T HE CLASSICAL VIEW is a psychological theory about how concepts are represented in humans and other species. In philosophy, the origins of this view go back to Aristotle, while in experimental psychology the view can be traced to Hull's 1920 monograph on concept attainment. In assembling our rendition of the classical view, however, we have relied mainly on contemporary sources. These sources include philosophically oriented studies of language (for example, Katz, 1972, 1977; Fodor, 1975); linguistic studies (Lyons, 1968; Bierwisch, 1970; Bolinger, 1975); psycholinguistics (Fodor, Bever, and Garrett, 1974; Miller and Johnson-Laird, 1976; Anglin, 1977; Clark and Clark, 1977); and psychological studies of concept attainment (Bruner, Goodnow, and Austin, 1956; Bourne, 1966; Hunt, Marin, and Stone, 1966).

While we think we have captured some common assumptions in these various sources, we are less sure that we have been faithful to the spirit of these works. For instance, the psychological studies of Bruner, Goodnow, and Austin (1956) were more concerned with the strategies people use in determining the relevant features of concepts than with supporting the classical view. Indeed, these authors even devoted one chapter of their influential book to concepts structured according to the probabilistic view. Still, the bulk of their effort employed artificial concepts structured according to the classical view, and there is no guarantee that the strategies that Bruner, Goodnow, and Austin turned up will be easily extendable to other views of concepts. Similar caveats apply to many of the other sources.

In terms of distinctions drawn earlier, we will be concerned here exclusively with feature descriptions, since all the sources given above (as well as many not listed) have analyzed concepts in terms of features. Also, it seems that practitioners of the classical view

#### The Classical View

have been primarily interested in characterizing the core of concepts, not their identification procedures, and as mentioned earlier, our treatment of the view will focus on the core. Finally, a word about the role of process models is in order. The classical view is a proposal about representations, not about processes. Once we have described the representational assumptions that make up the classical view and the criticisms of these assumptions, we will give some consideration to process models that can be generated from the view.

#### **Representational Assumptions**

#### SUMMARY REPRESENTATIONS

The first assumption is as follows: The representation of a concept is a summary description of an entire class, rather than a set of descriptions of various subsets or exemplars of that class. To illustrate, in representing the concept of bird we would not list separate descriptions for different species (like robin and chicken) or for specific instances (like our pet canary Fluffy), but rather would give a summary representation for all birds. As Rosch (1978) has emphasized, condensing a concept into a single summary greatly reduces the amount of information we need to store.

This notion of a summary representation is sufficiently important that it is worth specifying some explicit criteria for it. A summary representation, then, (1) is often the result of an abstraction process, (2) need not correspond to a possible specific instance, and (3) applies to all possible test instances. Thus: (1) one's summary representation for fruit is often based on induction from specific instances (as well as on facts one has been told about fruits in general); (2) the representation might contain fewer features than would be found in the representation of any possible instance; and (3) whenever one is asked whether or not a test item designates an instance or subset of fruit, the same summary representation is always retrieved and examined.

## Necessary and Sufficient Features

The heart of the classical view is contained in its second assumption: The features that represent a concept are (1) singly necessary and (2) jointly sufficient to define that concept. For a feature to be singly necessary, every instance of the concept must have it; for a set of features to be jointly sufficient, every entity having that set must be an instance of the concept. It is convenient to illustrate with a geometric concept – squares again. Recall that the concept of square may be represented by some in terms of the following

## Categories and Concepts

features: closed figure, four sides, sides equal in length, and equal angles. Being a closed figure is a necessary condition, since any square must have this feature; the same is true of the features of having four sides, the sides being equal, and the angles being equal; and these four features are jointly sufficient, since any entity that is a closed figure, has four sides equal in length, and has equal angles must be a square. We will sometimes refer to such necessary and sufficient features as *defining* ones.

Many scholars who have written about the classical view have emphasized that this assumption is about necessity or essentialism, not probability (see Cassirer, 1923; Katz, 1972). It is not just that all squares happen to have four sides, but rather that having four sides is essential to being a square. Or take another example: the defining features of bachelor – male and unmarried – are not only true of all bachelors (which is merely a statement about conditional probabilities), but are essential conditions for being a bachelor. To appreciate this distinction between probability and essentialism, suppose that the feature of "not wearing wedding bands" is also true of all bachelors. While this feature has the same conditional probability as being unmarried, only the latter would be essential.

It is important to note that this assumption about defining features implies that natural concepts are never disjunctive. To illustrate, let us consider first a *totally disjunctive* concept, which says that an instance either has features  $F_1$ ,  $F_2$ ,  $F_i$ ,  $F_n$  or features  $F'_1$ ,  $F'_2$ ,  $F'_i$ ,  $F'_n$ ; that is, two instances need have no features in common. This means there are no necessary features, which violates the classical view's assumption about defining features. Now consider a *partially disjunctive* concept, which says that an instance either has features  $F_1$ ,  $F_2$ ,  $F_i$ ,  $F_n$  or  $F_1$ ,  $F_2$ ,  $F_i$ ,  $F_n$ ,  $F_n$  or fatures may differ ( $F_n$  versus  $F'_n$ ). This means there is no set of necessary features that are jointly sufficient:  $F_1$ - $F'_n$  are necessary but not jointly sufficient, while either  $F_1$ - $F_n$  or  $F'_n$ ). This too violates the assumption about defining features the assumption about defining the assumption about definite the state assumption features in common ( $F_1$ - $F_1$ ). This means there is no set of necessary features that are jointly sufficient:  $F_1$ - $F'_n$  are sufficient but include at least one nonnecessary features ( $F_n$  or  $F'_n$ ). This too violates the assumption about defining features.

## NESTING OF FEATURES IN SUBSET RELATIONS

The final representational assumption of interest is as follows: If concept X is a subset of concept Y, the defining features of Y are nested in those of X. It is again convenient to illustrate the assumption with geometric concepts. Suppose that people represent the concept of quadrilateral by two features: closed figure and foursided. These two features are the ones we have included in our previous example of the concept square, along with the features of

equal sides and equal angles. Hence a square is a subset of quadrilateral, and the defining features of quadrilateral are nested in those of square. Similarly, the defining features of bird (for example, animate and feathered) are nested in those of robin, since robins are a subset of birds. Of course the more specific concept – square or robin – must also include some defining features that are not shared by its superset; for example, robin must contain some features that distinguish it from other birds. This guarantees that the representation of a concept cannot be a realizable instance, since the concept must contain fewer features than any of its instances.

Although this nesting assumption is a common one among advocates of the classical view, some would not buy it wholesale. Fodor (1975) in particular questions the assumption, and suggests instead that related concepts may be defined by different sets of features. For example, the feature of bird that specifies "feathered" may not be identical to any specific feature of chicken. If we accept this possibility, the classical view is considerably weakened in the claims it makes about concepts. Given this, for the time being we opt for the version of the view that includes assumption 3.

#### SUMMARY

The three assumptions of the classical view are summarized in Table 1. Although they are not the only assumptions used by proponents of the classical view, they are the modal ones. Indeed, they are presupposed by most of the significant psychological work done on artificial concepts from 1920 to 1970 (for reviews, see Bruner, Goodnow, and Austin, 1956; Bourne, 1966; Bourne, Dominowski, and Loftus, 1979).

TABLE 1 THREE ASSUMPTIONS OF THE CLASSICAL VIEW

1. Summary representation

2. Necessary and sufficient (defining) features

3. Nesting of concept's defining features in it subsets

One last point: the three assumptions say nothing about possible relations between features—that is, the features are treated as if they were independent. This treatment may be adequate for certain semantic domains, called *paradigms*; an example would be kinship concepts, like mother, father, son, and daughter. Here the features—sex and age—seem to combine as independent entities. However, the idea of independent features does not fit with other

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semantic domains, called *taxonomies*; an example would be animal concepts, like robin, bird, animal, and organism. Here the features are clearly related; for example, the feature "living" (defining for organism) is implied by the feature "animate" (defining for animal). Though such relations are not mentioned in our three assumptions, we do not mean to exclude them from the classical view. Rather, we are trying to keep the assumptions down to a minimum, agreed-upon set. Even this small set will soon be shown to contain a great deal of debatable content.

## General Criticisms of the Classical View

Throughout the years there have been various general criticisms of the assumptions of the classical view. In what follows we consider four criticisms that seem especially widespread, along with possible rebuttals.

## FUNCTIONAL FEATURES

Some have argued as follows:

- 1. The classical view deals only with structural features fixed properties (of varying perceptibility) that describe an entity in isolation, like the handle or concavity of a cup and prohibits functional features, like the fact that a cup is used to hold something.
- 2. But for many concepts, particularly those corresponding to human artifacts like cups and chairs, the defining features are functional ones.
- 3. Therefore, the classical view cannot handle all concepts.

Cassirer (1923) put forth this argument some time ago, and Nelson (1974) and Anglin (1977) have recently reiterated it and suggested that it is devastating to the classical view.

