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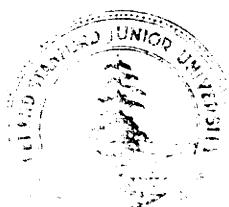
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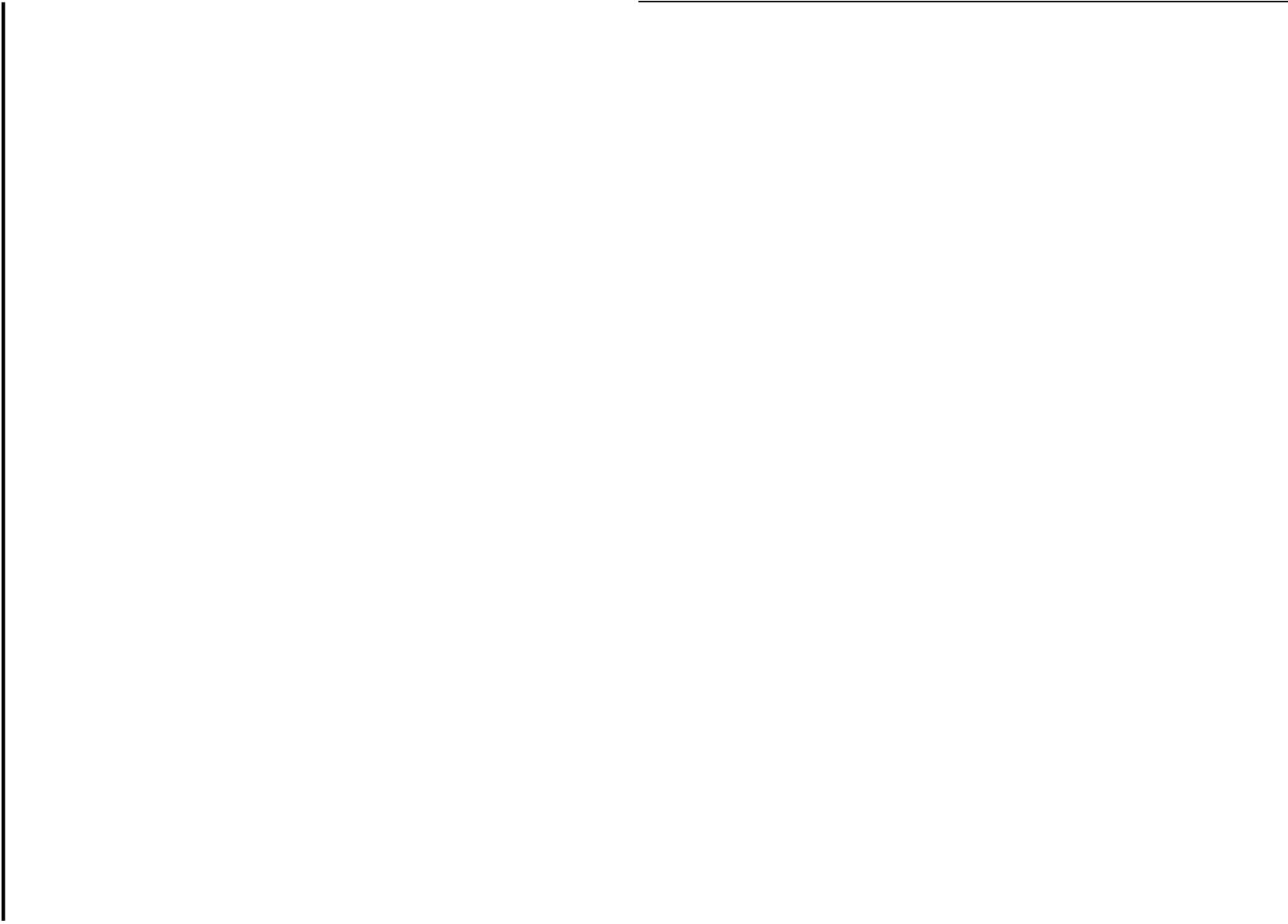
ON GENERALITY AND PROBLEM SOLVING  
A CASE STUDY USING THE DENDRAL PROGRAM

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**ABSTRACT:** Heuristic DENDRAL is a computer program written to solve problems of inductive inference in organic chemistry. This paper will use the design of Heuristic DENDRAL and its performance on different problems for a discussion of the following topics:

1. the design for generality;
2. the performance problems attendant upon too much generality;
3. the coupling of expertise to the general problem solving processes;
4. the symbiotic relationship between generality and expertness, and the implications of this symbiosis for the study and design of problem solving systems.

We conclude the paper with a view of the design for a general problem solver that is a variant of the "big switch" theory of generality.

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On Generality and Problem Solving:  
A Case Study Using the DENDRAL Program

Edward A. Feigenbaum, Bruce G. Buchanan and Joshua Lederberg

In discussing the capability of a problem solving system, one should distinguish between generality and expertness. Generality is being questioned when we ask: how broad a universe of problems is the problem solver prepared to work on? Expertness is being questioned when we ask: how good are the answers and were they arrived at with reasonable cost? Generality has great utility in some ways, but is not often associated with superior performance. The experts usually are specialists.

In analytic chemistry, there is a domain of inductive inference problems involving the determination of molecular structure by analysis of certain physical spectra of the molecule. We have written a problem solving program (Heuristic DENDRAL) that is prepared to attempt to solve any problem in this very large domain. By now, it has solved hundreds of structure determination problems and in many different chemical families. For some families of molecules, it is an expert, even when compared with the best human performance. For the other families, i.e., most of chemistry,

it performs as a novice, or worse.

This paper will use the design of **Heuristic DENDRAL** and its performance on many different problems it has solved as **raw material** for a discussion of the following topics:

1. the **design** for generality;
2. the performance **problems** attendant upon too much generality;
3. the **coupling** of expertise **to the general problem solving processes**;
4. the symbiotic relationship between **generality and expertness**, and the implications of this symbiosis for the study and design of problem **solving systems**.

We conclude the paper with a view of the **design** for a general problem solver that is a variant of the "**bigswitch**" theory of generality.

Previous papers have given a detailed exposition of **the workings** of the Heuristic DENDRAL program (Buchanan, et al, 1969) and a **discussion** of some general issues of representation and theory formation suggested by **the DENDRAL work** (Buchanan, et al, 1970). It is fair to ask for an integrated presentation of the **results of this application of heuristic programming** to an important chemical inference

problem. Several papers presenting these results to chemists have appeared or are in press (Lederberg, et al. 1969; Duffield, et al. 1969; Schroll, et al. 1963; Buchs, et al. 1970), but no summary of these results is available in the artificial intelligence literature.

