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Integrated Expert System Development Environment With an Automatic Verification Feature: An Application To Grinding Process

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by

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INTEGRATED EXPERT SYSTEM DEVELOPMENT ENVIRONMENT WITH AN AUTOMATIC VERIFICATION FEATURE: AN APPLICATION TO GRINDING PROCESS

Final Report

Submitted to the faculty of Miami University in partial fulfillment of the requirements for the degree of Master of Systems Analysis Department of Systems Analysis Miami University

by

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ABSTRACT

The concerns of this Master degree project are the implementations of an expert system development environment and an application to manufacturing. The environment is given the name of IDEA (Integrated expert system Development Environment with Automatic verification feature), and it integrates an expert system shell with a rule base verification facility and a rule base builder into a single program. The interface of IDEA has the look of a general purpose computer language. The knowledge bases referenced by IDEA consist of IF-THEN rules with the syntax of Object-Attribute-Value (O-A-V) triples and confidence factors. An application is developed for grinding chattering diagnosis problems using the IDEA.

The verification facility of IDEA realizes the main functions of CHECK, a verification tool, developed by T. Nguyen *et al*, 1987, so that it can identify some major problems of inconsistency and incompleteness. Different from CHECK, IDEA uses a unique approach based on rule tree algorithm to check for some logic problems in a rule base. IDEA also has the ability to draw the rule tree on a computer screen. This report discusses the problems of inconsistency and incompleteness in detail, including the influence of introducing confidence factors. Some problems identified in this report are not reported before.

The rule base builder of IDEA uses some techniques that can greatly reduce errors in a rule base during its input phase and greatly speed up the input work. First, the builder uses templates for rule input to eliminate the possible syntax errors of rules. Second, it uses an O-A-V file for declared O-A-V triples. By checking the O-A-V file and the O-A-V triples used in the rule base, the problems of illegal O-A-V triples and unreferenced O-A-V triples can be identified. Finally, it has the ability to automatically generate the ask-part of the rule base, so that no errors will exists in the ask-part and the input work can be efficient.

Grinding chattering is a kind of abnormal phenomenon in a grinding process, which greatly degrades the surface quality of the workpiece ground. The grinding chattering diagnosis rule base gives complete answers for the causes of chattering for external cylindrical grinding. This report describes the rule base development work and its related grinding knowledge.

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1. INTRODUCTION

The original objective of this research was to develop a knowledge acquisition environment for a rule-based expert system and an application to manufacturing system. As the project progressed, more emphasis was put on the expert system development environment. The environment is a rule-based expert system shell integrated with editing and verification facilities. The manufacturing application was developed with the guidance of Associate Professor Myron. Schmenk, from the Manufacturing Engineering Department.

This report is intended to provide a description of the expert system development environment developed in the Master degree project, and to discuss its associated concepts and the techniques. Chapter 2 introduces the concept of verification and discusses the potential problems in a knowledge base. Chapter 3 describes the work of developing the environment and explores the possible approaches. Chapter 4 describes the developing work for the grinding chattering diagnosis rule base.

In the following sections, the general concepts and terminology of expert systems and knowledge representation formalisms are introduced, the domain selection work is described, and a general description of the expert system environment developed is provided.

1.1 Expert Systems

Early expert systems arose in the late 1960's to the early 1970's. Well-known examples are

- DENDRAL, 1965, developed to identify the chemical molecular structures;
- MACSYMA, 1969, the first mathematics expert system; and
- MYCIN, 1973, an expert system for medical diagnosis.

Since then, expert systems have experienced tremendous growth and popularity. Today, their applications are found in all fields, including science, medicine, military, business, and engineering.

- **Definition.** A variety of definitions for expert systems has been offered by several prominent researchers [11, 12, 13, 19, 20]. In summary, an *expert system* is a computer program that emulates the decision-making process of human domain experts. Based on the definition, expert systems should have two basic features:
 - 1. The problem that an expert system intends to solve is difficult enough to require significant human expertise, usually in a narrow domain, for its solution.
 - 2. Expert systems are symbolic reasoning oriented. Usually, judgments, rules of thumb, and intuition are embodied in an expert system.
- Structure. The basic structure of an expert system is shown in Figure 1.1. The working memory contains the specific facts/data for a problem and the intermediate to final results produced by the system. The inference engine is the overall controller for the system. It provides reasoning methods and produces conclusions, advice and explanations for the end use. The knowledge base contains required expertise. The knowledge acquisition facility is the

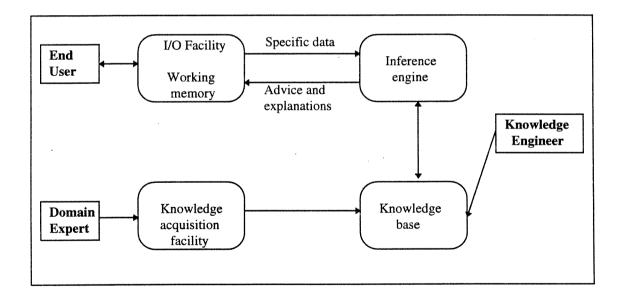


Figure 1.1 The structure of expert system

interface between domain experts and knowledge engineers. It includes methods and programs for developing, verifying and validating a knowledge base. Another important concept is *expert*

system shell, first introduced by MYCIN. Usually, an expert system shell consists of a user interface and an inference engine. Hence, the problem specific knowledge base is separated from the solution search procedures of the system.

1.2 Knowledge Representation

In principle, any consistent formalism in which the domain knowledge can be expressed can be used for knowledge base building. However, the most popular formalisms are *production rules*, semantic nets and frames. Their advantages and disadvantages are discussed below.

■ Production Rules. Because most of production rules have the IF...THEN format, they are also called *IF-THEN rules* (the terminology production rules and IF-THEN rules will be used alternately below). The corresponding expert systems are called *rule-based expert systems*.