Given our earlier discussion about the need to consider abstract, functional features in concept cores, it should come as no surprise that we think this argument is based on a faulty premise, namely premise 1. Nothing in our three assumptions excludes functional features. A functional feature, such as the fact that a cup can hold liquid, can be used in a summary description of an entire class, can be singly necessary and part of a jointly sufficient set, and can be nested in other feature sets. Furthermore, none of our constraints on features is inconsistent with functional features. A feature like "holdability" can bring out relations between concepts (for example, between cup and bowl), can apply to many different classes, and can be used as an input to categorization processes. In short, our rendition of the classical view is as hospitable to functional features as it is to structural ones.

Why, then, do so many psychologists think the classical view should be restricted to structural features? No doubt because such features are generally more perceptual than functional features. But then why do psychologists think the classical view should be restricted to perceptual features? We discussed one answer to this in the previous section—perceptual features greatly simplify the analysis of how people categorize perceptual objects. Another reason perceptual features have proved so attractive to psychologists is that such features are very easy to manipulate in experimental studies of concept attainment and utilization.

When Hull started his experimental study of the classical view in 1920, he used novel visual forms that were composed of multiple features. This allowed him to control precisely which features occurred in all instances of a concept, that is, which features were necessary. Had he used more abstract features, like functional ones, Hull would have had either to give his subjects real manipulable objects and let them discover the function (a messy task at best), or to give them pictures of objects that instantiated the function to varying degrees (which again is a relatively uncontrolled paradigm, though very likely a more ecologically valid one). Hull's emphasis on easily manipulable perceptual features proved so attractive that more than two generations of experimental psychologists have bought it, thereby making it seem that perceptual features are part and parcel of the classical view. It is only during the last fifteen years, with the influence of nonexperimental disciplines like linguistics and philosophy on psychology, that psychologists have begun to realize that some concepts may have functional features at their core.

#### DISJUNCTIVE CONCEPTS

A more powerful argument against the classical view is the following:

- 1. The classical view excludes disjunctive concepts.
- 2. But many concepts are clearly disjunctive, like that of a strike in baseball (which can be either a *called* or a *swinging* strike).
- 3. Therefore, the classical view cannot handle all concepts.

Certainly we agree with premise 1, for we noted earlier that the assumption of defining features excludes disjunctive concepts. Premise 2, though, is debatable. Specifically, how widespread are

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disjunctive concepts? Unfortunately, there is nothing resembling a clear-cut answer to this question. If we rely on intuitions (our own and those published by semanticists) and restrict ourselves to concepts about naturally occurring objects (flora and fauna), we can think of no obvious disjunctive concepts. Disjunctive concepts, then, may be rare, restricted to man-made concoctions (like a baseball strike), and constitute special cases that should not obscure the general conjunctive nature of concepts.

This reasoning may be too facile, however. There are alternatives to intuitive analyses of concepts, and at least one of these suggests that disjunctive concepts may be quite widespread. Rosch and her colleagues (1976) asked people to list the features of concepts, where the concepts varied in their level of inclusiveness (for example, kitchen chair, chair, and furniture). Their data suggest that the more inclusive or superordinate concepts may be disjunctive. For superordinate concepts like animal, plant, vehicle, furniture, clothing, and tool, people list few if any features; for concepts that are one level less inclusive, like bird, flower, truck, chair, hat, and hammer (what Rosch and colleagues call the *basic* level), people list a substantial number of features. This finding suggests that superordinate concepts are often disjunctive (and that basiclevel concepts are the most inclusive level at which conjunctive concepts appear).

There is, however, an alternative interpretation of the data compiled by Rosch and her associates, one that can save the classical view from a plethora of disjunctive concepts. The features that people listed may well have been part of the identification procedure, not the core. But then why should identification procedures be disjunctive only for superordinate concepts? The reason is very likely that the cores of superordinate concepts contain abstract features (remember "intended to be worn by a human"), and such features can only be instantiated disjunctively at the perceptual level. Under this interpretation, concept cores are as conjunctive as the classical view claims they are, and those who mistakenly think otherwise have confused the identification procedure with the core. To illustrate further the flavor of this argument, let us consider the concept "extreme." Some might deem it disjunctive because it implies one pole or the other, but this may be an aspect of the identification procedure, not the core, where the latter may mean "a value far from the central tendency." Another example is the concept of split personality: doesn't this mean personality X or personality Y, but not both (an exclusive disjunction)? Perhaps it does at the level of an identification procedure, but the concept core may mean "manifests different personalities," which is not inherently disjunctive. (Anisfeld, 1968, has made a similar argument using the notions of sense and reference.)

The upshot is that we have no firm evidence, intuitive or otherwise, about the prevalence of disjunctive concepts. Without such evidence, it is difficult to say how damaging the disjunctiveconcepts argument is to the classical view.

#### UNCLEAR CASES

A third argument against the classical view (see, for example, Hampton, 1979) takes the following form:

- 1. The classical view assumes that if concept X is a subset of concept Y, the defining features of Y are nested in those of X.
- 2. Given this, judgments about whether one concept is a subset of another should be clear-cut, since one merely has to compare defining features.
- 3. In fact, it is often unclear whether one concept is a subset of another. People disagree with one another about a particular subset relation, and the same person may even change his mind when asked the same question on different occasions (see McCloskey and Glucksberg, 1978). The classical view has no way of accounting for such unclear cases.

The weak part of this argument is premise 2, since a nesting of defining features does not guarantee that judgments about subset relations will be clear-cut. We can think of at least two reasons why this is so, and it is best to illustrate them by a specific example.

When asked, "Is a tomato a fruit?" many people, even collegeeducated ones, are unsure of whether this particular subset relation holds. One simple reason they may be uncertain is that their concepts of tomato and fruit may be faulty or incomplete-that is, they are missing some defining features of fruit and consequently cannot tell whether or not a tomato is a fruit. To put it more generally, the classical view does not stipulate that every adult has mastered every familiar concept; rather, it allows for the possibility that many of us are walking around with incomplete concepts, just as long as whatever features we do have are at least necessary ones. (Such incomplete concepts could not be too incomplete, however, since adults obviously do a good job of using their concepts in dealing with their environment.) A second way to reconcile the classical view with unclear cases is to assume that some concepts have two definitions, a common and a technical one (Glass and Holyoak, 1975). Thus one might be unsure about what concept a tomato belongs to because a tomato meets the technical definition of a fruit

(for example, it has seeds) but the common definition of a vegetable (it plays a particular role in meals).

Specifying the Defining Features of Concepts

Of all the arguments against the classical view, the best-known one goes as follows:

- 1. The heart of the classical view is its assumption that every concept has a set of necessary and sufficient features.
- 2. Decades of analysis have failed to turn up the defining features of many concepts.
- 3. Therefore, many concepts simply dó not have defining features.

It was essentially this argument that Wittgenstein (1953) pursued in his well-known critique of a classical-view approach to natural concepts. One of Wittgenstein's most famous examples was that of the concept of games, and we can use it to illustrate the flavor of his argument. What is a necessary feature of the concept of games? It cannot be competition between teams, or even the stipulation that there must be at least two individuals involved, for solitaire is a game that has neither feature. Similarly, a game cannot be defined as something that must have a winner, for the child's game of ringaround-a-rosy has no such feature. Or let us try a more abstract feature-say that anything is a game if it provides amusement or diversion. Football is clearly a game, but it is doubtful that professional football players consider their Sunday endeavors as amusing or diverting. And even if they do, and if amusement is a necessary feature of a game, that alone cannot be sufficient, for whistling can also be an amusement and no one would consider it a game. This is the kind of analysis that led Wittgenstein to his disillusionment with the classical view.

Although this argument clearly has merit, it is by no means ironclad, for its conclusion – that many concepts do not have defining features – is based on a lack of progress by the classical view. When Wittgenstein – or anyone else – asserts: "There are no defining features of concept X," it is equivalent to asserting: "No one has yet determined the defining features of concept X," since both assertions would be refuted by a cogent proposal of these features. Moreover, one could claim that part of the reason progress has been so slow is that we have been looking for the wrong kind of defining features – perceptual ones that are likely to be part of an identification procedure – when we should have been seeking abstract, relational, or functional features that may well make up the core of many concepts. Thus the Wittgenstein argument is nothing like a principled disproof of the classical view; it is instead an empirical argument about the observed rate of progress of a theoretical approach to concepts. Once this is appreciated, one can acknowledge that the argument certainly has force (like any excellent empirical argument) but that it deals no death blow to the classical view.<sup>1</sup>

A NOTE ON SCIENTIFIC CONCEPTS

It is worth pointing out that the last two criticisms of the classical view of psychological concepts – unclear cases and failure to specify defining features – have also been raised as criticisms of the classical view when it is used as a metatheory of scientific concepts. That is, in addition to its use as a psychological theory, the classical view has also served as a metatheoretical prescription of what scientific concepts should look like, and here it has run into problems similar to those we just described.

There are numerous unclear cases for classically defined biological concepts. For example, there is no uniform agreement among biologists as to whether *Euglena*, a mobile organism that manufactures chlorophyll, should be classified as an animal or a plant. Cases like this are occurring with sufficient frequency to lead scientists to question the validity of the classical view for biological classification (see Sokal, 1974).

