent role B list, then study the A-C list, and then are call). In this procedure, subjects study the Aamine this retroactive interference effect unto recall Next, subjects study a second list of A-C pairs the basic data on retroactive interference. two very different theories that both explain stronger response dominates. Here we have sociations are independent, with the earlier view (McGeoch, 1942) that asthe individual A-B associations. This contrasts the decrease in tral tenet of associative interference theory wood unlearning-recovery hypothesis—a cenphenomenon. According to the Melton-Understrating the classic retroactive interference call fewer B items and more C items, demonincreasing trials of A-C learning, subjects reboth the B and C responses in any order. With given each of the A items and asked to recall MMFR stands for modified-modified free reproposed the now-classic MMFR procedure learning comes a decrease in subjects' ability earning (see Postman & Underwood, 1973, for learning results from specific unlearning of the original A-B associations. To ex-A-B recall following A-C but that para-Ĕ

Applying a contingency analysis to the MMFR data, Dapolito (1967) examined the dependence between B and C recall. According to the unlearning-recovery hypothesis, learning A₁-C₁ causes specific weakening of the A₁-B₁ association. As a consequence, recall of B₁ and C₁ should be negatively correlated. In contrast to these predictions, Dapolito found near independence between recall of B₁ and C₁. Numerous additional studies supported the independence position (Abra, 1969; Greeno et al., 1971; Martin, 1971; Wichawut & Martin, 1971), Because these studies did not report their results in terms of Yule's Q, it is hard to

conditions obtained using the MMFR techcompare their observations with other studies nique. Across these conditions, the dix A presents a database of 32 experimental of correlations among successive tests. Appen-Chappell & Humphreys, 1994; Mensink & tinguishes among theories of associative interreveal a regularity of human memory that disation of 0.31. Again, contingency analyses value of Yule's Q is 0.08, with a standard devi-Raaijmakers, 1988) have been framed in a way human memory (e.g., Murdock, 1982, 1997; ference. Several major mathematical models of A-C independence. that captures the basic phenomenon of A-B, mean

The Recognition Failure Paradigm (Tulving & Thomson, 1973)

ognition-failure paradigm (Flexser & Tulving, pendency between item recognition and cued recall to the probability of recognition itself well fit by a quadratic formula relating the conditional probability of recognition given wide range of experimental conditions, was item recognition and cued recall, across a cued recall. Tulving and Wiseman (1975) obsively, first by item recognition and then a list of word pairs and are then tested succesbriefly in the previous section, subjects study Wiseman, 1975). In this approach, discussed the influential, if somewhat controversial, rectingency analyses to the study of memory is Perhaps the most common application of confunction, describes a moderate degree of de-This function, known as the Tulving-Wiseman served that the dependency relation between 1978; Tulving & Thomson, 1973; Tulving

are subject to a number of bility of recognition given recall, P(Rn|Rc) by either (1) knowing the probabilities of successful outcomes on tests one and two as well on test 1 and on test 2. of both Yule's Q and the probability of success can be avoided by examining results in terms of P(Rn) and P(Rc). These potential constraints predicted values of Yule's Q derived from the Hartry, 1990; Hintzman, 1992). Second, the P(Rc) in experiments where the probability of P(Rn|Rc) is constrained to be less than P(Rn)/ contingency table are completely determined -1 or greater than +1 for some combinations Tulving-Wiseman function, can be less than recall exceeds that of recognition (Hintzman & Analyses based on the conditional proba-The cells in a 2×2 problems: First

test 2 conclitional upon test 1 success, and the success on test 1, the probability of success on outcomes, or (2) knowing the probability of as the correlation (Yule's Q) between these and Bowyer (1980) as well as Batchelder and upon test 1 failure. (A proof of this assertion probability of success on test 2 conditional Riefer (1995) This later approach is used by Humphreys is available from the author upon request.)