Yet the attention given to the program as an application of **artificial** intelligence research has tended to **obscure** the more general concerns of the project investigators. These are:

1. To **study** and construct **detailed** information processing models of processes of scientific inference. By scientific inference we mean the inferential process by which a **model** is constructed to explain a given set of empirical data.
2. To study **experimentally** the "**operating** characteristics@\*" and the **effectiveness** of different designs (strategies) for the deployment of task-specific knowledge in a scientific area.
3. To develop a method for eliciting from an expert the heuristics of scientific judgment and choice that he is using in the performance of a complex inference task,
4. To solve real problems in an area of significance to modern science, and to do so with a level of performance high enough to have a noticeable impact upon that area of science.
5. To **discover** the heuristics which lie behind efficient selection. As we conclude later, the significant problem may

not be so much tuning a specialist with a new set of heuristics as learning how to acquire these heuristics,

#### THE TASK ENVIRONMENT

For the **sake** of **completeness** and **review**, we include here a brief description of the scientific problem that was chosen **as** the task environment in which to pursue the project's goals (publications **listed** in the References **will** give the interested reader the complete story). The problem given to the program is the usual problem of **the** analytic chemist: to **determine** the molecular structure of an unknown **compound**. While the chemist may **use** many analytic techniques, the program uses only two of the most important tools to collect **data** about the unknown sample. The **primary** source of empirical data is a mass spectrometer, an instrument that fragments **molecules** of a chemical sample (using an electron **beam**) and records the results. A mass spectrum, **the** output of **the mass** spectrometer, is a two-dimensional record of the abundance of various fragments plotted as a function of their molecular weights. A secondary source of data is a nuclear magnetic resonance (**NMR**) spectrometer, which uses variations in magnetic field strengths to provide information about certain specific kinds of structure internal to a molecule. (In addition, there is no difficulty in utilizing a third source of **data**, **the** infrared (**IR**) spectrometer, as soon as it

becomes sufficiently important to do so.)

The problem solver is given the mass spectrum, the **NMR** spectrum if it is available, and the elementary formula if it is available (number of atoms of each kind). For the classes of molecules reported in this paper, the program need not be given the formula **but** can infer it directly from the spectrum by a **heuristic** procedure,

The output of the problem solver is a graph, i.e., a topological model, of the molecular structure of the unknown compound. Or, if more than one graph is a plausible explanation of the given data, **the** output is a list of **the** plausible molecular graphs, rank ordered, with their relative plausibility scores.

The determination of molecular structure by these electronic instrumental techniques is seen by physical chemists to be a significant advance over older chemical methods, and is enticing because of the speed **and** economy of the analysis and the generality of **the** approach. However, the almost bewildering variety of fragmentations and reactions that can be induced **by** the high energy of the electron beam in a mass spectrometer are far from being completely understood, so that the science of mass spectrum analysis, though no longer an infant, has still not reached its maturity.

## GENERALITY VS SPEED AND ECONOMY

"A view of existing problem solving programs would suggest, as common sense would also, that there is a kind of "law of nature" operating that relates problem solving generality (breadth of applicability) **inversely** to power (solution successes, efficiency, etc.) and power directly to specificity (task-specific information). (Feigenbaum, 1968)

"Evidently there is an inverse relationship **between** the generality of a method and its power. Each **added** condition in the problem statement is one more item that can be exploited in finding the solution, hence in increasing the **power.**" (Newell, 1969)

One does not need a view of generality in problem solving systems of the scope of CPS (Ernst **and** Newell, 1969) to appreciate the importance of this tradeoff between generality (breadth of applicability) and effectiveness in solving a given problem (particularly speed and cost). The story of the **DENDRAL** program's success as an application is in part a story of this tradeoff, which the remainder of this paper will sketch. We approach this discussion of generality of problem solving systems with some caution since the history of the

search for generality in **problem solvers** (primarily the GPS effort) will tend to color the discussion no matter what **we** say or do not say about it.

Structure determination by mass spectral analysis is a technique pursued by its scientific practitioners because of its generality: its broad applicability **to** all types of molecules. The **designer** of a problem **solving** system to interface with this empirical data is inclined, at least initially, to try **to match the generality** of the physical process with generality of the reasoning process. Yet he soon finds, paradoxically, that he can not **afford** this **match**, that he must retreat and rework his analysis into more **specialized** forms if he is to be able to use **his** problem solver on real problems.

The Heuristic **DENDRAL** program has solved **hundreds** of structural inference problems, most recently of structures **in** the family of organic **amines**, **for which** the analysis is reasonably complex. The difference in running **speed** between **solving these** problems by the most general methods known **to** the program **and** solving them by its heuristic methods specialized for this type of problem is estimated to **be as** large as a factor of thirty thousand!

The world known to **the DENDRAL** program is the world of organic

chemical structures. For the purposes of this paper DENDRAL's world will be taken to be the world of non-ringed (acyclic) organic molecules, although not all parts of the program are so constrained.\*

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\*As of July, 1970, the Structure Generator could delineate all acyclic isomers and all mono-cyclic (single-ringed) isomers of a given chemical formula, the Predictor could predict mass spectra for acyclic molecules (and manipulate the internal structure of any cyclic molecules), and the Planner could infer structural information from the spectral data of any saturated acyclic monofunctional molecule.

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In the discussion to follow generality will mean breadth of applicability within the confines of the DENDRAL world. Some procedures apply to all possible structures in this world, and they will be considered the most general. If there were a procedure that applied to only a single molecule, that procedure would be the least general. Thus, generality is to be taken to mean relative generality in the DENDRAL world.

#### THE GENERAL PROBLEM SOLVERS OF THE DENDRAL WORLD

In another place, we have summarized our overall design

philosophy as follows: **"Some** of the essential features of the **DENDRAL** program include:

- 1)** Conceptualizing organic chemistry in terms of topological graph theory, i.e., a general theory of ways of combining atoms.
- 2)** Embodying this approach in an exhaustive hypothesis generator. This is a program which is capable, in principle, **of** "imagining" every conceivable **molecular** structure.
- 3)** organizing the generator so that it avoids duplication and irrelevancy, and moves from structure **to structure** in an orderly and predictable way. The key concept is that induction becomes a process of efficient selection from the domain of all possible structures. Heuristic search and evaluation is used **to** implement. this efficient selection2

This is a design philosophy which is clearly aimed at the most general kind of problem solving capability within the DENDRAL world, that is any mass spectrum and associated chemical formula within the DENDRAL world-can be treated.

From another point of view, the DENDRAL program can be seen to be implemented within a generate-and-test paradigm, to use **Newell's** terminology (Newell, 1969). The **"generate"** part is the Structure Generator program and the **"test"** part **is** the Predictor program. Hypothesis generation and hypothesis-validation are equally appropriate labels for these two stages

of the problem solving.

The Structure Generator incorporates:

1. an algorithm that allows it to **proceed** systematically from one possible candidate to the next, i.e., a legal **move** generator that defines the space;
2. general criteria for instability of organic molecules **that** allow it to **avoid** working on **chemically irrelevant** structures;
3. procedures for **treating** subgrphs as if they were atoms, allowing particularly important combinations of atoms to be treated as a unit in **the** combinatorial work of the generator, **Because of the** structure of molecular graphs, this task environment lends itself to partial solutions using the techniques described below.

The Structure Generator program knows nothing of the theory of mass spectrcmetry. Given a chemical formula, it will generate all the isomers (structural variants,) that, are chemically plausible a priori. These **are the** candidates that are input to the "**test**" part of the generate-and-test procedure.