Similarly, there has been substantial difficulty in specifying the \*defining features of biological species, at least in terms of structural features (Sokal, 1974; Simpson, 1961). Toward the end of the eighteenth century Linnaeus proposed that any biological species can be characterized by three kinds of features: (1) features that comprise the essence of the species, which are features that every member of the species must have and that correspond to what we have called defining features; (2) features called properties, which are common to all members of the species but are not part of the essence; and (3) features called accidents, which characterize some but not all members of a species. According to Linnaeus, only features comprising the essence should be used in classification. This classicalview approach has had great influence, but it now seems problematic as a guide to biological classification. For one thing, taxonomists have generally been unable to distinguish features comprising the essence from those called properties. For another, taxonomists have found that the so-called accidents, features not true of every species member, are sometimes genetically based and important for understanding and defining the species.

These developments in biological classification are relevant to a psychology of concepts. Recall that in the Introduction we noted that there was little hope for classically defined mental representa-

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tions if there was little evidence that such concepts could be given a classical definition in some language. The most likely place to look for classical definitions of flora and fauna is the language of biology, and to the extent that the classical view fails here, it will likely fail as a psychological theory as well.

## Summary

Table 2 summarizes the four general criticisms we have discussed. How badly do they damage the classical view? In answering this, we must distinguish between an *in principle* criticism – one that shows that the view could never handle a particular problem – and an *empirical* criticism – one that shows that specific embodiments of the view have thus far failed to handle a particular problem. All four criticisms seem to be mainly empirical ones.

TABLE 2 FOUR GENERAL CRITICISMS OF THE CLASSICAL VIEW

1. Exclusion of functional features

2. Existence of disjunctive concepts

3. Existence of unclear cases

4. Failure to specify defining features

The first criticism – that the classical view excludes functional features – is clearly about typical applications of the view in psychology, and not about what this view can accomplish in principle. We showed that the assumptions of the classical view are as compatible with functional features as they are with structural ones. How successfully one can use the view with functional features, however, remains an open question. The work of Miller and Johnson-Laird (1976) at least suggests that one can develop a classical-view model of categorization that employs functional features.

The second criticism – that the view excludes disjunctive concepts – comes closest to offering an in-principle argument against the classical view. If some natural concepts are clearly shown to be disjunctive, they simply fall outside the domain of the classical view. Such convincing demonstrations, though, have been fare.

With regard to the third criticism, we argued that unclear cases are not necessarily inconsistent with a classical-view description of natural concepts because people may have incomplete, or multiple, definitions of a concept. Again, the criticism is hardly a proof against the classical view.

Finally, there is the criticism that the classical view has made lit-

tle progress in specifying defining features. In discussing this criticism, we emphasized its empirical nature—it is a statement about what has happened so far, not about what can happen. Still, as an empirical criticism, it is one of the strongest arguments against the classical view.

We emphasize the empirical nature of these criticisms because we wish to dispel the popular notion that the classical view has been proved wrong by a priori arguments and consequently that no empirical work is needed. A more correct reading of the situation is this: serious empirical criticisms have been raised against the classical view of natural concepts – serious enough to make us have grave reservations about the view, but not serious enough to say that the view should be discarded at this point.

## Experimental Criticisms of the Classical View

Though the general criticisms discussed above are telling, they are not the only reasons why psychologists are currently forsaking the classical view in droves. There are other criticisms of this view that stem from experimental findings about how people use natural concepts, such as how they decide that apples are fruit. Before delving into these findings, we would like to interject a cautionary note. Since the findings deal with how people use concepts, they reflect categorization processes as well as concept representations. This means that we cannot go directly from the findings to claims about how concepts are represented; instead, we must interpret these findings in terms of both representations and processes - in short, in terms of models. This point has been missed in a good deal of recent research on natural concepts, where it has often been assumed that categorization data directly inform us about the nature of concepts. The best way to document the need for a model in interpreting categorization effects is to consider some results of interest and then show that their implications for the classical view depend on the specific model used to instantiate this view. This is the procedure we adopt in the following discussion.

#### SIMPLE TYPICALITY EFFECTS

#### Experimental Results

Of all the experimental findings used as evidence against the classical view, perhaps the best known are the effects of *typicality* (also called *prototypicality*). The most critical result is that items judged to be typical members of a concept can be categorized more efficiently than items judged to be less typical. The details of this result are as follows: People find it a natural task to rate the various

| TABLE 3. TYPICALI | TY RATINGS FOR BIRD |
|-------------------|---------------------|
|-------------------|---------------------|

AND MANDEAL INCLANCES

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| AND MAMMAL INSTANCES |        |          |        |  |
|----------------------|--------|----------|--------|--|
| Instance             | Rating | Instance | Rating |  |
| Robin                | 3.00*  | Deer     | 2.83   |  |
| Sparrow              | 3.00   | Horse    | 2.76   |  |
| Bluejay              | 2.92   | Goat     | 2.75   |  |
| Parakeet             | 2.83   | Cat      | 2.67   |  |
| Pigeon               | 2.83   | Dog      | 2.67   |  |
| Eagle                | 2.75   | Lion     | 2.67   |  |
| Cardinal             | 2.67   | Cow      | 2.58   |  |
| Hawk                 | 2.67   | Bear     | 2.58   |  |
| Parrot               | 2.58   | Rabbit   | 2.58   |  |
| Chicken              | 2.00   | Sheep    | 2.58   |  |
| Duck                 | 2.00   | Mouse    | 2.25   |  |
| Goose                | 2.00   | Pig      | 2.17   |  |

Source: After Rips, Shoben, and Smith (1973).

a. Higher numbers indicate greater typicality.

subsets or members of a concept with respect to how typical or representative each one is of a concept.<sup>2</sup> Such ratings were first reported by Rips, Shoben, and Smith (1973) and by Rosch (1973). Table 3 presents the ratings of Rips, Shoben, and Smith for the concepts of birds and mammals. As can be seen, robin and sparrow are considered typical birds, hawk and eagle less typical, and chicken and penguin atypical. Ratings like these have now been obtained for many noun categories in English, and have been shown to be highly reliable across raters (Rosch, 1973) and to be relatively uncorrelated with frequency or familiarity (Mervis, Catlin, and Rosch, 1976).

What is most important about these ratings is that they predict how efficiently people can categorize the various members of a concept in a semantic categorization task. One variant of this task is illustrated in Figure 3. On each trial, the subject is given the name of a target concept, like bird, followed by a test item; the subject decides whether the test item names a subset or member of the target concept, like robin or chicken, or a nonmember, like dog or pencil. The main data of interest are the times for correct categorizations. When the test item, or probe, in fact names a member of the target concept, categorization times decrease with the typicality of the probe (see Rips, Shoben, and Smith, 1973; Rosch, 1973). For example, when bird is the target concept, test items corresponding to robin and sparrow are categorized more quickly than those corresponding to eagle and hawk, which in turn are categorized faster

| 1000 | $\sim$ |         | 2.71 |
|------|--------|---------|------|
| Ine  | L      | assical | View |

| b.;                | Target<br>Concept | Test<br>Item | Correct<br>Response |
|--------------------|-------------------|--------------|---------------------|
| Trial <u>n</u>     | Bird              | Robin        | Yes                 |
| Trial <u>n + l</u> | Fruit             | Сир          | No                  |
| Trial <u>n + 2</u> | Bird              | Chicken      | Yes                 |

**Figure 3** One variant of semantic categorization task (other variants require verification of sentences of form "A robin is a bird" or "All apples are fruit")

than the probes of chicken and goose. Furthermore, to the extent that there is any variation in the accuracy of these categorizations, error rates also decrease with the typicality of the probe (see Rips, Shoben, and Smith, 1973). These effects are extremely reliable: they have been documented in more than 25 experiments that have used many different variants of the semantic categorization task (see Smith, 1978, for a partial review).

Though most studies of typicality have been concerned with categorization times, Rosch and Mervis have demonstrated a host of other typicality effects. For instance, the typical members of a concept are the first ones learned by children, as judged by either a semantic categorization task (Rosch, 1973) or by how accurately children can sort pictured objects into taxonomic categories (Mervis, 1980). (The latter finding is not strictly relevant to issues about concept cores.) Further, the typical members of a concept are likely to be named first when subjects are asked to produce all members of a category (Mervis, Catlin, and Rosch, 1976). And typical members are also more likely to serve as cognitive reference points than are atypical members (Rosch, 1975); for example, people are more likely to say "An ellipse is almost a circle" (where circle, the more typical form, occurs in the reference position of the sentence) than "A circle is almost an ellipse" (where *ellipse*, the less typical form, occurs in the reference position). This list of effects could be extended (see, for example, Rosch, 1974, 1975, 1978), but it is adequate for our purposes.

What does all this have to do with the classical view? Simply this: typicality effects reveal that not all members of a concept are equal, or to put it more positively, that concepts possess an internal structure that favors typical members over less typical ones. The representational assumptions of the classical view, however, sug-

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gest that all members of a concept are equal, since all members of concept X must have the defining features of X. At first glance, then, typicality effects seem incompatible with the classical view, a conclusion that has been drawn many times. A more thorough analysis is needed here, however—one that considers processes as well as representations.