Production rules have remained the most frequently used knowledge representation scheme. This resulted from their three significant advantages:

- Naturalness. A rule-based system mimics human experts in a natural way. IF-THEN
 rules could be easily understood and accepted by non-computer-oriented experts.
 This is a great benefit in knowledge acquisition.
- 2. *Modularity*. A rule base is highly modular. Each rule can be added, modified or deleted independently of other rules.
- 3. *Uniformity*. All knowledge in the rule base is represented in exactly the same way. This eases the work of rule base development.

On the other hand, production rules also have the following disadvantages:

- 1. Flatness. This formalism is quite flat and great numbers of rules are needed to model a real-world domain. This causes inefficiency when the rule base is executed.
- 2. Difficulty for tracing. Usually, knowledge in a rule base is not well organized, so it is often difficult to trace over the rule base. Efforts are made in this project to address this problem by introducing rule tree (see Section 3.3).
- 3. *Non-descriptive*. It is not natural to represent descriptive knowledge, such as the characteristics of an automobile, by using IF-THEN rules.

- Semantic Nets and Frames. Semantic nets and frames are two similar formalisms in which large sets of concepts and facts are directly attached to entities which they describe. Semantic nets and frames have the following advantages over production rules:
 - 1. *Inheritance*. Entities' properties could be inherited from their ancestors. The formalisms are deep in nature and the knowledge bases are easy to trace.
 - 2. *Efficiency*. Because facts are directly connected to related nodes or frames, the search time for a particular fact can be greatly reduced.

Frames also have the following additional advantages:

- 1. *Object-oriented nature*. Frames are actually objects, so they are adopted in a natural way by object-oriented programming. This is the major reason why frames are steadily growing in popularity.
- 2. *Procedure attachment*. As mentioned above, procedures can be attached in a frame. This feature greatly expands the application domains for frame-based systems.

The disadvantages of semantic nets and frames include:

- 1. Difficulty of interpretation. Because the knowledge is stated in a less English like way in semantic nets and frames, and there is lack of interpreting standards, the knowledge bases are difficult to understand by domain experts and may be interpreted in different ways.
- 2. *Non-heuristic*. It is difficult to represent heuristic knowledge without using IF-THEN rules.

1.3 Domain Selection

Selection of an appropriate domain is very important for an expert system, because it has significant influence over the knowledge representation scheme, the knowledge acquisition facilities and the inference engine. The domain chosen should meet the basic requirements:

- The domain is narrow enough to ensure the success of the project.
- The problems in the domain require symbolic reasoning for their resolutions.
- There exists at least one expert who is available and willing to work on the project.

In our case, a lot of effort has been spent in surveying the literature for an application area. The domain was finally limited to grinding problems, mainly because Associate Professor M. Schmenk, a metal cutting expert, showed interest in development of the knowledge base. Two possible tasks in grinding were identified:

- 1. grinding parameters selection; and
- 2. quality problem diagnosis.

The first task is suitable for data-driven expert systems but involves numerical calculations. The second task is suitable for goal-driven expert system and involves only symbolic reasoning. Therefore, the latter one is found to be more appropriate for the project. To further narrow the domain, the final task was chosen as grinding chattering diagnosis. Grinding problems are discussed in more detail in Chapter 4.

1.4 IDEA

IDEA (Integrated expert system Development Environment with Automatic verification feature) is the expert system development environment developed in this project, using Prolog. The motivation for IDEA is to develop a prototypal expert system environment that would have the appearance of a general purpose computer language. In other words, the objective for IDEA is to have functions analogous to the editing, compiling, debugging and running options in C or Pascal.

■ Selection of Knowledge Representation Formalism. IDEA uses IF-THEN rules to represent the knowledge. This decision was simply triggered by the availability of an inference engine that could be utilized for a rule-based system built in Prolog.

■ Syntax. A sample of toy rule base is shown in Appendix A. The rule base is headed by a *top-goal* "problem". This means the top-goal of the inference engine is to find what the "problem" is.

The body part of the rule base consists of IF-THEN rules. Each IF-THEN rule in turn consists of a rule ID, *left-hand side* (LHS) and *right-hand side* (RHS), and has the syntax:

```
Rule <RuleID>

IF [not] < Object1 > < Attribute1 > [ < Value1 > ]

[AND [not] < Object2 > < Attribute2 > [ < Value2 > ]

:

AND [not] < ObjectM > < AttributeM > [ < ValueM > ]]

THEN < ObjectN > < AttributeN > [ < ValueN > ] [cf < CF > ].
```

Where the contents inside <> are supplied by the knowledge engineer; the contents inside [] are optional; the cf stands for confidence factor; cf has a value in the range of [-100, 100]; the default value for Value is yes; the default value for cf is 100. The LHS consists of IF conditions. An IF condition includes an object-attribute-value triple (O-A-V triple) following an IF or an AND. For example, the LHS of the first rule in Appendix A is:

IF engine turn-over

AND auto battery bad

The RHS consists of only a *THEN conclusion*. The THEN conclusion in turn includes a *THEN*, an O-A-V triple and a confidence factor. There are two kinds of IF conditions: IF conditions that can be matched with other rules' THEN conclusions and IF conditions that can not be matched. The latter IF conditions are given the name of *leaf IF conditions*. Note that an Object-Attribute pair could be a variable. Because the inference engine used in IDEA does not have the ability to deal with variables, they are not introduced in the application rule base.

Following the rules is the ask-part of the rule base. This part consists of queries to acquire known basic facts from the end user. Each query corresponds to a leaf IF condition.

2. KNOWLEDGE BASE VERIFICATION

Verification is a part of the evaluation work for a written knowledge base. Different from validation, verification does not deal with expertise, instead, it checks for only the faultiness in an abstract knowledge base. In other words, a verified knowledge base is the one built in a "right" way, but not necessarily a "right" one itself.