The Structure Generator, even when used alone, has performed valuable service for chemists by exhibiting the sizes **and** structures of the **analytic chemist's** problem spaces. The number of chemically possible structural models, as shown in

Table 1, is an **important** boundary on a **chemist's problem** hitherto known only for a few classes of problems (see Lederberg, et al. 1969).

The Predictor program is the "**expert**" on the general theory of mass spectrometry. It **answers** this question for the system: Though the candidate may **be** chemically plausible on a priori **grounds**, is it a good candidate to **explain** the given mass spectrum? In other words, does its-predicted spectrum fit the data?

The **Predictor** incorporates a general theory of the fragmentation and recombination processes that can take place in a mass spectrometer, insofar as these are known to our chemist collaborators. The Predictor program is continually under development as the theory of mass spectrometry develops.

Any chemical structure in the DENDRAL world can be handled by the Predictor. In this sense, the **Predictor** is as general a problem solving element as the Structure Generator; in fact, it is the necessary complement.

The Heuristic DENDRAL program contains a great deal more **than** just this generate-and-test **team**, as will be described **subsequently**. But it is instructive to ask: how powerful are these "generalists" in solving mass spectral analysis

problems?

Table 2 exhibits the results for selected members of the family of amino acids. This family is distinguished from the other families **with** which we have worked by virtue of containing a relatively large **number of heteroatoms** (atoms not **carbon** or hydrogen) relative to the number of carbon atoms. For **each** entry, we give its common name, **its** chemical formula, **the** size of **the** problem space **in** terms of the number of topologically possible isomers, the number **of** chemically plausible isomers **actually generated by** the Structure Generator (using the "**zero-order**" theory explained below), and the rank **order** assigned to the correct candidate (i.e., the "**right answer**") by **the** Predictor. It will be seen that the heuristics concerning unstable **molecules** have a substantial effect for amino **acids**, i.e., **the** number **of** chemically plausible molecules is **much less** than the number of topologically possible **candidates**. This will not in general be true for molecules with **fewer** types of **atoms** for example, ketones, ethers and **amines**, as **we** shall later set.

#### **PROBLEMS ATTENDANT UPON TOO MUCH GENERALITY**

Experiments such as **those just** summarized pointed up design problems that were consequences of **the program's** generality.

As a result of **having** to be prepared to **handle** in a homogeneous and **complete** manner any formula or any structure presented, **the** programs are costly in terms of computer running time and use of main memory. **With** respect to the Predictor, this means that it is feasible to test only a relatively small number of candidate solutions. **In** the **Structure Generator** this means that it is **feasible** to **start** with only a **small** collection of atoms.

The generality of the Structure Generator, which employs only relatively weak **a priori** constraints and no constraints imposed by the data, tends toward producing too many "plausible" candidates. The **generate-and-test** procedure breaks down **because** the generator is too prolific and the **test** is too expensive.

The solution to this design problem is to strengthen the heuristic **controls** over the generation of candidate solutions. There are a number of ways available to do this, some of which **were** tried with success, some with failure. The failures were at least as illuminating as **the successes**.

The **most** obvious way will be **mentioned** first, and then discussed no further in this paper. It is this: **review** carefully **the** tricks in the- heuristic programmer's **toolkit** (particularly those that apply to **the** search of **AND-OR** problem

reduction trees) and do not fail to apply them when they are applicable. The following examples from the Structure Generator illustrate the point:

1. At an OR node (in DENDRAL, the selection of a particular partitioning of the remaining unassigned atoms), try the easiest subproblem **first**. At an AND node (in DENDRAL, making radicals from partition elements), **try the hardest subproblem first**.
2. Limit the **number of subproblems considered** at an OR node by evaluating the "**quality**" of subproblems and discarding those below a **threshold value**.
3. For difficult problems, allow human intervention in the choice of **subproblems** (this potentially powerful heuristic procedure is available in DENDRAL, but has never been used in solving problems).

#### HEURISTICS RELATED TO PROBLEM DATA: THE **EMERGENCE** OF "SPECIALISTS"

By far the most powerful method of gaining effective control over the generator is to force its search to be relevant to specific problem data **given** as the input (the spectral data). That is, the candidates produced by the generator must be not only chemically plausible a priori but also likely solutions

to the specific problem at **hand**.

In DENDRAL, one method for doing this is as follows: whenever a move in the problem space defines a new piece of an emerging structure, validate the move with respect to **mass** spectral theory by predicting its consequences in terms of **expected** spectral lines; an? prune **moves** that can not **be** so validated. In other words, **reduce** the search in light of the problem data by applying the theory of mass spectrometry to **nodes** in the problem space. For example, prune all structures to **be** built out of a cluster of 2 carbon atoms, 3 hydrogens and 2 oxygens if there is no corresponding data point (mass = **59**). A simple version of **this** method was used in early versions of **the DENDRAL** program. The theory of mass spectrometry used was so oversimplified that we called it derisively "**the** zero-order theory of mass **spectrometry**". Yet it **turned** out to be a cheap an3 effective pruning criterion for some problems, namely the amino acids, for whose fragmentation the zero-order theory **was** not a bad theory.

The zero-order theory failed, of course, on more complex problems, but a better theory was available, the general theory in the Predictor. A **procedure** was developed by which the Predictor was called every time there was a need for validation of a partial structure.

When in doubt consult the "generalist"! But the ~~design~~ experiment failed, for these reasons:

1. The "generalist", as we have said, is too expensive ~~even~~ for partial structures; and it was called too frequently.
2. The theory is most powerful in making statements about fragmentation at termini of chemical graphs; ~~but the~~ Structure Generator builds ~~candidate~~ graphs ~~by~~ starting at the center of ~~the~~ graph and building toward ~~the~~ termini. Thus the theory ~~was most powerful~~ precisely when it was ~~having the~~ least heuristic effect! This representational mismatch could ~~have~~ been remedied by considerable reprogramming (although a total correction would have ~~benefitted~~ by a complete reconceptualization and reprogramming of the Structure Generator), ~~but~~ it points up how critical are ~~the~~ problems of representation when one considers using ~~the~~ knowledge held by one process to control another.

There are other heuristic ~~methods~~ available in this concrete, running program, however. These we ~~shall~~ call "~~aggregation~~" and "~~planning~~". Both have general (an? well recognize?!) importance ~~quite~~ apart from their power in the ~~DENDRAL~~ application. In DENDRAL, both are employed prior to the search for candidate solutions, and serve to "~~preset~~" the generator to ~~work~~ on only ~~those~~ families of structures ~~that~~ meet certain conditions inferred from the problem data. To be effective, these processes ~~must be~~ cheap, relative to a search

unconstrained by their inferences. **As** we shall see, this is achieved by the use of highly specialized rules for interpreting the "meaning" of the problem **data** (spectral lines). These rules are the formal representation of what the chemist considers to be his **good** judgment in properly organizing his inference problem.

**Aggregation** is a **self-evident general** technique for reducing the number of alternatives produced by any combinatorial generator. **Aggregate** the combinatorial **elements** into **bigger** units and treat these as if they were elements. In **DENDRAL**, any **subgraph** can be treated as a "**superatom**" **with** a valence. The internal structure of the superatom is not manipulated by the combinatorial generator.