## A Classical-View Model for Simple Typicality Effects

Since we need to interpret categorization results in terms of process models, the real question for the classical view is this: Can a model based on this view (that is, one that incorporates its three representational assumptions) account for the typicality effects we have described? The answer is clearly yes, and we will describe such a model in this section. We caution the reader, however, that the model presented will run into problems when later confronted with other findings; our reason for describing the model here is to demonstrate how easy it is to come up with a classical-view model that can account for effects frequently claimed to be inconsistent with this view.

We call our proposal the *complexity model*. It assumes that concepts are represented just as the classical view says they are, and that in a categorization task these representations are processed by two sequentially ordered stages, the *access* and *comparison* stages. The stages operate as follows:

- 1. When given the target and probe concepts, the subject starts accessing the defining features of both concepts, with access order being random over trials.
- 2. As soon as any defining features are available, the subject compares those of the target concept to those of the probe. The subject responds affirmatively ("Yes, it's a member") only when every feature of the target has matched a feature of the probe, but can respond negatively ("No, it's not a member") as soon as any feature of the target mismatches a probe feature. This stage is limited in capacity, and therefore the time needed to compare probe and target concepts increases with the number of features in either concept.

Only one more assumption is needed, and it is the critical one: Typical members of a concept have fewer features than atypical ones; that is, typicality is in inverse measure of complexity. Figure 4 illustrates this idea. In the figure, features are assigned to concepts in accordance with the classical view; for example, the defining features of bird are contained in the defining features of robin and 37

| <u>Animal</u> | Bird           | Robin          | Chicken        |
|---------------|----------------|----------------|----------------|
| F             | F <sub>1</sub> | Fi             | F,             |
|               | F2             | F2             | F2             |
|               |                | F <sub>3</sub> | F <sub>4</sub> |
|               |                |                | $F_5$          |

Figure 4. Concept representations in a complexity model

chicken. In addition, chicken is assumed to contain more of its own defining features – those that distinguish it from other species of birds – than does robin. This is in keeping with our critical assumption.

These assumptions suffice to explain all simple typicality effects. Since atypical probes contain more features than typical ones, atypical probes will require longer comparison stages, and consequently they will be categorized more slowly. Furthermore, because atypical probes require more comparisons than typical ones, they are more likely to lead to an error (assuming each comparison has some fixed probability of being in error). The complexity model also provides a reasonable explanation of the other typicality effects reported by Rosch and Mervis. The fact that typical concepts are learned before atypical ones becomes just another example of simple concepts being mastered before complex ones (see Brown, 1973). And the fact that typical concepts serve as reference points may just indicate our preference for simple concepts as anchors. Then there is the question of how this model would interpret the typicality ratings themselves. The simplest possibility is that subjects rate the similarity of the probe to the target concept, with similarity (1) increasing with the number of features shared by probe and target and (2) decreasing with the number of probe features not present in the target (Tversky, 1977). Now, (1) is constant across all members of a target concept (each contains all the defining features of the target), but (2) must increase with the number of defining features in a probe; it follows that typical probes, which contain fewer of their own defining features, will be judged more similar to the target concept and hence rated more typical. The fact that typical items are more similar to their parent concepts also accounts for the one remaining

simple typicality effect – when given a concept and asked to produce its instances, a subject will name typical members first. The concept is essentially a memory probe, and research on memory retrieval indicates that items similar to the probe are retrieved first (see Tulving, 1974).

Despite what we have just discussed, we have no faith in the complexity model. For one thing, it is inconsistent with the finding that it takes no longer to respond to atypical than to typical probes when the probe is not a member of the target concept. For example, it takes no longer to disconfirm "a chicken is a fish" than "a robin is a fish," even though chicken supposedly has more features than robin and consequently should require more comparison time (see Smith, Shoben, and Rips, 1974). Another problem for the complexity model comes from studies in which subjects are asked to list features of various members of a concept. Such studies have found either no difference in the number of features listed for typical versus atypical members, or that more features are listed for typical members (Ashcraft, 1978; Malt and Smith, 1981a). To the extent that the listed features correspond to the true core features of the instances involved, these findings contradict the complexity model's assumption that atypical members have more features than typical ones. To reiterate, our point in presenting the complexity model was merely to show that simple typicality effects can readily be accounted for by a model based on the classical view.<sup>3</sup>

# Determinants of Typicality: Family Resemblance Measures

#### **Experimental Results**

In addition to research showing the effects of typicality on various measures of performance, there have been some studies that have tried to specify the determinants of typicality. The most important of these is Rosch and Mervis's work on family resemblance (1975). We first present their results and then take up the question of whether these findings are incompatible with the classical view, as has often been claimed.

Some of Rosch and Mervis's experiments used natural concepts. The subjects were asked to list features of various subsets of a superordinate concept, like those of furniture, where the subsets varied in typicality (table is typical, lamp atypical). Rosch and Mervis showed that the distribution of listed features could provide a basis for typicality. Their analysis is illustrated in Table 4. Each feature listed for a subset is weighted by the total number of subsets that it is listed for; then, for each subset, the weights of all of its

| TABLE 4 FAMILY RESEMBLA | NCE ANALYSIS |
|-------------------------|--------------|
|-------------------------|--------------|

| Article of<br>furniture |                     | Listed             | features           |                    | Family resemblance<br>measure |
|-------------------------|---------------------|--------------------|--------------------|--------------------|-------------------------------|
| Chair                   | F <sub>1</sub> (5)" | F <sub>2</sub> (4) | F <sub>3</sub> (3) | F <sub>4</sub> (2) | 14                            |
| Sofa                    | F,(5)               | F <sub>2</sub> (4) | F <sub>3</sub> (3) | F <sub>s</sub> (2) | 14                            |
| Cushion                 | F,(5)               | F <sub>2</sub> (4) | F <sub>6</sub> (1) | F,(2)              | 12                            |
| Rug                     | F <sub>1</sub> (5)  | F <sub>3</sub> (3) | F,(2)              | $F_s(1)$           | 11                            |
| Vase                    | F <sub>1</sub> (5)  | F <sub>5</sub> (2) | $F_{9}(1)$         | $F_{t0}(1)$        | 9                             |
| Telephone               | F <sub>2</sub> (4)  | F4(2)              | F.,(1)             | $F_{12}(1)$        | 8                             |

Source: After Rosch and Mervis (1975).

a. Numbers in parentheses indicate how often each feature occurs in set of instances.

features are summed, yielding a measure called *family* resemblance. In Rosch and Mervis's study, these family resemblance measures were very highly correlated with typicality ratings of the subsets. In short, an item is a typical subset or member or a concept if it contains features shared by many other members of that same concept.

To back up this conclusion, Rosch and Mervis performed experiments with artificial concepts. In these experiments, subjects learned to assign visually presented letter strings to categories, with six strings belonging to one category and six to another. Table 5 illustrates two of the categories used. Each string can be treated as an instance of a concept, each letter in a string as a feature. Since no information but the letters was available to subjects, we may assume that the cores of the concepts were restricted to the letters. (This kind of assumption – that the core is restricted to the obvious perceptual features - is standard in studies of artificial concepts.) Note that the strings were constructed so that no feature (letter) was common to all instances of a concept, that is, there were no salient defining features for the concept. Though all strings in Table 5 contain the same number of letters, they vary with respect to their family resemblance scores; for example, a high family resemblance string like .- IMQB contains letters that were usually shared by other strings in its category, while a low family resemblance string like IXPHM contains letters less likely to be shared by other strings. The results showed that the higher the family resemblance score of a string, the sooner it could be learned as a concept instance, the more quickly it could be categorized once learned, and the more typical it was rated of its concept. Thus an experimental manipulation in the distribution of features-that is, in family resemblance – produced many of the simple typicality effects found

| LABLE 5 | ARTIFICIAL CATEGORIES | USED BY | r Rosch | AND | Mervis | (1975) |
|---------|-----------------------|---------|---------|-----|--------|--------|
|---------|-----------------------|---------|---------|-----|--------|--------|

| Category   | Letter string | Family resemblance<br>measure |
|------------|---------------|-------------------------------|
|            |               | 111643016                     |
| Category A | JXPHM         | 15                            |
|            | XPHMQ         | 19                            |
|            | PHMQB         | 21                            |
|            | HMOBL         | 21                            |
|            | MQBLF         | 19                            |
|            | QBLFS         | 15                            |
| Category B | GVRTC         | 15                            |
|            | VRTCS         | 19                            |
|            | RTCSF         | 21                            |
|            | TCSFL         | 21                            |
|            | CSFLB         | 19                            |
|            | SFLBQ         | 15                            |

with natural concepts. This is solid evidence that the distribution of features is the cause of simple typicality effects. Rosch, Simpson, and Miller (1976) have replicated some of these important findings.

## Implications for the Classical View

What exactly are the implications of these results for the classical view? From the results with natural concepts, one could make the following argument.

- 1. The typicality variations observed when people categorize members of a superordinate concept are highly correlated with variations in family resemblance scores of the members.
- 2. The variations in family resemblance scores are due to features that are not common to all members. (A common feature would simply add a constant to all family resemblance scores see Tables 4 and 5 and hence could not influence the correlation with typicality.)
- 3. Therefore, typicality variations cannot be explained by variations in the defining features of the superordinate concept.
- 4. Therefore, typicality variations must be accounted for in terms of nondefining features, but the classical view precludes the latter.