T. Nguyen, Walton, Perkins and Pecora (1987) wrote that the problems in a knowledge base can be classified as inconsistency and incompleteness [1]. For inconsistency, they identified the potential problems as redundant rules, conflicting rules, subsumed rules, unnecessary IF conditions, and circular rules. For incompleteness, they identified the potential problems as unreferenced attribute values, illegal attribute values, unreachable conclusions, and dead-end IF conditions. Gonzalez and Dankel (1993) described the similar problems [13].

It should be pointed out that the potential problems in a knowledge base can be detected only if the syntax of knowledge representation is sufficiently restricted. More important, different kinds of knowledge base will have different kinds of problems. For the backward-reasoning IF-THEN rules used in IDEA, we identified the problems:

- 1. syntax error
- 2. illegal O-A-V triples
- 3. inconsistency
- 4. incompleteness

The first one is self-explanation. The second one refers to the O-A-V triples which have not been declared. In the case of IDEA, the illegal O-A-V triples are the triples that are not included in the O-A-V file (see Section 3.1). This problem is caused by typing mistakes and/or misusing terminology. The last two are discussed in more detail in the next two sections. This discussion attempts to be as complete as possible, including some inconsistency forms that are not reported before as well as the influence of introducing confidence factors, but it is limited in the scope of pure theory. The implementation work for the verification facility will be discussed in the next chapter. Again, the discussion is based on the rule base syntax defined in Section 1.4.

2.1 Checking for Consistency

Inconsistency refers to a logical incorrectness in a rule base. Six potential problems are identified:

■ Redundant Rules. Two rules are redundant if they have actually the same IF conditions and the same THEN conclusion. For example, consider the two rules that follow:

Rule 1

IF not engine turn-over

AND auto battery bad

THEN problem auto battery.

Rule 2

IF car battery bad

AND not engine turn-over

THEN problem car battery.

These two rules would be redundant, because they actually referred to the same conditions and the same conclusion, though Rule 2 used "car battery" instead of "auto battery" and its IF conditions were in a different order.

It should be pointed out that the above definition, based on T. Nguyen *et al* and Gonzalez *et al*, is only a special case of redundant rules. In most cases, a redundant rule would be the one that can be derived from the other two or more rules. For example, with the notation of predicate calculus, consider the set of rules that follow:

$$a(x) \to c(x)$$
; $a(x) \to b(x)$; $b(x) \to c(x)$.

Apparently the first rule can be derived from the other two rules. Though this case is still quite simple, it is already quite difficult to be detected by programs.

Note that with the introducing of confidence factors, two THEN conclusions that have the same O-A-V triple but different confidence factors should be regarded as different

conclusions. For example, if rule 1 and rule 2 have different confidence factors, they should not be regarded as redundant rules. Instead, they become conflicting rules (discussed below).

Although redundancy usually does not cause logic problems, it might affect efficiency. In such cases, it is better to eliminate the redundancy.

■ Conflicting Rules. Two rules are conflicting if they have actually the same IF conditions but different THEN conclusions. For example, consider the two rules that follow:

Rule 3

IF engine turn-over

AND auto radio weak

THEN auto battery bad.

Rule 4

IF engine turn-over

AND auto radio weak

THEN gas run-out.

These two rules are conflicting, because given the same IF conditions, different conclusions are derived. Conflicting rules could cause serious logic problems. Usually only one rule among the conflicting set is correct. The above description of conflicting rules is suitable for both cases of with or without confidence factors. The way to fix this problem is to identify the incorrect conflicting rules and eliminate them.

Note that conflicting rules do not necessarily cause logic problems. Under some circumstances, conflicting rules are legal and they work as a single rule in which the THEN conclusion has OR logical relations:

Rule < RuleID>

IF < condition 1>

AND < condition 2>

AND < condition m>

THEN < conclusion 1>

OR < conclusion 2>

OR < conclusion n >.

■ Subsumed Rules. One rule is subsumed by another if two rules have the same THEN conclusion but the first one's IF conditions is a subset of the second one's. For example, consider the two rules that follow:

Rule 5

IF auto lights weak
THEN auto battery bad.

Rule 6

IF auto lights weak

AND auto radio weak

THEN auto battery bad.

Rule 6 is subsumed by rule 5 because whenever rule 6 succeeds rule 5 also succeeds.

If rule 5 and rule 6 have different confident factors, they should not be regarded as subsumed rules, because their THEN conclusions become different. The following two rules are reasonable:

Rule 5'

IF auto lights weak
THEN auto battery bad cf 70.

Rule 6'

IF auto lights weak

AND auto radio weak

THEN auto battery bad cf 80.

The second IF condition in rule 6' could be purposely added to change the confidence. Because the second IF condition favors the THEN conclusion, rule 6' has a larger confidence factor value than that of rule 5'. Otherwise, a logic error must exist in the rule base.

Subsumed rules will cause logic errors. At most, only one rule among the subsumed set is correct. The way to fix this problem is to identify the right one and to delete the rest of the rules in the subsumed set.

■ Unnecessary IF Conditions. Two rules contain unnecessary IF conditions if they have the same THEN conclusion, a subset of IF conditions in the first rule is the negation of the correspondent IF conditions in the second rule, and all other IF conditions are equivalent. For example, consider the two rules that follow:

Rule 7

IF engine turn-over

AND auto radio weak

THEN auto battery bad.

Rule 8

IF not engine turn-over

AND auto radio weak

THEN auto battery bad.

The first IF condition in both rules is unnecessary, because the THEN conclusion will be true if only the second IF condition is true.

Unnecessary IF conditions will effect efficiency, although it does not cause logic errors. The way to fix this problem is to delete the unnecessary IF conditions and collapse the two rules into one. For example, the above two rules could be collapsed into the single rule

Rule 7'

IF auto radio weak THEN auto battery bad. If rule 7 and rule 8 have different confidence factors, one of the first IF condition could be still needed, so treat the two rules, as the case of rule 5' and rule 6'.