The most **general** view of the **aggregation** heuristic in DENDRAL is this:

**Use** whatever specialized knowledge and processes and whatever auxiliary data are available to infer pieces (partial structures) of the solution. **Make** these superatoms, **For** the remaining atoms, uncommitted to superatoms, use the **general** structure generating machinery to build the interstitial structures in all the ways allowed by the heuristics defining chemical **plausibility**.

This **general** approach has been used in many particular ways.

For example:

1. The Structure Generator can be supplied with a list of **superatoms** that are known a priori to **be** highly stable and therefore likely to occur in nature,
2. A nuclear **magnetic** resonance spectrum, **important** auxiliary data to a **mass** spectrum analysis, often **provides** clear and easily obtained information **about** the number of methyl superatoms (**CH<sub>3</sub>**) in the structure. Infra-red and ultra-violet spectra can reveal other kinds of **substructure**, which can be similarly treated as **superatoms**.
3. **The key** subgraphs of a molecule (those containing the heteroatoms) **usually leave** their **particular "fingerprints"** in the lines of the **mass** spectrum. Complex pattern recognition **criteria** have been **developed** by us for identifying these **key subgraphs**, which **are** then treated **as** superatoms. A few of these rules are shown in Table 3.
4. Sequence extrapolation **and** **left** numerology have been used to infer some simple structures, such **as the** longest **unbranched** chain in **the molecule**. Once identified, they **become** superatoms.
5. By direct human intervention, any **aggregation--any** **superatom--** can be **established**. This is of great importance when **the** program is used as an **"assistant"** in a very **complicated** problem. The human chemist often knows in advance **basically what** kind of structures he is working with, i.e., **he** knows most of the structure **ah initio**. **The** known piece of

structure is input as a **superatom**; DENDRAL then is of assistance in **analyzing** the unknown part and connecting all parts to **form** complete **molecules**.

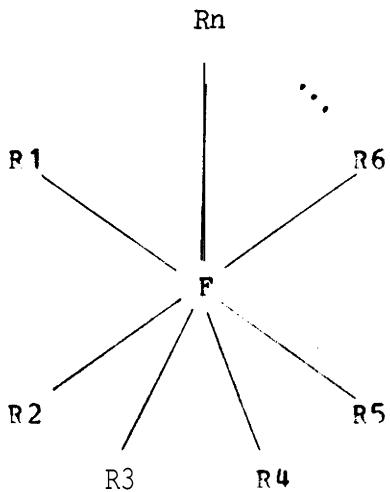
Aggregation, as just described, is a part of **the more** formal, more organized, **more** complete **heuristic process** in DENDRAL that we call planning.\* We have organized the planning

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\*The **aggregation** heuristics are currently the most important parts of our planning process, **but** not **the** only parts. For example, the heuristics which infer the **weights** of **radicals** attached to the central **subgraph** (see discussion in **text**) for later use **in** search control in the generator are not aggregation heuristics. Planning, in our view, can be a much broader process than **just** aggregation. A plan can contain any information that subsequently will be useful in controlling the search for solutions.

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process **around** a planning model shown below:



where **F** is the key subgraph of the molecules (that which determines its chemical family), and R<sub>1</sub> . . . R<sub>n</sub> are the subgraphs (radicals) that are connected to it. At the planning stage in a particular analysis, more than one **F** may be possible. The number of radicals attached to the various possible **F**'s may differ.

A plan given to **the** Structure Generator by the Planner consists of:

1. one or more **F**'s, as superatoms
2. for each **F**, the "**molecular**" weights of the radicals **attached** to the various valence bonds
3. other information about aggregation.

The plan delineates the subset of the set of **all** plausible structures that will be allowed as solution candidates, **in**

effect, it determines that the search for solutions will take place in some particular **subtree** of the **DENDRAL** space. How far **below the root** of the **space** (i.e., how much of the "upper levels" need not be searched) is a function of how much aggregation there is in the **F's**.

In the early forms of the planning process (previously called a "preliminary inference" process), the **F's** and the pattern recognition rules for identifying **F's** were determined in basically an ad hoc fashion, by the thorough, careful but painstaking technique involving chemist, computer, and **DENDRAL** staff member that has been described as "Eliciting a Theory from an Expert". (Buchanan, et al, 1970). In a series of carefully chosen steps up the ladder of structural and mass spectral complexity, heuristically powerful sets of **F's** and rules for the **acyclic** monofunctional (i.e., one **F** at a time) chemical families were worked out. The aggregation heuristics previously discussed were employed. The Planner developed into the system's "specialist" on the meaning of spectral lines--3 collection of special f-acts and special.-purpose heuristics organized around particular chemical families.

The use of the Planner as a specialist controlling a general search process is powerful. Results for the analysis of mass spectra of the chemical **families** of ketones and ethers are illustrative. See Tables 4 and 5. The differences between

numbers of structures in the columns labeled "Number of Chemically Plausible Structures" and the columns labeled "Number of Structures Generated" exhibit the power of planning in limiting search in **these problems.**

#### THE PLANNING PROCESS

The primary fact of life for heuristic program designers is that increases in complexity of problems are accompanied by exponential increases in the size of the problem spaces to be searched. Successful heuristic designs cope **by** increasing the number **and/or** power of **the** heuristics to match increases in the size of the space.

The chemical family of **amines** presents such a challenge for DENDRAL. **Amines** contain a nitrogen atom **as** the key heteroatom. Since nitrogen has three valence bonds compared with **oxygen's** two, **amines** represent the next logical step up in complexity from the ketones and ethers. For any **fixed** number of carbon atoms there are many more **amines** than **either** ketones or ethers. That is, there is a marked increase in **the** size of the spaces to **be** searched.

Early experiments with **amines** showed the usual pattern of system breakdown symptomatic of too little heuristic power for

the size of the spaces. Since for **amines** the a priori stability heuristics that define chemical plausibility for the generator have little or no heuristic power, all of **the** heuristic control over the generator must come from the plan. Producing plans simply by extrapolating the techniques used for the ketone and ether families was grossly inadequate.

In such a situation, a **sensible** design change is to give the Planner the ability to specify **more completely** the form of acceptable solution candidates. The generator is thereby constrained to search a smaller space. One way to do **this** is by more **aggregation**--to cause more pieces or larger pieces of structure to be "predetermined" by special-purpose inference schemes.

In the DENDRAL development, increased **aggregation** in the planning stage was **designed** in as follows:

1. In a systematic way, the size of the **F's** was increased to incorporate more carbon and hydrogen atoms. **If** the set of **F's** is to be logically complete within the size bounds chosen, then by the ordinary **combinatorics**, the number of possible **F's** from which selections will be made must increase. This complicates the classification decision by which it is inferred that the spectral data indicates a particular **F** (or set of **F's**).

The systematic method used for enumerating the set of **F's** for **amines** was chosen very **carefully** to mate best with that part of the theory of mass spectrometry that seemed most powerful in aiding the classification decision. The system for constructing the **F's** and the mass spectral theory to which it mates (alpha-carbon fragmentation theory) are described in detail elsewhere (Buchs, et al, 1970) and will not be explicated here.