We have no quarrel with premises 1 and 2, nor with conclusion

3, but conclusion 4 is fallacious. This is illustrated by Table 6, which is the same as Table 4 save two exceptions: (1) one common feature,  $F_0$ , has been added to all members; this feature is assumed to be defining for the superordinate concept of furniture, and it adds a constant of 6 to the family resemblance scores of all members; and (2) all features listed for a member are assumed to be defining of that subset (for example, chair's defining features are  $F_0$ - $F_4$ ). This example is consistent with the first three steps of the argument just presented but is inconsistent with the critical conclusion that typicality variations *must* be accounted for in terms of nondefining features. For Table 6 shows that, at least in principle, typicality variations can be explained in terms of variations in the defining features of the members; that is, the defining features of the individual members vary in frequency, and this variation may be responsible for the concomitant variation in typicality.

We seem to have saved the classical view from family resemblance. There are problems, however, with our rescue mission. One stems from our assumption that the features listed for concept members are defining ones. Inspection of the features actually listed makes this unlikely. For example, many people list as features of chair "made of wood" and "has four legs." Clearly these features are not true of all chairs, and consequently they are not defining of chair. Other problems arise from our assumption that the superordinate concept, furniture, may be represented by a defining feature. If there is such a feature (or features), why did subjects not list it? (No feature was listed for all instances of furniture.) Perhaps the feature was too abstract for naive subjects to verbalize; perhaps it was too obvious for anyone to mention; but then again perhaps it just wasn't there. Furthermore, this assumption about a defining feature is clearly unnecessary to explain the data of interest. Rosch and Mervis found the same relation between typicality and family resemblance with their artificial concepts, and these concepts were constructed in such a way that they had no obvious defining feature.

The problems mentioned above hinge on whether listed features are valid indicators of the true features of concepts. Putting this issue aside, there are further problems when we try to construct a classical-view model for how the representations in Table 6 could be processed so as to yield a correlation between family resemblance scores and categorization time (the latter being known to correlate with typicality). The complexity model won't do. It holds that the critical factor is the number of features in a concept member, whereas the data show that the critical factor is the distribution of the member's features. Let us try to construct another model of the same sort, that is, one in which the features of the probe and

| Table ó | How family resemblance measures can  |
|---------|--------------------------------------|
|         | BE CONSISTENT WITH DEFINING FEATURES |

| Article of<br>furniture |                     |                    |                    |                    |                    | Family resemblance<br>measure |
|-------------------------|---------------------|--------------------|--------------------|--------------------|--------------------|-------------------------------|
| Chair                   | F <sub>0</sub> (6)" | F.(5)              | F <sub>2</sub> (4) | F <sub>3</sub> (3) | F <sub>4</sub> (2) | 20                            |
| Sofa                    | F₀(6)               | F <sub>1</sub> (5) | F <sub>2</sub> (4) | F <sub>3</sub> (3) | F <sub>5</sub> (2) | 20                            |
| Cushion                 | F <sub>0</sub> (6)  | F <sub>1</sub> (5) | $F_{2}(4)$         | F <sub>6</sub> (1) | F,(2)              | 18                            |
| Rug                     | F₀(6)               | F <sub>1</sub> (5) | F <sub>3</sub> (3) | F <sub>7</sub> (2) | $F_{a}(1)$         | 17                            |
| Vase                    | F <sub>0</sub> (6)  | F.(5)              | F <sub>5</sub> (2) | F <sub>9</sub> (1) | $F_{10}(1)$        | 15                            |
| Telephone               | F <sub>0</sub> (6)  | F <sub>2</sub> (4) | F4(2)              | F.,(1)             | $F_{12}(1)$        | 14                            |

a. Numbers in parentheses indicate how often each feature occurs in set of instances.

target concepts are accessed and compared, with the comparison process starting as soon as any features are available and continuing until all the features of the target have been matched or at least one has been mismatched.

In this kind of model the distribution of members' features could affect either the access or comparison processes. Both possibilities have their problems. Consider first the possibility that the access process is affected. Since a positive categorization must be based on retrieval of  $F_0$  from this probe (it is the only feature that matches the defining feature of the superordinate concept; see Table 6), we seek a model in which the access time for  $F_0$  can be affected by the access times for the other features in the probe concept. Such a model is embodied in the following assumptions:

- 1. Assume that the processing capacity for accessing features is limited.
- 2. Assume further that the amount of capacity needed to access a particular feature decreases with the frequency of that feature in concept members. This means that the amount of capacity needed to access features that define typical members (like  $F_1$  and  $F_2$ ) will be less than that needed to access features that define atypical members (like  $F_{11}$  and  $F_{12}$ ).
- 3. Therefore, more capacity can be devoted to accessing F<sub>0</sub> when it occurs in a typical probe than in an typical one.
- 4. Therefore,  $F_0$  should be accessed faster in typical than in atypical members, which in turn implies that typical members should be categorized faster.

## The Classical View

Though this model accounts for the typicality effects associated with Table 6, it has a serious problem. The model implies that the fewer features a concept member contains, the faster it will be categorized (the fewer the features, the more capacity can be devoted to accessing each one); but Malt and Smith (1981a) found that those concept members that had a minimal number of features (as determined by attribute listings) are categorized slowest of all.

A similar situation results if we try to construct a model in which the distribution of members' features affects the comparison process. Now we would assume that the capacity needed to compare a feature decreases with the frequency of the feature in concept members. Consequently, less capacity will be needed for comparing features that define typical members, which means that more comparisoncapacity can be devoted to  $F_0$  (the feature that defines the superordinate) when it occurs in typical members, which in turn implies that typical members should be categorized faster than atypical ones. Again, though, this model erroneously predicts that the fewer features a concept member contains, the faster it will be categorized.

What this shows is that some simple classical models cannot explain the relation between certain typicality effects and family resemblance scores. We can construct more complex classical models that will do the job, but the ones we have tried all require some ad hoc assumptions (like the one that the common feature,  $F_0$ , though it occurs very frequently, is so complex that it is processed slower than features that occur less frequently). In short, we feel that if the Rosch and Mervis results are to be explained by the classical view, ad hoc processing assumptions are needed, which reflects badly on the view.

To summarize all we have said about the Rosch and Mervis results, the critical result is that typicality variations are due to the distribution of features of concept members. This by itself is not, in principle, inconsistent with the classical view, as is demonstrated by our example in Table 6. But to make the result consistent with the view, we had to make some precarious assumptions about the relations between the listed features and the true features of the concepts. To account for the results in terms of a classical model, we would have to make even more precarious assumptions. Like other findings we have covered and will cover, the present results do not decimate the classical view, but they do provide reasons for lessening our belief in it.

Use of Nonnecessary Features

Though Rosch and Mervis's work (1975) centered on an explanation of typicality effects, we noted that one of its main implications

#### **Categories and Concepts**

for the classical view involved the possible use of nondefining features. The work we now wish to discuss offers a more direct approach to the issue of nondefining features: it specifically tries to show that people use nonnecessary features in categorization.

A good example of this work is a study carried out by Hampton (1979). One group of subjects listed features that characterized concepts like bird, fruit, tool, and so on. Next, they rated the extent to which subsets of these concepts had the features listed. For example, if subjects listed "flies" for bird, they might specify that robin has this feature while chicken does not. These ratings were then used to predict categorization times for another group of subjects. The more features that were shared by a concept and one of its members, the faster that member could be categorized; that is, the number of shared features between member and concept was a good measure of the typicality of that concept member. There are two critical points:

- 1. Some features listed for a concept were nonnecessary ones (for example, "flies" for bird).
- These nonnecessary features were correlated with categorization performance.

It follows by a correlational syllogism that

3. Nonnecessary features are used in categorization.

While this argument is similar to the one ascribed to Rosch and Mervis (1975), it is stronger because it involves a direct assessment of the nonnecessary features of the concept.

The conclusion from the argument given above is difficult to reconcile with categorization models based on the classical view. It is clearly incompatible with the complexity model, which assumes that when one has to decide whether a probe names an instance of a target concept, only necessary features of the target concept are considered. Indeed, any classical-view model that restricts itself to necessary features (an obvious restriction) must be incompatible with the use of nonnecessary features in categorization.

How can the classical view get around this argument? The weak point in Hampton's experiment, of course, is the assumption that the listed features correspond exactly to the true defining features of a concept, a point noted in our earlier discussion of Rosch and Mervis. It seems most unlikely that just anyone off the street can readily list the features of a concept, particularly since such a list may be sensitive to contextual factors, and since sophisticated semanticists have been unable to compose lists of defining features after decades of study. Thus a proponent of the classical view might argue that the "features" listed for a particular concept are simply an epiphenomenon in the following sense:

- 1. Categorization is based on defining features that are relatively inaccessible, or at least difficult to report or introspect on.
- 2. These defining features, however, are correlated with other nonnecessary features that are relatively easy to report.
- 3. Consequently, the corrrelation between categorization and nonnecessary features is being mediated by defining features.

Though this rebuttal is legitimate, it is based on the assumption that unspecifiable defining features just happen to be correlated with specifiable undefining ones. The rebuttal thus capitalizes on its own ignorance of defining features—a rather dubious state of affairs, and one that makes us take the findings on nonnecessary features as a serious problem for the classical view.