T. Nguyen *et al* pointed out the second form of unnecessary IF conditions [1]. With the notation of predicate calculus, this form can be represented as:

$$a(x)$$
 and not $b(x) -> c(x)$; $b(x) -> c(x)$.

To illustrate this consider the following two rules:

Rule 9

IF auto radio weak

AND not auto lights weak

THEN auto battery bad.

Rule 10

IF auto lights weak
THEN auto battery bad.

In this case, the second IF condition in rule 9 is unnecessary, but two rules are still needed. If rule 9 and rule 10 have different confidence factors, they could be reasonable.

There is still the third form of unnecessary IF conditions that is not reported. With the notation of predicate calculus, this form can be represented as:

$$a(x)$$
 and $b(x) -> c(x)$; $a(x) -> b(x)$.

To illustrate this, consider the following set of rules:

Rule 11

IF auto radio weak

AND auto battery bad

THEN problem auto battery.

Rule 12

IF auto radio weak

THEN auto battery bad.

In this case, the first IF condition in rule 11 is unnecessary, even if the two rules have different confidence factors. However, the two rules are still needed.

■ Circular Rules. A set of rules is circular if they form a closed reasoning chain. The following two rules are an example of circularity:

Rule 13

IF auto radio weak
THEN auto battery bad.

Rule 14

IF auto battery bad

THEN auto radio weak.

This loop is apparent. But if a loop is formed by a long reasoning chain, it will be very difficult to be detected by a knowledge engineer.

Gonzalez and Dankel (1993) reported another form of circularity [13]. Consider the following rule:

Rule 15

IF X is bald

THEN X's son is bald.

If X = John and John is bald, this rule will derive that John's son is bald too. This conclusion in turn satisfies the rule's only IF condition and will derive that John's son's son is bald, and so forth.

The above descriptions of circular rules are suitable for both cases of with or without confidence factor. Circularity will lead the computer into an infinite loop, so it should be avoided.

■ Duplicated O-A Pairs. This problem was identified when the grinding chattering diagnosis rule base was being tested. Two O-A pairs are duplicated if they are identical and belong to the same rule's IF conditions. For example, consider the following rule:

Rule 16

IF engine idle-speed >1000rpm

AND engine idle-speed < 2000rpm

THEN engine idle-speed normal.

Rule 16 contains a duplicated O-A pair because "engine idle-speed" appeared twice in IF conditions. The consequence of the duplicated O-A pair is the user will be asked twice for the same question:

```
engine idle-speed [<1000rpm, >1000rpm, <2000rpm, >2000rpm]?
```

The user will be confused by the question and probably will choose an unwanted answer. For example, the user might choose the answer ">1000rpm" twice. This rule should be modified as:

Rule 16'

IF engine idle-speed 1000-2000rpm THEN engine idle-speed normal.

When the inference engine looks at rule 16', it will ask the user:

engine idle-speed [<1000rpm, 1000-2000rpm, >2000rpm]?

In this time, the user will not be confused any more.

2.2 Checking for Completeness

Incompleteness refers to the missing of rules, fact-asking and/or O-A-V triples. Incompleteness is not affected by confidence factors. Three potential problems are identified

- Unreferenced O-A-V Triples. A declared O-A-V triple is unreferenced if it is not covered by any rules. Unreferenced O-A-V triples usually imply missing rules or IF conditions. The rules or IF conditions that cover the unreferenced O-A-V triples should be added into the rule base, otherwise these O-A-V triples should be removed from the list of declared O-A-V triples.
- Unaskable IF Conditions. Unaskable IF conditions are usually called dead-end IF conditions or dead-end rules. But the terminology "unaskable IF conditions" is easier to be understood by the end user. An IF condition is unaskable if it is not referenced in any other rule's THEN conclusion and is not referenced in the ask-part of the rule base. Unaskable IF conditions might cause run time errors, so the correspondent queries should be added to the ask-part.
- Unreachable Rules. A rule is unreachable if it is not linked with the top-goal by any reasoning chain. Unreachable rules are usually caused by missing rules or IF conditions, or caused by typing errors, so they might imply the existence of logic problems in the rule base. In this case, the missing rules or IF conditions should be added to the rule base. Sometimes, an unreachable rule could be a trash rule. In this case, it should be eliminated from the rule base.

3. IMPLEMENTATION OF IDEA

IDEA consists of a rule base builder, a verification facility, a rule tree building facility, and an inference engine.

3.1 Rule Base Builder

The rule base builder of IDEA consists of a rule base editor and an ask-part generator. There are two main goals for the development of the builder. First, the builder should be a useful tool for the user to build up a rule base with minimum errors. Second, the builder should be efficient. To realize the first goal, the causes of potential problems are analyzed and appropriate approaches for eliminating these problems are adopted. The second goal is mainly guaranteed by the ask-part generator.

3.1.1 Causes of Inconsistency and Incompleteness

For an inconsistency and/or incompleteness rule base, four major causes are identified:

- 1. *Typing errors*. Typing errors are the most common errors and the major source for all kinds of problems discussed in Chapter 2.
- 2. *Misusing O-A-V triples*. Misusing O-A-V triples is the direct cause for illegal O-A-V triples and may be responsible for some inconsistency and incompleteness problems. They in turn could be the results of typing errors.
- 3. Missing rules, IF conditions and/or fact queries. Missing directly cause incompleteness in rule base. It could also cause conflicting or subsumed rules.
- 4. Logic errors. Logic errors are the cause for inconsistency.

3.1.2 Approaches

The rule base builder follows the rule of thumb that all potential problems should be detected and eliminated as early as possible. Its goal is to prevent, detect and eliminate syntax errors, illegal

O-A-V triples, unreferenced O-A-V triples and unaskable IF conditions when a rule base is in its editing phase.