2. Heuristics for the interpretation of nuclear magnetic resonance spectra were added to the Planner. As previously mentioned, these auxiliary data **are** useful for inferring the number of CH<sub>3</sub> superatoms in **the** structure (also how many of these superatoms are linked to a carbon, how many **to** the heteroatom). A complete interpretation of the NYR spectrum often is impossible to make, whether the interpreter is human or DENDRAL, but in any event is not necessary. Whatever partial interpretation can be done unambiguously by the heuristics will be reflected in the plan by corresponding aggregation information,

A new Planner (for historical reasons called "**Inference Maker**" in Buchs, et al, 1970) implements these ideas. The structure of this program is very simple, but the mass spectrum interpretation heuristics are quite complex. These rules developed **by** the **DENDRAL** group stand on their own as a

contribution to the methodology of **mass** spectrum analysis.

Because of their complexity, however, they are best applied by a computer **program**, not a human chemist, giving DENDRAL a substantial performance edge over human analysts for the class of problems handled by the rules.

The Planner has the following organization:

1. If an **NMR** spectrum is given as problem information, infer all that can be inferred about the **methyl** superatoms. Include this information in the plan. In addition, use it in the test part of step 4 below.
2. Generate a list of the relevant **F's** for the chemical family being **considered** (for example, generate the 31 **F's** relevant to **amines**).
3. Associate with each **F** a property list which contains a number of criteria of applicability ("diagnostic" criteria) for that **F**. In large measure these criteria are inferred from mass spectral theory. (We mentioned earlier that the method of structuring the **F's** was chosen to make this application of theory easy.)
4. Test each superatom against the given mass spectrum to ascertain whether all of the "diagnostic" criteria for it are satisfied by the data. If any part of this validation test series fails, discard the **F**.
5. All **F's** not discarded are included in the plan. For each of these, infer the weights of the attached radicals from the

spectral data and include these sets **of weights** in the plan.

Table 6 exhibits the results of using this planning process on a group of amine compounds. There **are some** noteworthy things about the data in this table, for example:

1. The size of the problem spaces for some. of the **amines** (over **14** million isomers of **C<sub>20</sub>H<sub>43</sub>N!**);
2. The impotence of the mass spectrum alone in finding the **answer** (or a small set of **answers**). This difficulty is not caused by a lack of expertise in the **program**. **Human** experts are in exactly the **same** situation, or perhaps worse.
3. The extraordinary effect of the NMR data to assist the mass spectrum analysis. Every time a "**1**" appears in the right **most column, it.** indicates that the plan contained **so much** information about the solution, that the plan in fact uniquely determined the solution? Even in the other cases, the number of isomers in the plan-constrained space is trivially **small**.

This is remarkable, The **Planner**, which is the specialist **at "understanding"** the data and inferring conditions on the solution, is so powerful that the need for the general problem solving processes of the system is obviate?, Another way **to** view this is that all the **relevant** theoretical knowledge to solve these amine problems has been mapped over from its general form in the **Predictor ("first principles")** to efficient special forms in the Planner ("cookbook **recipes**").

The details of how each specialist works have been described elsewhere. In each particular case, new constraints on the problem lead to new heuristics for shortcircuiting the general combinatorial theory. When the shortcuts can be discovered a specialist emerges; otherwise, the program relies on its **general capabilities**.

On the average, the problems of Table 6 each took about 0.5 seconds of computer time to solve, whereas the average ketone or ether problem shown in previous tables took a few minutes to solve; and the average amine problem done by the method used for the ketones would take much longer.

#### PLANNING RULE GENERATOR

At this point, we will review the most important features of the planning process.

Though it houses a few general practitioners performing aggregation, the Planner is primarily a house of specialists. The areas of specialty are chemical families such as ketones, ethers, and amines. One process makes the necessary plan-formulation decisions for all the specialists. The expertness of a specialist is contained in what it knows about its family of specialization, particularly the expected

patterns of mass spectral lines for a set of subclasses of **the** family.

There is, in effect, an N-position switch at the very front end of DENDFAL, which is set when a heuristic procedure or human intervention **declares** the family of **molecules** to be considered,\*

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\*Deciding on an appropriate **setting** of the switch may involve some "**active**" processing, e.g., some search. **Unless told** by human intervention, **DENDRAL does** not know at **the outset** what the appropriate specialist. is. **It** discovers this **by some** trial and error search. This involves, first, **guessing the** correct heteroatom (assuming that the empirical formula is not **given**). If, as a **result** of this guess, the specialist **that** is appropriate can not validate even one **F**, a "backtracking" takes place in which the **guess** is abandoned, and a new guess as to heteroatom is made.

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Setting **the switch** calls the appropriate specialist. If there is none, the switch is set to a default position which calls only general practitioners. The specialist knows how to generate the central superatoms relevant to its family **and the** associated validation criteria for each superatom.

The specialist was given this information by us, the designers. The designers, who know the theory of mass spectrometry, have selected some of this theory--first order effects--as the basis for a preliminary interpretation of the data. The slice of theory so selected determines what size and structural form the central superatoms must have. The designers then deduce the actual structures of all of the logically possible central superatoms of that size and form. The designers also deduce from the-first-order theory specific values for the validation criteria to be associated with each central superatom. The results of these two deductive steps (superatoms and criteria) taken together constitute a set of planning rules to be used at the time the specific plans are formulated. Thus a set of planning rules makes the Planner a specialist for a chemical family. Once alive and tested, the new specialist is added to the "big switch?

It is evident that when the designer has chosen the slice of theory he wishes to use for planning purposes, the remainder of his work, the generation of planning rules, can be, in fact should be, done by program. As the molecular families treated become more complex; necessitating the addition of heuristic power in the planning stage if the generator is to be properly controlled, the planning analysis involves increasingly more theory, which in turn leads to increased difficulty+ for humans in generating logically complete and accurate sets of planning

rules. In addition, a Planning Rule Generator program can create, automatically, specialists for each of the member-families of the broad class of families to which the theory now applies. This is an automatic mass production process that can replace the tedious and expensive process of eliciting knowledge from an expert that we have used in the past.

A Planning Rule Generator has been written for DENDRAL. It deals with the very general class of saturated (i.e., no double bonds or rings), acyclic monofunctional compounds. Plan schemes have been generated by this program for the following families: thiols and thioethers (heteroatom is sulphur); ethers; alcohols; and amines. These planning rules were then used by DENDRAL in solving problems in these areas (i.e., the ordinary DENDRAL performance mode). The results are shown in Tables 5, 6, and 7. The comments we made earlier concerning Table 6 apply also to Tables 5 and 7.

The Planning Rule Generator is a complex program, the details of which can not be described here. Those interested can find a description of the program from a chemical point of view in a recent publication (Buchs, et al, 1970).

The DENDRAL Planner is a performance process. The Planning Rule Generator is not. It is a higher level planning process

by which it is determined how planning shall **be done** in particular classes of problems. For us it is the first small step up the ladder of programs for theory manipulation and theory formation "**meta**" to the DENDRAL performance program, we view the building of such programs as a promising endeavor. DENDRAL as a performance program is complex. enough and rich enough in internal structure and theory to provide many firm foundation points on which to erect a **meta-level** for the study of theory formation processes.