There is another source of evidence for the use of nonnecessary features in categorization that does not require subjects to list features. This source is based on multidimensional scaling studies (for example, Rips, Shoben, and Smith, 1973; Caramazza, Hersch, and Torgerson, 1976; Shoben, 1976). Subjects are given pairs of concepts from a particular domain, like robin-sparrow and robin-hawk, where each pair contains two subsets of a generic concept (birds); they are also given pairs that include the concept and one subset (for example, robin-bird, hawk-bird). The task is to rate each pair for its similarity of meaning. These ratings then become input to a scaling program whose output is a geometric space. The points in the space represent the items involved, while the distance between any pair of points reflects the dissimilarity between the two items. Figure 5 illustrates such a space for the concept of bird and 12 of its subsets. The representation seems a reasonable one, since similar birds are close together (for example, hawk and eagle) while dissimilar ones fall far apart (robin and goose).

There are two critical points about this multidimensional space:

- The horizontal dimension appears to reflect variations in size while the vertical one depicts predatory relations or ferocity, where neither of these dimensions specifies anything necessary about being a bird or being any particular subset of a bird.
- 2. The distance between bird and any of its subsets correlates highly with how long it takes to categorize that subset as a bird.

Again it follows by a correlational syllogism that

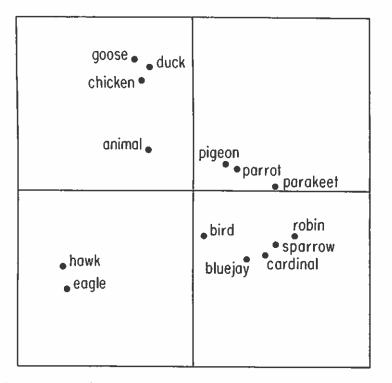


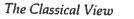
Figure 5 Multidimensional space for bird and 12 of its subsets

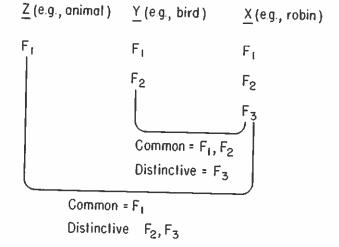
3. Nonnecessary properties are being used in categorization. (The properties here take the form of dimensions rather than features, but recall that any dimension can be represented by a set of nested features.)

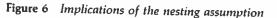
Though one can raise arguments against taking these results at face value (see, for example, Clark and Clark, 1977, chap. 11; Tversky, 1977), the nature of such arguments differs from those raised against the feature-listing studies. In short, accepting that nonnecessary features are used in categorization is beginning to seem more parsimonious than accepting the arguments needed to salvage the classical view.

## NESTED CONCEPTS

A final problem for the classical view stems from its third assumption: If concept X is a subset of concept Y, the defining features of Y are nested in those of X. Figure 6 shows the implica-







tions of this assumption for a nested triple of concepts X, Y, and Z, where X is a subset of Y and Y a subset of Z. As is clear from the figure, X and Y have more common features and fewer distinctive ones than X and Z, which implies that X should be judged more similar to Y than to Z (Tversky, 1977). Thus the third assumption of the classical view implies that a subset (for example, robin) should always be judged more similar to an immediate superor-dinate (for example, bird) than to a distant one (for example, animal). That this prediction has sometimes failed is a major problem for the classical view.

To judge the severity of this failing, we need to consider some specific results. One indirect way to measure the similarity between a subset and its superordinates is to give the subset to a group of subjects and ask them to produce its superordinates. The frequency with which a particular superordinate is produced is then a measure of its similarity to the subset. To illustrate, if the subset is rose, and 18 of 20 subjects produce flower as a superordinate, while only 10 produce plant, then rose is more similar to flower than to plant. This technique was used by Loftus and Sheff (1971), and their results showed that a subset was no more likely to produce its immediate superordinate than a distant one. Taken at face value, this finding is at odds with the classical view (see Smith, Shoben, and Rips, 1974).

As usual, there is good reason not to take the results at face

value. The production frequency of a superordinate term probably depends not only on its similarity to the subset but on its general accessibility as well. Thus a better way to measure the similarity of a subset to its superordinates is by direct similarity ratings, that is, by having subjects give numerical ratings of the similarity of a subset to both its immediate and its distant superordinates. Numerous studies have used this technique (Rips, Shoben, and Smith, 1973; Smith, Shoben, and Rips, 1974; McCloskey, 1980; Roth and Shoben, 1980), and it was generally found that the majority of subsets are rated as more similar to their immediate than to their distant superordinates. There are some exceptions to this finding, however; some of them might be due to the use of unfamiliar superordinates such as alloy and mammal (McCloskey, 1980), but others cannot be explained away by familiarity. Thus chicken and duck are consistently rated as more similar to animal than to bird, and these seem to be clear-cut counterexamples of the classical view's prediction that a subset is more similar to its immediate than its distant superordinates. The bottom line is that although the classical view's prediction works in most cases, it does not work in all.

In addition to the similarity problem, the use of nested triples of concepts in categorization studies had led to another difficulty for the classical view. The problem is that classical-view models, like our complexity one, would predict that a probe concept should be categorized faster when the target concept is a distant superordinate than when it is an immediate one. For example, robin should be categorized faster as an animal than as a bird. But this prediction has often been disconfirmed. The reasoning behind the critical prediction is as follows. For a nested triple, the distant superordinate must contain fewer features than the immediate one - for example, animal has few features than bird-which is just the third assumption of the classical view at work again. And the fewer features there are in the target concept, the fewer must be compared in the comparison stage of our classical-view model, and the less time is needed to decide that the probe concept is indeed a member of the target. This prediction falls out of any classical-view model that assumes categorization is based on a limited-capacity comparison of probe and target features.

We will briefly summarize the experimental literature on this point. Early studies showed that categorizations were faster when the target was an immediate than a distant superordinate (Landauer and Freedman, 1968; Collins and Quillian, 1969; Meyer, 1970). This finding directly contradicts classical-view models of the sort described above. Later studies, however, found few consistent

## The Classical View

effects (Smith, Shoben, and Rips, 1974). And the most recent experiments show that in a majority of cases categorizations are faster with immediate than distant superordinates, though there are some exceptions (for example, Roth and Shoben, 1980). The exceptions turn out to be just those cases where the instance was rated as more similar to its distant than its immediate superordinate. For example, it takes longer to categorize chicken as a bird (an immediate superordinate) than as an animal (a distant superordinate).

We can summarize the foregoing discussion by two critical points:

- 1. With respect to similarity judgments, the classical view predicts an advantage (higher similarity ratings) for immediate over distant superordinates; this prediction works for a majority of cases, but there are definite exceptions.
- 2. With respect to categorization times, straightforward models based on the classical view predict an advantage (faster times) for distant over immediate superordinates; this prediction fails for a majority of cases, but works in a minority.

We can see no way to overcome point 1. It rests on the notion that similarity increases with common features and decreases with distincitive ones, and to maintain otherwise seems downright implausible. We can try to get around point 2, however, by going to a different kind of categorization model. To illustrate, we might have a model that computes the similarity between the probe and target concepts, and responds affirmatively ("the probe is a member of the target") as soon as the similarity score exceeds some threshold. This model predicts faster times for immediate than distant superordinates, which is consistent with the majority results, but it no longer handles the exceptions. In sum, the results with nested triples are difficult to reconcile with the classical view.

## SUMMARY

Table 7 summarizes the four sets of experimental findings just discussed. Again we need to ask, how badly do they damage the classical view? We argued at length that the first set of findings, simple typicality effects, do not really tarnish the classical view because they can be readily explained by the complexity model that is based on this view. The other three sets of results, however, pose serious problems for models based on the classical view.

The second set included Rosch and Mervis's family resemblance results (1975), which showed that typicality variations are cor-

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TABLE 7 FOUR EXPERIMENTAL CRITICISMS OF THE CLASSICAL VIEW

1. Simple typicality effects – ratings, categorization times and errors, ease of learning, order of production, and cognitive reference points

2. Determinants of typicality – typicality and the distribution of features across concept members

3. Use of nonnecessary features

4. Nested concepts-similarity and categorization times

related with variations in the distribution of features across concept members. To accommodate these results to the classical view required many ad hoc assumptions; some were needed to explain the relation between listed and true features, while others arose in the effort to specify a classical-view model that could handle the critical results. All told, these results place a heavy burden on the classical view.

The third set of findings consisted of experimental demonstrations that people use nonnecessary features in making semantic categorizations. Taken at face value, these demonstrations constitute strong evidence against the classical view because any model based on the view would presumably be restricted to necessary and sufficient features. One could, however, challenge whether these experiments tapped the real features of concepts, but such challenges seem to invoke more tenuous assumptions.

Finally, we considered findings on nested concepts. The classical view clearly predicts that a subset should be judged more similar to its immediate than it distant superordinate. Although this is generally true, there are some counterexamples. Unless all counterexamples can be explained away by artifacts, they constitute solid evidence against the classical view. We also considered categorization results showing that times are generally faster for immediate than for distant superordinates, but there were some definite exceptions. The general result was inconsistent with most straightforward models based on the classical view. Though we could come up with another classical model that would handle the general result, the model would then be inconsistent with the exceptions.