In order to prevent syntax errors, our approach is to use a template for rule input. With the rule template, the rule base developer is forced to type in syntactically correct rules.

For the problems of illegal and unreferenced O-A-V triples, a special file, called *O-A-V file*, is introduced. An O-A-V file consists of O-A-V triples to be declared. An O-A-V triple has the syntax:

Object, Attribute, [Value1, Value2, ..., ValueN].

When an O-A-V triple has been entered by the rule base developer, the editor will search the O-A-V file for the same O-A-V triple. If the O-A-V triple is not found, the rule base developer will be warned. When the rule base is saved, the set of referenced O-A-V triples and the set of declared O-A-V triples will be compared. If any declared O-A-V triple is not referenced, the rule base developer will also be warned.

To eliminate unaskable IF conditions, the ask-part generator is developed. When a rule base is saved, the generator will look up all IF conditions and determine if it is a leaf IF condition. Then the generator will automatically write a fact query for each of leaf IF condition onto the rule base file. In this approach, no rule will contain unaskable IF conditions, meanwhile, no other error will exist in the ask-part of the rule base.

3.1.3 Implementation

First, a rule template is developed according to the syntax described in Chapter 1. The rule ID could be either a number or a word, but must be unique. When an object or an attribute is entered, the rule editor searches for the O-A-V database to determine if it is a new O-A pair. If it is new, the user will be given options to create a new O-A-V triple or to re-enter the O-A pair. When a value is entered, if it belongs to an existing O-A-V triple, it will be checked if it is a member of the value list; otherwise, it will not be checked. In the first case, if the value is new, the user is again given options to create a new value or to re-enter it. When the user is prompted to enter an object, all of the pre-declared objects are listed on the screen. The user can type in only the index number instead of the whole object. This greatly saves typing work and reduces typing errors. Similarly, when the user is prompted to enter an attribute, all of pre-declared

attributes that followed the object that was typed in are listed. When the user is prompted to enter a value, the set of pre-declared values that belongs to the O-A pair are listed.

Second, more editing functions are provided. If the user types in a rule ID that already exists, he/she is given options to modify, delete or rename the rule, or to re-enter the rule ID.

Finally, when the user chooses to save the rule base file, the ask-part is automatically generated. Meanwhile, illegal O-A-V triples in the rule base and unreferenced O-A-V triples in the O-A-V file, if exisit, are checked. If there are unreferenced O-A-V triples, the user will be informed whether it is the whole triple that is unreferenced or only some values that are unreferenced.

3.2 Rule Base Verification Facility

The need for a rule base verification facility for IDEA is due to the fact that the rule base builder is not capable of detecting inconsistency problems, though it can greatly reduce potential problems. The rule base verification facility was developed to detect inconsistency and unreachable rules.

Several verification tools have been reported since 1980's. One example is CHECK, developed by T. Nguyen *et al*, 1987, for use with *LES*, a generic rule-based expert system building tool [1]. CHECK is reported to be capable of detecting all problems described by the developers (see Chapter 2).

PREPARE is another verification tool, implemented by Zhang and D. Nguyen, 1993 [14]. PREPARE used a special modeling methodology, called *Predicate/Transition (Pr/T) nets*, a subclass of *Petri nets*. Pr/T nets are also a graphical representation of first order predicate logic. Pr/T nets are used to model the rule base so that it may be verified by detecting the problem patterns associated with inconsistency and incompleteness.

3.2.1 Approaches

To detect redundant, conflicting, subsumed rules and unnecessary IF conditions, the approach used in IDEA is similar to the one used in CHECK. For each rule in the rule base, the verification facility compares it with the other rules one by one. If a rule is found to have the same LHS, the facility will check if a redundant or conflicting rule exists. If a rule is found to

have the same RHS, the facility will check if a subsumed rule or an unnecessary IF condition exists.

For the problems of circular and unreachable rules, the approach is unique. This approach is based on the fact that the rule bases referenced by IDEA are goal-driven rule bases. Starting from the top-goal, the verification facility traces over the whole rule base and puts all of the rules it encountered into a database. This database has a tree structure. The root node is the top-goal. The leaf nodes are the rules that contain leaf IF conditions. Each node (rule) of the tree in turn has a *pointer/4* structure:

pointer(Level, RuleID, Children, History).

Where *Children* is a list containing the IDs of the rules the node points to; *History* is also a list containing the IDs of the rules from the top-goal to the previous node. If a loop exists, it can be detected by checking the *History*. The rules not covered by the database of *pointer/4* are unreachable rules.

3.2.2 Implementation

The actual work is a program that is capable of detecting the major problems described in Chapter 2. These problems cover redundant rules, conflicting rules, subsumed rules, unnecessary IF conditions, circular rules, and unreachable rules. This verification facility does not deal with confidence factors, because the original scope of this project was merely to realize the basic functions of CHECK.

For redundant rules, only the case illustrated by rule 1 and rule 2 in Chapter 2 is covered. IDEA does not detect the rules that can be derived from the other two or more rules. Detecting these rules might require considerable effort and consume considerable amount of memory space and computing time. Considering that redundant rules do not cause logic errors, it is usually safe without implementation of this function at this point.

For unnecessary IF conditions, IDEA covers only the case illustrated by rule 7 and rule 8. The case illustrated by rule 9 and rule 10 will be detected as subsumed rules, because the program ignores the "not" in the second IF condition of rule 9. IDEA does not detect the case illustrated by rule 11 and rule 12, because of the lack of time.

IDEA does not cover duplicated O-A pairs, because this problem was not identified until the grinding chattering diagnosis rule base was tested. They should be detected by the rule base builder.

For circular rules, the case illustrated by rule 15 dose not exists in the rule bases referenced by IDEA, because these rule bases contain no variable, as mentioned in Section 1.4.

As to illegal O-A-V triples, unreferenced O-A-V triples and unaskable IF conditions, they are detected by the rule base builder.