summarized by Nilsson and Gardiner (1993) nition and cued recall under varying conditions (see Nilsson & Gardiner, 1993, for a parmethodology yield the moderate dependenexperiments using proximate ly 0.55 in these experiments. Not all sents a basic fact of human memory. consistently obtained moderate correlation bethese studies, the mean value of Yule's Q is 0.55 and the standard deviation is 0.12. The test followed by a cued-recall test). Across pairs and are then given a yes/no recognition ods (i.e., subjects study a list of common word nificant arnounts of data using standard meth-Yule's Q values from studies that gathered sigconstructed to push the correlation up or down. Appendix B presents a database of that many of these studies were especially value of 0.55. reveals significant variation around the mean tial review). looked at the dependency between item recog-Wiseman function. A great many studies have cases as boundary conditions on the Tulvingor from semantic redundancy of study pairs. pendencies can result from shallow encoding Gardiner, 1993). Higher recognition-recall detion (Nilsson, Law, & Tulving, 1988; Nilsson & cies described by the Tulving-Wiseman func-Yule's Q. one typically obtains a value of aptween item recognition and cued recall repre-Using a measure of dependency such as and Gardiner (1991) refer to The reason for this variation is Examining the correlations the successive testing these

unity (Q = 0.94). tive recognition tasks correlations approach tially (Q = 0.81). Finally, for identical associaand cued recall), the correlation rises substansuggests that the reason for the moderate desociative recognition tasks (Q = 0.59). This that the moderate dependency obtained in tion (e.g., successive associative recognition trast, whem both tasks tap associative informaother test taps associative information. In conpendency is that the one test taps item and the tasks is also found in successive item and assuccessives item recognition and cued recall Kahana and Rizzuto (1999) have shown

> classes of memory models including Met-calfe's (1982, 1985) CHARM model, Mur-dock's (1982) TODAM model and the matrix gency analyses place constraints on the modexplicit predictions about subject performance ues. These results show that if models make mental data for all reasonable parameter valsume identical representations for individual have distinct representations. Models that assumes that item and associative information man, 1987). Each of the preceding models asin the goodness of encoding (see also Hintzcan all produce the moderate dependencies model of Humphreys, Bain, and Pike (1989) distributed storage of auto- and hetero-assomemory model that assumes composite and certain conditions, CHARM (a mathematical els and, in doing so, provide insight into the in the successive testing paradigm, continthat are too high, deviating from the experirequired, but only if you allow for variability Tulving-Wiseman function and conditions ciative information) can account for both the findings. Metcalfe (1991) shows that under much debate over the interpretation of these function of human memory. items and associations produce correlations Kahana and Rizzuto (1999) found that several that result in deviations from the function Though the data are clear, there is still

Independent Associations Associative Symmetry vs.

retrieval have been taken as evidence for the stored item (Kohler, 1940). In this model, fortional links between stored items (e.g., Ebbinghaus, 1885/1913; Robinson, 1932). The holiscal literature; The independent associations models of association are present in the classimemory is the nature of associations. Two asymmetric retrieval have been taken as eviposition of associative symmetry; findings of Findings of equivalent forward and backward has a minimal effect on associative strength merous studies, suggests that order of study ward recall (see Ekstrand, 1966, and Kahana, formed patterns combining elements of each tic model considers associations to be newly model considers associations to be unidirec-Another basic question in the study of human 1999, for a review). The evidence, from nuwhether forward recall is easier than backbetween these two positions focused on 1962). Early studies addressing the differences items prew and backward associations between are symmetrical (Asch & Ebenholtz,

> tions (e.g., Wollen, 1970a,b; Wolford, 1971). dence for the position of independent associa-

gency analyses applied to the outcomes of successive cued-recall tests. First, one associatested (B is presented as a cue to recall A) recall B) and then the other association is associations, and hence the recall probabili-B), the strength of the forward and backward try requires that for a given studied pair (Ation is tested (say, A is presented as a cue to this question may be gleaned from continties, must be identical. Evidence relevant to A much stronger test of associative symme-

ward or backward-backward), the average Yule's Q value was 0.88 (SE = 0.012). For recall tests of all pairs in the list. Across the two word pairs and then gave successive cued-reor backward-forward), the average Yule's verse successive tests (i.e., forward-backward identical successive tests (i.e., forward-forverse successive tests (separate Yule's Q val-ues were calculated for each subject). For ward, T1-backward/T2-forward, T1-backward/ T1-forward/T2-forward, T1-forward/T2-backcombinations of forward and backward orders: successive tests, word pairs were tested in all backward recall are near unity. ciations model of paired-associate learning not differ statistically. The independent assovalue was 0.91 (SE = 0.017). These effects did Yule's Q was computed for identical and re-T2-backward. To test the symmetry principle, finding that correlations between forward and Kahana, 1999), but they cannot explain lenge to symmetrical associative models (see for asymmetries that pose a resolvable chal-(Wolford, 1971; Wollen, 1970a,b) can account Kahana (1999) had subjects study lists of