Each of the last three sets of results offers some evidence against the classical view. In no single case is the evidence unimpeachable, but taken together the three sets of results start to mount a strong case. Furthermore, the three sets of results fit together like a glove. Certain nonnecessary features appear to be used in categorization, and the distribution of these features apparently leads to typicality effects as well as to cases where a subset is judged more similar to a distant than an immediate superordinate. In contrast, salvaging the classical view from the three critical sets of results involves a great deal of patchwork, where the patching needed in one spot is of no help in another.

When the three critical sets of results are combined with the general criticisms of the classical view discussed earlier, the case against the classical view of concepts starts to look very imposing.

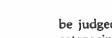
## Radical Attempts to Salvage the Classical View

So far we have attempted to defend the classical view by either challenging the basis of the evidence used against it, or by making minimal additions to the view (like the addition of the processing assumptions of the complexity model). As noted, however, these challenges and additions are too piecemeal, and they do not add up to a parsimonious proposal. It is time to consider some more sweeping and radical attempts to salvage the classical view.

In this section we consider three such attempts. The first is based on the assumption that categorization depends on interconcept links, not on the actual features of concepts. The next approach starts by dropping the third assumption of the classical view (the one dealing with the nesting of defining features) and explores the consequences of this move for the findings we have considered. Finally, the third approach takes off from our notion that every concept contains both a core and an identification procedure; while the core may conform to the classical view, the identification procedure need not, and the latter is assumed to be the source of the findings we have considered.

## Access Links between Concepts

Suppose that in addition to acquiring classical-view concepts, people also learn direct links between them that can be used to access one concept from another. This assumption has important implications for performance in a semantic categorization task, the task that has produced most of the experimental evidence against the classical view. Specifically, when asked whether one concept is a subset of another, people check the interconcept links and not the features of concepts. That is, performance in a semantic categorization task does not reflect the contents of concepts at all but rather the ease with which one can move from one concept to another. This idea is illustrated in Figure 7, where the interconcept links are represented by labeled paths. This kind of representation is borrowed from the current network models of knowledge representations (for example, Anderson and Bower, 1973; Collins and Loftus, 1975; Norman and Rumelhart, 1975; Anderson, 1976), but the



be judged more similar to animal than to bird, and will also be categorized faster as animal than as bird. Conversely, whenever a subset is judged more similar to its immediate than its distant superordinate, there are no shortcuts (or if there are, they are relatively inaccessible); hence the subset will be categorized faster at the immediate superordinate level.

The Classical View

Despite its apparent successes, this access-path approach has serious drawbacks. First, although it accounts for some of the experimental findings that embarrassed the classical view, by no means does it account for all. In particular, it offers no explanation of why (1) typicality variations in natural concepts are correlated with variations in the distributions of listed features; (2) typicality variations in artificial concepts can be induced by variations in the distribution of features; and (3) various experiments have revealed evidence for the use of nonnecessary features in categorization. Second, the access-path approach does not even address some of the general empirical arguments raised against the classical view. There is nothing in the approach, for example, that comes to grips with the possible existence of disjunctive concepts, or the failure to specify the defining features of concepts. (Remember, the accesspath approach assumes that there are defining features for concepts, even though they are not used in categorization tasks.) Third, in cases where the approach succeeds – as in explaining typicality effects – the access-path approach seems too unconstrained. Though there are some suggested empirical measures of the accessibility of one concept from another (roughly, the frequency with which one concept name leads to the other; see Glass and Holyoak, 1975), the approach still lacks criteria for specifying when paths are formed, what affects their lengths, and so on.

These drawbacks are serious, if not overwhelming. The move away from feature-based processes in categorization, and toward path-search processes, seems to have solved few problems. The next two attempts to salvage the classical view maintain a featurebased approach and seem a bit more promising.<sup>4</sup>

#### **TRANSLATIONS BETWEEN FEATURES**

Another radical modification of the classical view starts by dropping its third assumption – that the defining features of a concept are nested in those of its subsets. This assumption seems to be the cause of many shortcomings of the classical view: it is solely responsible for the prediction that a specific subset must always be judged more similar to its immediate than its distant superordinate, and it plays a major role in the prediction that there should be no unclear cases, since the nesting of a concept's defining features in

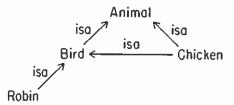


Figure 7 Part of a network of interconcept links

present proposal differs from most netword models in its claim that the concepts being linked together conform to the classical view. Indeed, the only existing network model that is close to the spirit of Figure 7 is that proposed by Glass and Holyoak (1975), for they also assumed that concepts have necessary and sufficient conditions. Some of the assumptions we make in the following analysis are taken from Glass and Holyoak (1975), while others are borrowed from Collins and Loftus (1975).

There are two critical points about the links or paths in Figure 7. First, the paths between adjacent levels vary in length, with longer paths reflecting longer access times. This assumption is capable of accounting for most simple typicality effects. When asked to judge the typicality of a probe concept to a target one, subjects base their judgments on the length of the path between the two concepts; when categorizing the probe as a member of the target concept, subjects traverse the probe-target path, where these paths are shorter for typical probes (like the robin-bird path) than for atypical probes (like the chicken-bird path); when subset relations between concepts are being learned, relations characterized by shorter paths are acquired earlier.

The second important point about the paths in Figure 7 is that shortcuts are possible; that is, there are direct paths between concepts more than one level apart, and some of these may be shorter than paths between adjacent levels. For example, there is a shortcut between chicken and animal that is shorter and hence more accessible than the path between chicken and bird. This notion helps explain the data on nested concepts. Whenever a subset is judged more similar to its distant than its immediate superordinate, this can be attributed to the existence of a shortcut between the subset and the distant superordinate that is shorter than the path between the subset and the immediate superordinate. And since the shorter path is more accessible, the subset will be categorized faster vis-àvis its distant than its immediate superordinate. Thus chicken will those of its subset is the basis of the classical view's supposed algorithm for determining subset membership. The nesting assumption may even be partly responsible for the failure of semanticists to find defining features for many concepts, because the search may have been overly constrained to features that meet the nesting criterion. Thus, dropping the nesting assumption seems a reasonable starting point for a modification of the classical view.

Under this modification, classical-view concepts are still assumed to be summary representations containing necessary and sufficient features, but now at least some features of a particular concept do not have to be identical to those in the concept's subsets. But in such cases, how can one use features to determine if one concept is a subset of another? The simplest answer is that there are rules or relations for directly translating one feature into another. This idea is illustrated in Figure 8. One defining feature of bird is "animate," and it is listed for robin but not for chicken. However, chicken includes the defining feature "egg-laying," where the latter implies animate. Now one can establish that a target concept includes a probe either by matching the target's features to those of the probe (as in the bird-robin case in Figure 8) or by using the interfeature relations to translate one feature into another (the chicken-bird case).

These ideas can be used to account for most of the experimental findings discussed earlier. To explain simple typicality effects, we assume that the less typical a subset, the more features it contains that require translation when it is compared to its parent concept. In Figure 8, for example, the atypical chicken requires translation of its egg-laying feature, while the more typical robin needs no translation at all. Assuming a translation operation requires more time and is more error-prone than a simple matching operation, we would expect atypical instances to be categorized slower and less accurately than typical ones. To account for Rosch and Mervis's finding (1975) that typical subsets contain features that are fre-

| <u>Bird</u> | Chicken          | Robin     |
|-------------|------------------|-----------|
| living      | living           | living    |
| animate <   | – egg-laying ——> | > animate |
| feathered   | feathered        | feathered |
|             | <del></del>      |           |

**Figure 8** Direct translation between features

quently listed for other subsets of the concept, we might assume that features listed for many subsets are more likely to be part of the concept's representation and hence require no translation. Again, typicality comes down to an inverse measure of the amount of translation needed in relating two concepts.

The data on nested concepts also fit nicely with this translation approach. When the features of a subset do not require any translation vis-à-vis the parent concept, we have a perfect nesting of the concept's features in those of the subset, and we expect the similarity predictions of the classical view to hold; when some features do require translation, we have less than perfect nesting, and exceptions to the similarity prediction are expected. Moreover, since atypical subsets are more likely to require translation, such subsets should constitute the bulk of the exceptions, which seems to be the case.

There is an interesting aspect to the above arguments. In all cases, the typical members of a concept are treated in roughly the way the classical view specifies (few or no translations are needed), whereas atypical members are treated differently (translations are frequently required). The direct-translation approach thus has the character of a rule-plus-exception approach, typical members being handled by the rule (the classical view) and atypical members being handled as exceptions. This approach has the desirable property that much of the machinery of the classical view is salvaged, the view now being limited to the more typical members.

The preceding discussion highlights some strong points of the translation approach. There is, however, one sore point for this approach when it comes to accounting for experimental findings. In explaining Rosch and Mervis's results (1975), we explicitly assumed that features listed of many subsets were often defining of the concept, and implicitly assumed that features listed for any subset were defining of it. Inspection of feature listings like those collected by Rosch and Mervis, however, provides little support for either assumption. Some frequently listed features are clearly non-necessary for parent concept or subset. This relates to a more general problem: the translation approach we have sketched has no natural way of dealing with the use of nonnecessary features in categorization, which is one of the major experimental findings.