#### GENERALITY AND THE DESIGNS FOR PROBLEM **SOLVING** SYSTEMS

We **shall** conclude this paper with a return to the theme with which we began: generality, expertness, and the design of problem solvers. **As** a case study, we have traced the evolution of designs for a **system** that solves difficult scientific inference problems. The forcing function for the evolution of designs was **primarily** the set of demands placed upon the organization of the DENDRAL program by increasingly more complex **and** difficult tasks. The design we now have is "**natural**" (i.e., shaped **by** the real world), not "**artificial**" or **abstract**.

**Many** threads have been woven into our discussion: general processes and representations in DENDRAL; the cost of

generality; heuristic power; the specialization of knowledge in the planning process; planning as a method for translating problem data into search constraints and solution conditions; higher-level planning as a method for building specialists from general theory. We now ask whether these threads form a meaningful fabric.

The study of generality in problem solving has been dominated by a point of view that calls for the design of "**universal**" methods and "**universal**" problem representations. These are **the** GPS-like and Advice Taker-like models. This approach to generality has great appeal, **but there** are difficulties intrinsic to it: the difficulty of translating specific tasks into the general representation; and the tradeoff between generality and power of the methods.

**In** recognition of **these** difficulties, a viewpoint at the other extreme has emerged, informally called "**the** big switch hypothesis?\*

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\***We** first heard **the** phrase "**big switch hypothesis**" in a lecture given by **A Newell** at Stanford University in 1966.

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In this view, general problem solvers are too weak to be used as the **basis** for building high-performance systems. The

behavior of the **best general** problem solvers we know, human problem solvers, is observed to be weak and shallow, except in the areas in which the human problem solver is a specialist. And it is observed that the transfer of expertness between specialty areas is slight. A chess master is unlikely to be an expert algebraist or an expert mass spectrum analyst, etc. In this view, the expert is the specialist, with specialist's knowledge of his area and specialist's methods and heuristics.

The "**big switch hypothesis**" holds that generality in problem solving is achieved by arraying specialists at the terminals of a big switch. The big **switch** is moved from specialist to specialist as the problem solver switches its attention from one problem area to **another.\***

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\*In this paper, we merely **state** the hypothesis without discussing it. The kinds of problem solving processes, if any, which are involved in "Wetting the **switch**" (selecting a specialist) is a topic that obviously deserves detailed examination in another **paper**.

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Our case study of the DENDRAL program **suggests** a synthesis of these extreme points of view. The features that characterize a general problem solving process are present. **Within** the

DENDRAL world, the search for solution **candidates** in the Structure Generator and the validation procedure of **the Predictor** are "**universal**" **methods**, and the representation employed is "**universal**". The general methods do solve DENDRAL problems, sometimes well as with **some** amino acid spectra, but they **are** relatively weak and inefficient.

To increase accuracy and efficiency, specialists emerged, but in a design which **called** for compatibility and coexistence with the general **processes**. The existing internal representation was maintain& throughout as a "common language" understood by both generalist and specialist. The specialists did not replace the generalists. They were written to function as planners, providing search constraints and solution conditions. The "**big** switch" in **DENDRAL** is at the front **end** of the Planner Program. Despite the array of powerful specialists on the switch, perhaps **the** most important position is the default position--the "**not** elsewhere classified" **bypass**--that calls the general problem solving processes when the **knowledge** of a specialist is not available.

**The** Planning' Rule Generator makes the symbiosis of generalist and specialist **mutual**. **The** theory of mass spectrometry that is used by the Predictor to validate candidates (or some part of it) **is used** by the Planning Rule Generator to **deduce** a new specialist for the "**big** switch?

Herein we think lies the germ of another method for problem solvers. A general problem solving process in part achieves generality because it employs a general theory **of** the nature and behavior of **the** objects and operators of its world. This theory can be used in **what we** might call "**execute mode**", as for example when **DENDRAL's Predictor** is validating a candidate **solution**. **But** this theory can also be used in what might **be** called "**compile mode**", as for example when the Planning Rule Generator is deducing a new specialist.

This idea needs an extended discussion, which we are not prepared to give here. But we shall **make a** few brief observations.

The first observation is **that** the idea closely parallels **the** line of argument given by Simon in his book of essays on heuristic programming entitled "**The New Science of Management Decision**" (Simon, 1960). In discussing human decision making, particularly **in** organizations, Simon draws a dichotomy between the **routine** repetitive decision problems, which he calls "**programmed decisions**" and the novel one-shot decision problems, which he calls "**nonprogrammed decisions**". Concerning "**programmed decisions**", the organization "**develops** specific processes for handling **them.**" Examples are: habits (an individual's "**compiled subroutines**"), Standard Operating

Procedures (an organization's "compiled **subroutines**"), mathematical models from **Operations** Research, and EDP procedures. The "non programmed" decision **problems** are "handled" by general problem-solving processes? To a large extent., it is the **repetitiveness** with which a decision problem presents itself that determines whether it is economic for an organization to invest resources in **routinizing** and **specializing** the decision making process, i.e., "compile" **general processes** into special-purpose routines.

The second observation is that **the idea** may be **much** more difficult to implement than it appears at first for the simple reason that the tradeoff **between** generality and power holds for processes at the **meta-level** just as it holds for performance level processes. Thus, for **example**, **DENDRAL**'s Planning Rule Generator is powerful for ths supra-family of all saturated, acyclic, monofunctional compounds, but is useless for all other classes of compounds. When we extend **DENDRAL**'s capability **to families** of cyclic molecules, we **may** have to write a new Planning Rule Generator. Or is there yet another process lurking at a higher level, a Generator of **Planning Rule** Generators?

The **appropriate** place for an attack on the problem of generality **may be at** the **meta-levels** of learning, knowledge transformation, and representation, not at. the level of

performance programs. Perhaps for the designer of intelligent systems what is most significant about human general problem solving behavior is the ability to learn specialties as **needed--to** learn expertness in problem areas by learning problem-specific heuristics, by acquiring problem-specific information, and by transforming general knowledge and general processes into specialized forms.



TABLE 1  
Numbers of Possible Non-Cyclic Molecular Structures of Selected Formulas (1)

Chemical Formula	Number of Carbon Atoms						
	4	5	6	7	8	9	10
C <sub>n</sub> H(2n+2)	2	3	5	9	18	35	75
C <sub>n</sub> H(2n+2)O	7	14	32	72	171	405	989
C <sub>n</sub> H(2n+3)N	8	17	39	89	211	507	1238
C <sub>n</sub> H(2n+3)NO							

(1) These numbers define the size of the search space for problems involving molecules of a given chemical formula. The size of the space increases dramatically with both the number of carbon atoms and the number of other types of atoms in the formula. This table is abstracted from Lederberg, et al, 1969.

TABLE 2  
Amino Acid Results

Name of "Unknown"	Chemical Formula	Size of Problem Space (1)	Number of Plausible Structures (2)	Number of Structures Generated (3)	Rank Order of Correct Candidate (4)
Glycine	C <sub>2</sub> H <sub>5</sub> NO <sub>2</sub>	38	12	8	1st, 7 excluded
Alanine	C <sub>3</sub> H <sub>7</sub> NO <sub>2</sub>	216	50	3	1st
Serine	C <sub>3</sub> H <sub>7</sub> NO <sub>3</sub>	324	40	10	1st, 9 excluded
Threonine	C <sub>4</sub> H <sub>9</sub> NO <sub>3</sub>	1758	238	2	1st
Leucine	C <sub>6</sub> H <sub>13</sub> NO <sub>2</sub>	10000 (approx.)	3275	288	Tied for 2nd, 277 excluded

(1) The total size of the problem space is the number of topologically possible molecular structures generated within valence considerations alone.