Other Applications

(Hayman & Tulving, 1989a, 1989b; Tulving et al., 1982; Tulving, Hayman, & Macdonald, spoon & Moscovitch, 1989). These studies 1991; sively applied to the study of implicit memory ular, contingency analyses have been extenanalyses that are not reviewed here. In particof other important applications of contingency strain theories of memory. There are a number dramatically and how they can be used to conhow intertask contingencies can vary quite The previous three applications illustrated have shown that successive implicit tasks can (with identical cues). In contrast, successive yield dependencies ranging from near zero nonoverlapping cues) to around 0.5 Tulving & Hayman, 1995; Wither-

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fected by subjects' intentionality (i.e., whether or not subjects are instructed to focus retrieval ment completion tasks can be dramatically afresults is that the dependency between tragidentical cues). An important aspect of these around 0.5 (with different cues) to 1.0 (with explicit tasks yield dependencies ranging from dence, to support a multiple memory systems view (e.g., Schacter & Tulving, 1994). gether with findings of functional indepenand colleagues have used this evidence, toon their memory for the study list). Tulving

The Controversy

analyses to a broad range of memory paralapsed data. In practice, what does this mean? at worst uninterpretable. His arguments are correlations between tasks, as measured by versy surrounding their use as an analytic digms, there has been considerable controson, 1973). Recall that in these experiments, and cued-recall tests (e.g., Tulving & Thommiliar example of successive item-recognition Here we examine two cases, both using our farelations that were not present in the pre-coldata across subjects or items can give rise based Yule's Q, are at best difficult to interpret and tool. Hintzman, a vocal critic, has argued that Despite the fruitful application of contingency Yule's Q is almost always between 0.30 and 1980). This refers to the fact that collapsing on Simpson's paradox (Hintzman,

coding recall; rather it is a spurious correlation in-duced by a third actor—variability in endiscover that all responses are segregated into cells A (+/+) and D (-/-) of the contingency mon operations underlying recognition and exactly +1.0. This result does not reflect comsponses in cell D). The resulting correlation is recognized nor recalled (accounting for the retest 2 (accounting for the responses in cell A). sponses. The few items that subjects rehearsed table, with D (-/-) having the majority of representation. Upon tabulating our results, we those pairs throughout the duration of the list strategy: they attend to the first few pairs and being presented, subjects adopt the following list of A_i-B_i word pairs for study. As the list is were both recognized on test 1 and recalled on then close their eyes and keep rehearsing The remaining, unseen items, were neither Case 1. The experimenter presents a long

1934

contingency table, most of the pairs in group of B, by varying its word frequency or its simi-B_i. Suppose we now divide our pairs into two selected pairs to induce a negative correlation. on recognition and recall and then specially we identified factors that have different effects This illustrates another spurious correlation: nition and cued recall would be close to -1.0. 1 would be in cell C (+/-) and most of the B, by varying the associative relations between easy to recall. We can manipulate recognition we make the B-items hard to recognize and to recognize but hard to recall, and in group 2 groups: in group 1, we make the B-items easy sive item recognition and cued recall tests for ents a list of A₁-B₁ pairs, and then gives succesthis case, the correlation between item recogpairs in group 2 would be in cell B (-/+). In A_i-B_i and between A_i-B_j. After tabulating our larity to the lures. We can manipulate recall of Case 2. Here again, the experimenter pres