Though the direct-translation approach can also be stretched to deal with some of the general empirical arguments raised against the classical view, most of what can be said is quite vague. We have already noted that it may be easier to specify defining features of concepts if we drop the nesting constraint. But this is merely a promissory note, particularly since the direct-translation approach

still assumes a substantial degree of feature nesting between concepts and their typical members. The occurrence of unclear cases may be attributed to the need to translate between features. Perhaps there are subset relations, like "tomato-fruit," in which the needed translation relations are imperfect or unknown. Without some further specification of the actual interfeature relations, though, it is hard to evaluate this possibility. And finally, there is the question of what the direct translation approach has to say about disjunctive concepts. One possible answer might go as follows: Concepts are truly conjunctive, but they may appear disjunctive when we focus on their members and notice that the latter do not share many features; to illustrate with the concept of bird, we might focus on robin and chicken and note that one contains "animate," the other "egg-laying," and mistakenly conclude that bird is disjunctive. Although this answer is possible, we are hard pressed to put much credence in it.

Some of the problems just mentioned may be alleviated by a more extreme form of the translation approach.<sup>5</sup> So far we have assumed a direct translation between features, but translation can also operate indirectly in that two defining features (of two different concepts) might be connected by a third feature. To illustrate, a defining feature of fruit might be "seeds," while such a feature of orange might be "acidic." To translate between them, one could use the information that (1) seeded objects are often juicy and (2) acidic objects are often juicy. Hence both seeds and acidic lead to juicy, so juicy serves to translate between the two defining features. Note that this translation process is probabilistic, since the intermediate feature, juicy, is not true of all seeded objects. This kind of indirect translation is consistent with the use of nonnecessary features, for intermediate features may be nonnecessary yet used in the categorization process.

An indirect-translation process can also be elaborated to account for other experimental findings. Two examples should suffice. First, variations in the typicality of concept members might reflect either the necessity of translation or the ease with which the defining features of the members can be translated into the defining features of the concept. Second, the Rosch and Mervis correlations between typicality and the distribution of listed features (1975) might be explained by assuming that the listed features are those used in the translation process; for example, features used to translate many concept members may be more powerful or accessible than features used to translate few members.

The obvious problem with the indirect-translation approach is that it is extremely ad hoc. No constraints of any kind have been

placed on the nature of intermediate features or on their relations to defining ones, and it is unclear what, if any, predictions follow from this approach. Nevertheless, indirect translation at least addresses a wide range of problems and may, if properly developed, have the potential to save the classical view. Such a rescue will come at a high price, however: for all explanations of empirical results will be in terms of intermediate features rather than defining ones. Indeed, the defining features, which are the heart of the classical view, seem to be doing no theoretical work at all, and it becomes unclear why one need posit them to explain categorization. To some extent, the same is true of the direct-translation approach, where most of the explanations hinged on relations between features rather than on the features per se. And as we will see, the identical problem arises in our third way of salvaging the classical view.

GREATER ACCESSIBILITY OF IDENTIFICATION PROCEDURES

Until now we have assumed the following about concept cores and identification procedures:

- 1. Though many concepts contain an identification procedure as well as a core, the core is more important psychologically because it must always be there and because it determines the contents of the identification procedure.
- 2. We must thus focus on empirical methods that are likely to involve the core rather than the identification procedure.
- 3. Therefore, we should concentrate on semantic categorization tasks rather than perceptual ones, since the semantic task requires a consideration of only the cores.

By challenging the third assumption, we generate our third attempt to salvage the classical view. Specifically, we now assume that the identification procedure is more accessible than the core: consequently, semantic categorization is often based on a comparison of the identification features of the target and probe concepts rather than on a comparison of the core's defining features. This new proposal also partly undermines assumption 1 above: if identification procedures are so widely used, it may be misleading to call the core "more important psychologically."

Given that semantic categorizations may be based on a comparison of identification procedures, most of the relevant experimental results fall into place. First, an immediate consequence is that many features used in categorization will be nonnecessary ones, since the features in identification procedures will often be

nonnecessary. Second, simple typicality effects can now be explained in terms of similarity of identification features. The more similar are the identification features of a concept to the identification features of one of its members, the more typical that member is judged to be. Now a categorization model that computes the featural similarity between target and probe concepts will yield faster and more accurate decisions for typical members. Third, the Rosch and Mervis finding (1975)-that typical members contain features common to many other members - can be interpreted solely in terms of identification features. Identification features common to many members are likely to be included in the identification procedure of the concept itself, so again typicality comes down to a matter of similarity of identification features between a concept and its members. Finally, since a concept's identification features need not be perfectly nested in those of its subsets, the similarity prediction of the classical view is no longer expected to hold in all cases.

In addition to accounting for the above results, the assumption that semantic categorizations are based on identification features has another important experimental consequence: semantic categorizations should resemble perceptual ones, since both rely on the same identification features. Hence our earlier stricture against using results from perceptual categorization to evaluate views of concepts must be temporarily suspended, and we need to look briefly at some comparisons of semantic and perceptual categorization.

Though there are few research reports that afford a detailed comparison between semantic and perceptual categorization, what is available shows marked similarities between the two. If one asks subjects to rate a concept's members for typicality, the typicality ordering will be virtually identical for members presented as pictured instances and for members expressed as words (see Smith, Balzano, and Walker, 1978). Moreover, aside from the fact that pictures are responded to slightly faster than words, the effects of typicality are the same in perceptual and semantic categorization (see Guenther and Klatzky, 1977). Also, to the extent that comparable data are available, items that lead to faster categorizations with immediate than with distant superordinates do so regardless of whether the item is presented as a picture (that is, as a specific instance) or as a word (that is, as a subset; compare Smith, Shoben, and Rips, 1974, with Smith, Balzano, and Walker, 1978). And finally, the features listed for items presented as words substantially overlap the features listed for these same items presented as pictures (compare, for example, Hampton, 1979, with Rosch et al., 1976). Although some of these comparisons are tenuous because they involve contrasting results from different studies, they at least suggest that many of the same features are used in semantic and perceptual categorization. This bolsters our new assumption that identification procedures underlie semantic categorization.

The idea of accessible identification procedures can also offset one of the general empirical arguments raised against the classical view, namely, that some concepts are disjunctive. As mentioned earlier in this chapter, some concepts may appear disjunctive when we mistakenly focus on their identification features rather than on their core features. However, a reliance on identification features is of less help in warding off other general empirical arguments. Positing accessible identification features does not explain the prolonged failure to specify the defining features of the core; nor does it get rid of the problem of unclear cases: for example, if asked, "Is a tomato a fruit?" and given sufficient time to mull it over, one should be able to use the core features to resolve the matter.

Despite the two deficiencies just noted, the present approach gets rid of many of the classical view's empirical problems and correctly predicts a similarity between semantic and perceptual categorization. And it does not seem as unconstrained as the other attempts to salvage the classical view, since identification features would at least be restricted to properties that people actually use in deciding that a physical object is an instance of a concept. However, as was the case with the translation approach considered earlier, the present approach seems to save the classical view by shifting all the theoretical action away from the defining features of the core. Again we may raise the question, why bother to posit classicalview concept cores at all?

One possible answer that seems reasonable to us is that while the identification procedure may form the front line of categorization, the core is used as a backup procedure (this is similar to an argument made by Katz, 1977). That is, difficult categorizations, ones that cannot be done by the identification procedure, are eventually tackled by the classically defined core. This claim is seriously challenged, however, by the presence of unclear cases. Another possible answer is that while an identification procedure generally takes care of categorization, the core plays a major role when we do things with concepts other than categorizing. We may, for example, work mainly with classically defined cores when we combine simple concepts into complex ones (the conceptualcombination function of concepts), or when we draw inferences from existent propositional representations. We will not take up the pros and cons of this proposal since it would take us too far from the mainstream of this book. All we can conclude is that,

## Categories and Concepts

when restricted to categorization phenomena, the proposal of accessible-identification procedures essentially salvages the classical view by ignoring it.

## SUMMARY

All three attempts to salvage the classical view have their problems. The first, or access-path approach, seems the most problematic. It fails to deal with most criticisms of the classical view – namely, the correlation between feature distributions and typicality, the use of nonnecessary features, the apparent presence of disjunctive concepts, and the inability to specify defining features. Moreover, the access-path approach runs up this rather impressive list of failings while imposing overly powerful and unconstrained assumptions.

The translation approach fares better. Direct translation seems to handle readily several experimental findings in a parsimonious fashion. However, it lacks convincing accounts of the general empirical problems of the classical view (for example, failure to specify defining features), and it cannot explain the use of nonnecessary features in categorization. The latter problem could be solved by positing indirect translation, with its notion of intermediate features, but at the cost of introducing many unconstrained assumptions.

The final approach, accessible identification procedures that determine semantic as well as perceptual categorizations, appears to have the most potential. However – and this is the critical point – all of its potential in handling categorization phenomena seems to be dependent on the nonnecessary (identification) features, and such nonnecessary features are the backbone of the probabilistic view. In short, the most promising attempt to salvage the classical view is promising because it moves toward the probabilistic view.