(2) The number of plausible structures is the number of molecular structures in the total space which also meet a priori conditions of chemical stability. The a priori rules have greater effect with increased numbers of non-carbon, non-hydrogen atoms.

(3) The number of structures generated is the number of molecular structures actually generated by the program as candidate explanations of the experimental data. Pruning has been achieved by using the "zero-order" theory during structure generation.

(4) The rank order of the correct structure is the validation program's assignment of rank to the actual molecular structure used as a test "unknown". The number of structures excluded in the validation process is also indicated.

TABLE 3  
Heuristics Used for Identifying Three Superatoms (1)

Superatom		Identifying Conditions
Name	Structure	
Ketone	$\text{--C}^{\text{O}}\text{--}$	<p>There are 2 peaks at mass units <math>x_1</math> &amp; <math>x_2</math> such that</p> <ol style="list-style-type: none"> <li><math>x_1 + x_2 = M + 28</math>,</li> <li><math>x_1 - 28</math> is a high peak,</li> <li><math>x_2 - 28</math> is a high peak,</li> <li>At least one of <math>x_1</math> or <math>x_2</math> is high.</li> </ol>
N-Propyl Ketone <sub>3</sub>	$\text{CH}_3\text{--CH}_2\text{--CH}_2\text{--C}^{\text{O}}\text{--CH}_2\text{--C--CH}_3$	<ol style="list-style-type: none"> <li>71 is a high peak,</li> <li>43 is a high peak,</li> <li>86 is a high peak,</li> <li>58 appears with any intensity.</li> </ol>
Ether	$\text{--C}^{\text{--O}}\text{--C--}$	<ol style="list-style-type: none"> <li><math>M-18</math> is a peak of 0 or 1% intensity,</li> <li><math>M-17</math> is a peak of 0 or 1% intensity,</li> <li>There are 2 peaks corresponding to the alpha-cleavage fragments.</li> </ol>

(1) See Duffield, et al. (1969), Schroll, et al. (1969), and Buchs, et al. (1970) for fuller discussions of these and other sets of heuristics used in planning.

TABLE 4  
Ketone Results

Name of "Unknown"	Chemical Formula	Size of Problem Space (1)	Number of Plausible Structures (2)	Number of Structures Generated (3)	Rank Order of Correct Candidate (4)
2-Butanone	C4H8O	11	11	1	1st
3-Pentanone	C5H10O	33	33	1	1st
3-Hexanone	C6H12O	91	91	1	1st
2-Methyl-hexan-3-one	C7H14O	254	254	1	1st
3-Heptanone	C7H14O	254	254	2	Tied for 1st
3-Octanone	C8H16O	698	698	4	1st
4-Octanone	C8H16O	698	698	2	1st, 1, excluded
2,4-Dimethyl-hexan-3-one	C8H16O	698	698	4	Tied for 1st, 1 excluded
6-Methyl-heptan-3-one	C8H16O	698	698	4	1st
3-Nonanone	C9H18O	1936	1936	7	1st
2-Methyl-octan-3-one	C9H18O	1936	1936	4	1st (5)
4-Nonanone	C9H18O	1936	1936	4	1st (5)

- (1) The total size of the problem space is the number of topologically possible molecular structures generated within valence considerations alone.
- (2) The number of plausible structures is the number of molecular structures in the total space which also meet a priori conditions of chemical stability. The a priori rules have no effect with formulas containing a single non-carbon, non-hydrogen atom.
- (3) The number of structures generated is the number of molecular structures actually generated by the program as candidate explanations of the experimental data. Pruning has been achieved by using the planning information from the Planning program.
- (4) The rank order of the correct structure is the validation program's assignment of rank to the actual molecular structure used as a test "unknown". The number of structures excluded in the process is also indicated.
- (5) Previous publication showed the correct structure excluded. The general rules of the program have since been modified to improve its performance.

Table 5  
Ether and Alcohol Results

<u>Alcohol</u>		Number of $C_nH_{2n+2}O$ isomers(1,2)		Number of inferred isomers(3)		<u>Ether</u>	Number of $C_nH_{2n+2}O$ isomers(1,2)		Number of inferred isomers(3)	
		A	B	A	B		A	B	A	B
n-butyl	C4	7		2	1	Methyl-n-propyl	C4	7	2	1
iso-butyl		7		2	1	Methyl-iso-propyl		7	3	1
sec-butyl		7		5	2	Methyl-n-butyl		14	2	1
2-methyl-2-butyl	C5	14		1	1	Methyl-iso-butyl		14	2	1
n-pentyl		14		4	1	Ethyl-iso-propyl		14	1	1
3-pentyl		14		1	1	Ethyl-n-butyl	C6	32	4	1
2-methyl-1-butyl		14		4	2	Ethyl-iso-butyl		32	4	2
2-pentyl		14		2	1	Ethyl-sec-butyl		32	2	2
3-hexyl	C6	52		2	1	Ethyl-tert-butyl		32	1	1
3-methyl-1-pentyl		32		8	4	Di-n-propyl		32	1	1
4-methyl-2-pentyl		32		4	1	Di-iso-propyl		32	1	1
n-hexyl		32		8	1	n-propyl-n-butyl	C7	72	2	1
3-heptyl	C7	72		4	1	Ethyl-n-pentyl		72	4	1
2-heptyl		72		8	1	Methyl-n-hexyl		72	8	1
3-ethyl-3-pentyl		72		1	1	iso-propyl-sec-butyl		72	3	2
2,4-dimethyl-3-pentyl		72		3	1	iso-propyl-n-pentyl		171	4	1
n-heptyl		72		17	1	n-propyl-n-pentyl		171	4	1
3-methyl-1-hexyl		72		17	6	Di-n-butyl		171	3	1
n-octyl	C8	171		39	1	iso-butyl-tert-butyl		171	2	1
3-octyl		171		8	1	Ethyl-n-heptyl	C9	405	34	1
2,3,4-trimethyl-3-pentyl		171		3	1	n-butyl-n-pentyl		405	8	1
n-nonyl	C9	405		89	1	Di-n-pentyl	C10	989	10	1
2-nonyl		405		39	1	Di-iso-pentyl		989	18	7
n-decyl	C10	989		211	1	Di-n-hexyl	C12	6045	125	2
6-ethyl-3-octyl		989		39	9	Di-n-octyl	C16	151375	780	1
3,7-dimethyl-1-octyl		989		211	41	Bis-2-ethylhexyl		151375	780	21
n-dodecyl	C12	6045		1238	1	Di-n-decyl	C20	11428365	22366	1
2-butyl-1-octyl		6045		1238	25					
n-tetradecyl	C14	38322		7639	1					
3-tetradecyl		38322		1238	1					
n-hexadecyl	C16	151375		48865	1					

A = Inferred isomers when only mass spectrometry is used.