ment completion tests, Hintzman and Hartry at work in successive recognition and fragcrease in the observed correlation. do point out, as all practitioners who use coronstrations do not teach us very much. They served correlation between tasks. Such demlated manmer can induce an increase or deaffects the two outcome variables in a corretions can occur. in particular, variability that relations ought to know, that spurious correlaitems could produce large changes in the ob-(1990) found that selecting different subsets of In an attempt to show Simpson's paradox

changed dramatically. correlation between successive tasks has not lapsed across only subjects, or only items. When these techniques have been used, the subjects and items to permit separate Yule's Q served value of Yule's Q would be somewhat and item difficulty both have positive effects analyses for contingency tables that are colrequires the collection of sufficient data across these sources of variability. Another approach nique for adjusting Yule's Q to account for items. Flexser (1981) provided a useful techhigher when collapsing across subjects or on memory performance. In this case, the ob-Suppose that variability in subject ability

test 1 and test 2, is much larger than the meaencoding of information important for both this external source of variability, affecting the tors not intrinsic to the tasks being studied the level of subject items that is caused by facthey do not adjust for potential variability at (e.g., trial by trial fluctuations in attention). If One problem with these approaches is that

> smaller than what is observed experimentally surable variability across subjects and items then it is possible that the "true" correlation is

nipulations produced dramatic changes in the correlations between successive tests. These cessive memory tasks. in determining the correlations between sucgoodness-of-encoding only plays a minor role findings strongly suggest that variability in by about 0.10). In contrast, informational machanges in variability only increased Yule's Q effects were extremely small (the largest structed, high-variability lists. However, the which presentation rate, or number of repeticomparing these mixed lists with pure lists (in on the dependency relations among tasks by ing number of repetitions or presentation is to create lists in which the variability is artiiments of this kind and found significant hana and Rizzuto (1999) conducted two exper tions, is uniform across all study pairs). Karate). One can assess the effect of variability items (where strength is manipulated by varyficially increased by mixing strong and weak increases in Yule's Q for the specially con-One way to address this potential problem

call performance for the recognized items A final class of criticisms relate to the "priming" of second test performance by the first test (e.g., Humphreys & Bowyer, 1980). As higher cued-recall performance for those items experiments, Humphreys and Bowyer found more than for the nonrecognized items, nized items may be strengthened during the recall by prior recognition. In essence, recogcreased by differential facilitation of later dency between recognition and recall is inand the effect of the first test on memory. a direct measure of the study event. Rather, compared with items only tested in the cueditem recognition and cued recall. In several thereby increasing the dependency between recognition test more than nonrecognized test, and therefore the second test is no longer first test may affect performance on the second stated previously, it is well known that the items. This transfer effect enhances cued-re-Humphreys and Bowyer argue that the depenthe second test reflects both the study event that were tested in the recognition phase (as

be largely due to this priming effect. The problem with this account is that nonrecognized relation between recognition and recall could prime subsequent recall, Humphreys and Bowyer show that the observed moderate cor-If one assumes that only recognized items

even experiments that fail to show significant effect accounts only for some part of the obman & Tulving, 1976). These results suggest (Begg, 1979; Donnelley, 1988). In addition, served dependency between these tasks. tion test does influence subsequent recall, this tween item recognition and cued recall (Wiseltems can also facilitate subsequent recall "priming" yield moderate dependency behat although the outcome of the first recogni-

1981; Gardiner, 1991). gency analyses we need to proceed with cauning away from the complexities of continthese potential confounds? Rather than runsistent and theoretically interpretable (Martin, method of successive tests, and the variation highly reliable and replicable results using the n dependencies among tasks have been conion. Researchers have managed to obtain Where, then, do we stand in the face of

Conclusions

high dependencies (Yule's Q = 0.90). With improvided evidence supporting the multiple memory systems view (e.g., Tulving, 1985; tween implicit and explicit memory tasks has containing identical information yield very episodic memory with identical cues or cues analyses applied to successive memory tasks. function of cue overlap. This difference observed: Yule's Q ranges from 0.1 to 0.5 as a plicit memory tasks, a very different pattern is (Yule's Q \approx 0.55) whereas successive tests of tasks yield moderate levels of dependency Successive item-recognition and cued-recall relation between successive tests varies in re-As shown in the applications section, the cor-This chapter reviewed the use of contingency liable ways across different task comparisons.