3.3 Rule Tree

The capability of drawing a rule tree is a notable feature of IDEA. This feature provides the user with

- a tool for the knowledge engineer to structurally build up a rule base;
- a tool for the knowledge engineer to structurally test the rule base; and
- a tool for the end user to trace the rule base.

As a developing tool, the rule tree gives the knowledge engineer a graphical view of the structure of a rule base. With the rule tree, the engineer knows the depth of the rule base, the number of branches contained in the rule base, the logical interrelationships among the rules and even for the rule loops. As a testing tool, the rule tree tells all the logical paths in the rule base. Using these paths, the knowledge engineer can systematically test the rule base without any unnecessary repetitive work. Moreover, the testing work can be independent of the inference engine. As a tracing tool, the rule tree, together with the rule base, tells the end user how the final conclusions are reasoned and why a suspected conclusion can not be reached.

In the subsections below, the alternative graphical representations of rule bases are compared, the reason why rule tree is chosen is given, and the approach of drawing a rule tree is discussed.

3.3.1 Graphical Representation of Rule Base

T. Nguyen *et al* reported that CHECK used a *dependency chart* to show the interactions among the goal-driven rules and the goal [1]. An example of dependency chart based on Appendix A is shown in Figure 3.1 where an * indicates one IF condition of the left rule is matched by the THEN conclusion of the top rule. The dependency chart has the advantage of simplicity. It is relatively easy to generate and draw onto the screen by a computer. But it has some disadvantages. First, it does not give a graphical picture of logical paths. Second, it does not distinguish the "AND" relation from the "OR" relation. Finally, it does not show how many IF conditions are contained in a rule.

	Rule 1	Rule 2	Rule 3	Rule 4	Rule 5	Rule 6	Rule 7
Rule 1		*	*	*			
Rule 2							
Rule 3							
Rule 4							
Rule 5							
Rule 6							
Rule 7							
Goal	*				*	*	*

Figure 3.1 A sample of dependency chart

Rule graph (logical path graph) is a more natural alternative [5]. An example of a rule graph based on the same rule base is shown in Figure 3.2 where an arc connecting two or more paths indicates an "AND" relation, and a cross means a negation. A rule graph overcomes the drawbacks of a dependency chart at the expense of being more complex. It requires quite complex programming to draw the rule graph onto the screen.

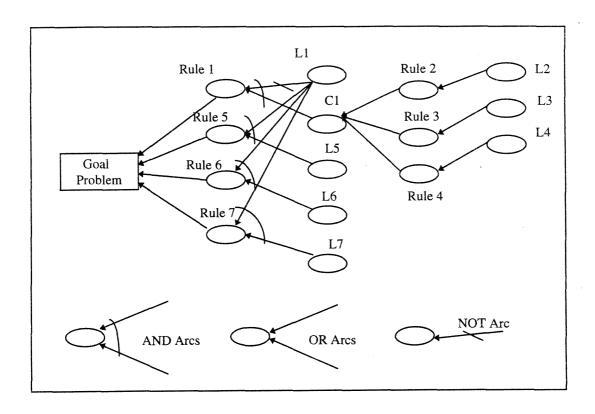


Figure 3.2 A sample of rule graph

L1: engine turn-over; L2: auto high-beam weak;

L3: auto low-beam weak; L4: auto radio weak;

L5: smell gas; L6: gas gauge empty;

L7; gas gauge low; C1: auto battery bad.

IDEA uses a *rule tree* to graphically represent a rule base. A rule tree is similar to a rule graph and has almost the same advantages. The difference is that in a rule tree a node (rule) is duplicated whenever it appears in a new path. It is easier to draw a rule tree than a rule graph. This is the major reason why we chose a rule tree to graphically represent a rule base. Figure 3.3 shows a rule tree based on the same rule base as Figure 3.2.

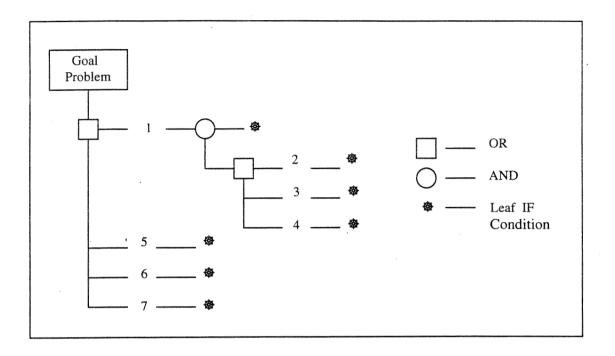


Figure 3.3 A sample of rule tree

3.3.2 Approach and Implementation

A rule tree is a kind of AND/OR tree. On a rule tree, the root node is the top-goal. On the next level, the nodes are *sub-goals* connected by only one "OR" logical connector. A sub-goal is a special rule that has the THEN conclusion as the top-goal. In the following levels, an "OR" connector always follows an "AND" connector if they exist.

The algorithm of drawing a rule tree is based on the database of *pointer/4*, discussed in Section 3.2. Starting from the root node, the rule tree is traced branch by branch. For each node (rule) encountered, its specific graphical feature is identified, and the feature is added into the database:

level(..., AndGate, AndGateFeature, OrGate, OrGateFeature).

The possible values of these attributes of *level* are shown in Table 3.1. For examples, the graphical feature of node rule 2 in the path "problem - 1 - 2- *" is represented as:

level(..., and, end, or, new).

That means the node rule 2 is led by an AND connector and an OR connector, and it is at the "end" branch of the AND connector and at the first branch of the OR connector. Similarly, the graphical feature of node rule 3 in the path "problem - 1 - 3 - *" is represented as:

level(..., and, written, or, branching).

That means the node rule 3 is led by an AND connector and an OR connector, and it is at a subbranch of the AND connector and at an intermediate branch of the OR connector. Because the database of pointer/4 is the by-product of verification, rule tree drawing must be preceded by a verification operation.