B = Inferred isomers when the number of methyl radicals is known from NMR data.

- (1) The total size of the problem space is the number of topologically possible molecular structures generated within valence considerations alone.
- (2) The number of plausible structures is the number of molecular structures in the total space which also meet a priori conditions of chemical stability. The a priori rules have no effect with formulas containing a single non-carbon, non-hydrogen atom.
- (3) The number of structures generated is the number of molecular structures actually generated by the program as candidate explanations of the experimental data. Pruning has been achieved by using the planning information from the Planning program.

Table 6

## Amine Results

Amine	Number of $C_nH_{2n+3}N$ isomers	Number of inferred isomers		Amine	Number of $C_nH_{2n+3}N$ isomers	Number of inferred isomers	
		A	B			A	B
n-propyl	C3	4	1	1	N-methyl-di- <u>iso</u> -propyl	C7	89
<u>iso</u> -propyl		4	2	1	<u>iso</u> -octyl	C8	211
n-butyl	C4	8	2	1	Ethyl-n-hexyl		211
<u>iso</u> -butyl		8	2	1	1-methylheptyl		211
<u>sec</u> -butyl		8	4	2	2-ethylhexyl		211
<u>tert</u> -butyl		8	3	1	1,1-dimethylhexyl		211
Diethyl		8	3	1	Di-n-butyl		211
N-methyl-n-propyl		8	4	1	Di- <u>sec</u> -butyl		211
Ethyl-n-propyl	C5	17	5	1	Di- <u>iso</u> -butyl		211
N-methyl-diethyl		17	4	1	Di-ethyl-n-butyl		211
n-pentyl		17	4	1	3-octyl		211
<u>iso</u> -pentyl		17	4	2	n-nonyl	C9	507
2-pentyl		17	2	1	N-methyl-di-n-butyl		507
3-pentyl		17	5	1	Tri-n-propyl		507
3-methyl-2-butyl		17	4	1	Di-n-pentyl	C10	1238
N-methyl-n-butyl		17	4	1	Di- <u>iso</u> -pentyl		1238
N-methyl- <u>sec</u> -butyl		17	3	1	N,N-dimethyl-2-ethylhexyl		1238
N-methyl- <u>iso</u> -butyl		17	4	1	n-undecyl	C11	3057
n-hexyl	C6	39	8	1	n-dodecyl	C12	7639
Tri-ethyl		39	2	1	n-tetradecyl	C14	48865
2-hexyl		39	8	1	Di-n-heptyl		48865
Di-n-propyl		39	8	1	N,N-dimethyl-n-dodecyl		48865
Di- <u>iso</u> -propyl		39	8	1	Tri-n-pentyl	C15	124906
N-methyl-n-pentyl		39	8	1	Bis-2-ethylhexyl	C16	321988
N-methyl- <u>iso</u> -pentyl		39	8	2	N,N-dimethyl-n-tetradecyl		321988
Ethyl-n-butyl		39	6	1	Di-ethyl-n-dodecyl		321988
N,N-dimethyl-n-butyl		39	10	1	n-heptadecyl	C17	830219
n-heptyl	C7	89	17	1	N-methyl-bis-2-ethylhexyl		830219
Ethyl-n-pentyl		89	16	1	n-octadecyl	C18	2156010
n-butyl- <u>iso</u> -propyl		89	11	1	N-methyl-n-octyl-n-nonyl		2156010
4-methyl-2-hexyl		89	16	4	N,N-dimethyl-n-octadecyl	C20	14715813
							1284792
							1

A = Inferred isomers when only mass spectrometry is used.

B = Inferred isomers when the number of methyl radicals is known from NMR data.

- (1) The total size of the problem space is the number of topologically possible molecular structures generated within valence considerations alone.
- (2) The number of plausible structures is the number of molecular structures in the total space which also meet a priori conditions of chemical stability. The a priori rules have no effect with formulas containing a single non-carbon, non-hydrogen atom.
- (3) The number of structures generated is the number of molecular structures actually generated by the program as candidate explanations of the experimental data. Pruning has been achieved by using the planning information from the Planning program.

Table 7

## Thioether and Thiol Results

Thioether	Number of $C_nH_{2n+2}S$ isomers(1,2)	Number of Inferred isomers(3)	Thiol	A		Number of $C_nH_{2n+2}S$ isomers(1,2)	A	
				B			B	
Methyl-ethyl	C3	3	1	1	n-propyl	C3	3	2
Methyl-n-propyl	C4	7	1	1	iso-propyl		3	1
Methyl-iso-propyl		7	2	1	n-butyl	C4	7	3
Di-ethyl		7	1	1	iso-butyl		7	1
Methyl-n-butyl	C5	14	5	1	tert-butyl		7	1
Methyl-iso-butyl		14	5	2	2-methyl-2-butyl	C5	14	1
Methyl-tert-butyl		14	1	1	3-methyl-2-butyl		14	2
Ethyl-iso-propyl		14	1	1	3-methyl-1-butyl		14	6
Ethyl-n-propyl		14	2	1	n-pentyl		14	4
Ethyl-n-butyl	C6	32	3	1	3-pentyl		14	5
Ethyl-tert-butyl		32	1	1	2-pentyl		14	6
Ethyl-iso-butyl		32	3	2	n-hexyl	C6	32	8
Di-n-propyl		32	2	1	2-hexyl		32	12
Methyl-n-pentyl		32	10	1	2-methyl-1-pentyl		32	8
Di-iso-propyl		32	1	1	4-methyl-2-pentyl		32	4
Ethyl-n-pentyl	C7	72	4	1	3-methyl-3-pentyl		32	1
n-propyl-n-butyl		72	5	1	2-methyl-2-hexyl	C7	72	8
iso-propyl-n-butyl		72	5	2	n-heptyl		72	17
iso-propyl-tert-butyl		72	1	1	2-ethyl-1-hexyl	C8	171	39
n-propyl-iso-butyl		72	3	2	n-octyl		171	9
iso-propyl-sec-butyl		72	4	3	1-nonyl	C9	405	89
n-propyl-n-pentyl	C8	171	4	1	n-decyl	C10	989	211
Ethyl-n-hexyl		171	8	1	n-dodecyl	C12	6045	1238
Di-n-butyl		171	5	1				
Di-sec-butyl		171	3	1				
Di-iso-butyl		171	3	1				
Methyl-n-heptyl		171	21	1				
Di-n-pentyl	C10	989	12	1				
Di-n-hexyl	C12	6045	36	1				
Di-n-heptyl	C14	38322	153	1				

A = Inferred isomers when only mass spectrometry is used.

B = Inferred isomers when the number of methyl radicals is known from NMR data.

- (1) The total size of the problem space is the number of topologically possible molecular structures generated within valence considerations alone.
- (2) The number of plausible structures is the number of molecular structures in the total space which also meet a priori conditions of chemical stability. The a priori rules have no effect with formulas containing a single non-carbon, non-hydrogen atom.
- (3) The number of structures generated is the number of molecular structures actually generated by the program as candidate explanations of the experimental data. Pruning has been achieved by using the planning information from the Planning program.



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