bate. Schacter & Tulving, 1994). Although other exses have played an important role in this deplanations may well exist, contingency analy-

diner, 1993; Wiseman & Tulving, 1976). This of performance do not seem to affect the taskers have shown that experimental variables sive tasks. overall performance levels on the two succesdependent variable that is independent of tertask contingencies: they represent an added finding highlights an important feature of in-1978; Kahana & Rizzuto, 1999; Nilsson & Gartask contingencies (see Flexser & Tulving, that have a significant effect on overall levels tion between recognition and recell, research-Applying contingency analyses to the rela-

analyses, reviewed in this chapter and elsesame, the exact form of the second test can tions are the same, and the target items are the where (e.g., Tulving, 1985; Tulving & Hayman (see Hayman & Tulving, 1989a; Tulving et al., first test remains the same, the study condidency varies systematically with the nature of lations between successive tests: the depen-1995) illustrate a striking feature of the corredramatically alter the observed correlation the two tests. Even in a situation in which the These and other examples of contingency

experience are frequently tested repeatedly. resemble life in which, too, the results of an require a different approach than that adopted variables. In this respect, these experiments for the study of the effects of independent exhibited, and problems of interpretation that the fear that many students of memory have of how human memory works—this despite mense value in enriching our understanding jects acquire in the laboratory has been of im-Repeated testing of the knowledge that sub-

Appendix A

A selected database of studies that examined the correlation between B and C responses in

the MMFR paradigm. The mean Yule's Q value for the included conditions was $0.08 \pm .11$ (95% confidence interval).

Tulving and Watkins (1974) Tulving and Watkins (1974) Tulving and Watkins (1974) Wichawut and Martin (1971) Wichawut and Martin (1971) Wichawut and Martin (1971)	Postman and Gray (1977) Tulving and Watkins (1974)	Koppenaal (1978) Koppenaal (1978) Koppenaal (1978) Koppenaal (1964) Postman (1964) Postman (1964) Postman (1964)	Delprato (1972) Delprato (1972) Delprato (1972) Delprato (1972) Koppenaal (1978) Koppenaal (1978) Koppenaal (1978) Koppenaal (1978)	Study Abra (1969) Abra (1969) Abra (1969) Abra (1969) Delprato (1972) Delprato (1972) Delprato (1972)
8'C' 8'C 8C' AC(4) AC(8) AC(12)	Substitution BC B'C' B'C BC' BC'	Retention Interval = 24 hours Retention Interval = 72 hours Retention Interval = 1 week set 1 set 2 set 3	() (6) (7) (8) (8) (8) (9) (9) (9) (9) (9) (9) (9) (9) (9) (9	Condition 48 hr0 hr. 48 hr24 hr. 48 hr48 hr. 24 hr24 hr. 24 hr24 hr. AC(4,2) AC(8,4) AC(8,4)
0.48 0.07 0.62 -0.24 0.13 0.00	-0,13 0.16 0.44 0.51 0.53 0.39	0.00 0.06 0.23 -0.32 0.31 0.35	-0.16 -0.38 0.14 -0.32 0.00 0.37 -0.62	Yule's Q -0.22 -0.13 0.47 -0.10 -0.05 0.29 -0.32

Appendix B

A selected database of successive item recognition—cued-recall experimental conditions. Studies that used pairs of common English nouns, or weak associates, were included; studies that used strong associates or non-English materials were excluded. Studies that

used categorization or free association before the successive tests were excluded. Finally, experiments where the standard error on Yule's Q was greater than 0.2 were also omitted. The mean Yule's Q value for included conditions was 0.55 ± 0.05 (95% confidence at interval).

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cussion. Correspondence should be addressed 9110. E-mail may be sent to: kahana@cs. Brandeis University, Waltham, MA, 02454 to M. J. Kahana, Center for Complex Systems, Howard for very helpful comments and disments on a previous version of this chapter. sity. I am grateful to Ben Murdock for his comby N.I.H. grant MH55687 to Brandeis Univer-Acknowledgments This research was funded Thanks are also due to Dan Rizutto and Marc

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Part II: Memory in the Laboratory