Table 3.1 Attributes and Values of Database level.

AndGate	AndGateFeature	OrGate	OrGateFeature
and	new	or	new
	branching		branching
	end		end
	written		written
	•		
not and	new	not or	new
	written		written
		شد حدد الله جان جان هند هن رفاد احد حدد خود شن باله العد باله المد سن حدد مند	

3.4 Integration

After the facilities of editing, verification and rule tree are developed and tested individually, they, together with the inference engine, are integrated into a single environment. This

environment is given the name IDEA, as mentioned above. IDEA has the user interface that has a look of a general purpose computer language. In more detail, the interface has the options such as file, edit, verify (similar to "compile"), run, trace, help, and so on.

3.4.1 Inference Engine

The inference engine is based on the work implemented in the expert systems class, 1993, lectured by Dr. Ozden. This work in turn is based on the textbook. Some modifications were done and the interface was re-designed.

The inference algorithm used is goal-driven, backward-chaining. Starting from the top-goal, the inference engine first checks if the top-goal is satisfied by an existing fact. If so, the conclusion is immediately reached. Otherwise, the inference engine searches for the sub-goals. The searching work is carried out by matching the top-goal with the THEN conclusions as the sub-goals. When a rule fires (THEN conclusion matches), it will check for the IF conditions contained by the sub-goal found. For each of the IF conditions, the inference engine will regard it as a goal and do the above work over again. If an IF condition is neither satisfied by any existing fact nor matched by any rule's THEN conclusion, the inference engine will ask the user to provide a basic fact that supports or rejects the IF condition. In both cases, the fact is added into the working memory. If all the IF conditions contained in a rule are satisfied, the THEN conclusion is established. Otherwise, the rule is rejected.

The inference engine uses confidence factors of MYCIN to deal with uncertainty. If a rule's IF conditions are supported by facts with confidence factors less than 100, the original THEN conclusion confidence factor is replaced by the product of the minimum IF condition confidence factor and the original THEN conclusion confidence factor itself. For example, if we have the rule:

Rule 17

IF auto radio weak

AND auto lights weak

THEN auto battery bad cf 90.

and the IF condition are supported by

auto radio weak cf 80; auto lights weak cf 60.

The new conclusion should be

auto battery bad cf 54.

The additional functions of the inference engine are:

- *Tracing*. This function is used to trace the reasoning process. The user will be told which rule is called and if it is fired or rejected.
- How Explanation. This function tells the user how the final conclusion is reached.
- Why Explanation. When the user is prompted by a question, he/she can choose this
 option to get the explanation for why this question is proposed.
- Why not Explanation. This function tells the user why an expected conclusion is not reached.

3.4.2 Structure of Interface

IDEA has three modes: the *main mode*, the *editing mode*, and the *run mode*. Each mode contains several push buttons. Each push button in turn may have several options. The hierarchical structure of IDEA interface is shown in Figure 3.4.

The interface of IDEA is push-button-driven, somewhat different from ordinary menudriven interface. This interface has the following features:

- 1. Quick interface. Compared with a menu-driven interfaces, it may save one key stroke to reach the same option.
- 2. Using modes. When the user chooses the edit option or the run option, it is likely that he/she will stay in this option for a relatively long time. With the editing mode and the run mode, the user can choose the related options more directly.

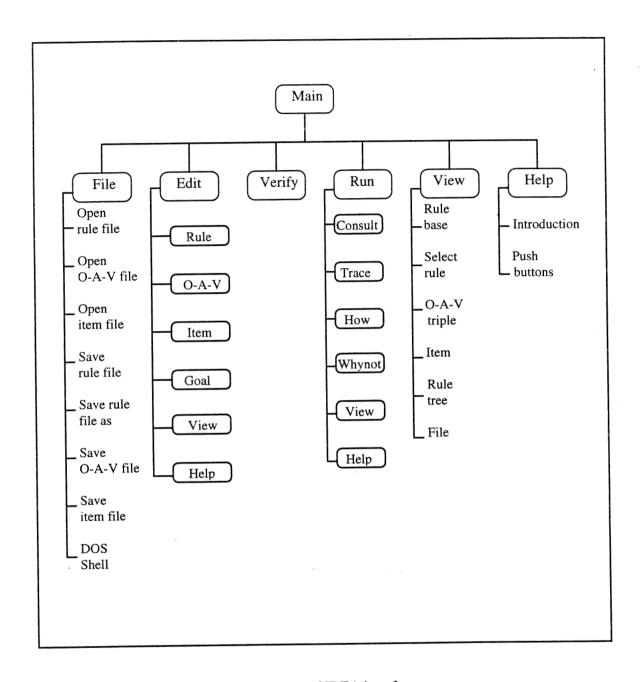


Figure 3.4 The hierarchical structure of IDEA interface

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4. APPLICATION TO GRINDING PROCESS

This application is another major work accomplished in this project. The motivation is to build up a prototype rule base that could be used to solve a real world problem using IDEA. As mentioned in Chapter 1, the domain of this application is quite narrow so that it covers only the chattering problems in external cylindrical grinding.

4.1 Grinding

Grinding is one of the major metal cutting process. The stocks on the surfaces of a work piece are removed by a rotating grind wheel which is composed of tiny and hard bonded abrasive grains. Each grain acts as a cutter. Grinding is usually used when the tolerance or finishing requirements of the part to be machined are very high or the material is too hard to cut by other metal cutting methods.

There are several reasons for building a grinding expert system:

- 1. Lack of experts. The mechanism of grinding is much more complex than other ordinary metal cutting methods, such as turning, shaping, milling, and so forth. When turning, shaping or milling a work piece, only one or a few well-positioned and shaped cutters cut materials at the same time. Whereas, in grinding process, there are a large number of abrasive grains participating in the cutting process simultaneously. The shape, position and size of each grain varies greatly. The consequence of the complexity is the difficulty of finding a grinding expert.
- 2. Cost. Grinding is an expensive and energy consuming process. The introducing of an expert system is likely to have a significant payoff.
- 3. *Importance*. Grinding is typically used to process important parts with high quality requirements, so it is worth having an expert system to provide process engineers with a quality management tool.

There are many variations for the grinding process. The basic types of grinding are surface grinding, external cylindrical grinding and internal cylindrical grinding, shown in Figure

4.1. External cylindrical grinding, for which the rule base is developed, is the one most commonly used.

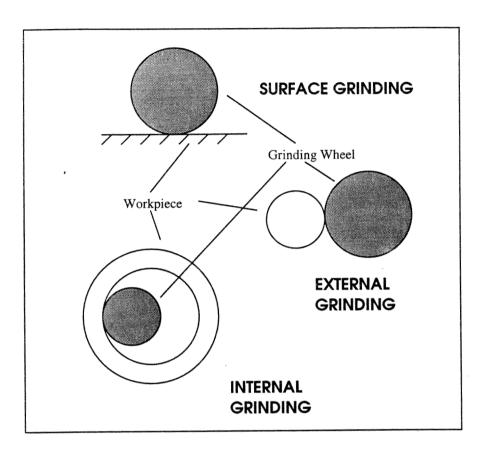


Figure 4.1 The types of grinding

4.2 Chattering

Chattering means vibrations occurred in a metal cutting process. For grinding processes, the influences of chattering are significant:

- 1. It will generate chatter marks which greatly degrade the surface quality of the part ground;
- 2. It might cause premature part failure as the result of fatigue;
- 3. It will create audible noise during the grinding process; and
- 4. It might disrupt lubrication.

There are tree types of vibrations that cause chatter marks: *random vibration*, *forced vibration* and *self-excited vibration* [22]. In the rule base, the chattering caused by self-excited vibration is called regenerative chatter.

Random vibration is caused by an occasionally created vibration source, for example, a passing forklift truck or a passing train. If chatter marks are created by a random vibration, their frequency could be irregular, and their amplitude could be variable. Random vibration does not often cause problems, but its source is difficult to identify.

Forced vibration is caused by a fixed vibration source which could be either outside or inside the grinder. The examples for an outside vibration source could be a punch press or an air hammer. The inside vibration source is usually an unbalanced rotating component, such as an unbalanced grinding wheel or an unbalanced idle wheel. If chatter marks are created by forced vibration, their frequency will be synchronous with the vibration source, and their amplitude will be constant.

The mechanism of self-excited vibration is quite complex. There are several explanations for self-excited vibration, but none of them is commonly recognized. Generally speaking, a self-excited vibration is agitated by a random event. For example, if the cutter encounters a hard point in the material to be removed, it will cause the initial vibration. When a self-excited vibration is stirred up, it can maintain itself by getting required energy from the cutting process. Self-excited vibrations only occur when a cutting processing is going on. The amplitude of chatter marks will become larger at the beginning of the cutting process until a balance between the energy consumed and obtained is reached.

4.3 Rule Base

The actual work for the application to grinding process is a grinding chattering diagnosis rule base that contains about forty rules (see Appendix B).

4.3.1 Implementation

The knowledge needed for building the rule base was mainly acquired from a grinding trouble-shooting table, written by Lewis and Schleicher, 1976 [25]. Other resources include expertise provided by Associate Professor M. Schmenk, materials provided by a grinding seminar,

sponsored by Institute of Advanced Manufacturing Science, Inc., 1993 [8]; and a chapter from a text book, written by Yueyi Yu, 1991 [22]. The knowledge acquired was then translated into IF-THEN rules

To write a well structured rule base, the most important thing is to group the rules properly. With well organized rules, the end user can reach a conclusion with a minimum number of questions answered, and the knowledge engineer will benefit when maintaining, testing, and possibly expanding the rule base. Because the original trouble-shooting table is quite flat, the chattering indications and causes were grouped according to the types of vibrations before they were translated into rules. Some modifications for the original table were also done.

Ordering is another consideration for the rule base development. For some inference engines, the order of rules will have an influence on efficiency. As a general principle, the rules dealing with the most frequent events should be placed first. Although the inference engine used for IDEA is not rule order sensitive, because it will look at all of possible conclusions, the rules in the chattering diagnosis rule base were still ordered according to the principle. In more detail, the rules about self-excited vibration were placed first, followed by the rules about forced vibration and random vibration.

4.3.2 Result

After the rule base was built up, it was verified by the verification facility of IDEA, discussed in Section 3.2. Through the verification, an inconsistency problem, duplicated O-A pairs, which was not reported before, was identified.

The rule base was then used to test the other facilities of IDEA. Through the testing work, some problems with IDEA were also found and fixed.

Finally, the rule base was checked by Associate Professor M. Schmenk. According to Associate Professor M. Schmenk, the rule base is moderately helpful for external cylindrical grinding chattering diagnosis. This rule base has the following positive features:

- 1. it provides the end user with quite complete answers for the causes of chattering for external cylindrical grinding;
- 2. the terms used in this rule base can be accepted by grinding process engineers; and
- 3. it is well structured, so it can be easily expanded to cover a wider range for grinding diagnosis.

The limitations of this application are mainly due to the limited functions of the inference engine. Because the development of an inference engine was beyond the original scope of this project, some useful functions are not developed. Associate Professor M. Schmenk especially pointed out the lack of the following functions:

- 1. Graphical View of Grinder. A picture of grinder will be very helpful for a less experienced grinding process engineer to locate the source of chattering.
- 2. *Undo option*. The end user is likely to type in some wrong answers. An undo option would be a great convenience for the end user to correct the errors.
- 3. Advises. After the source of chattering is identified, the program should provide the end user with advice or instructions to fix the problem.

5. CONCLUSION AND FUTURE WORK

The major work accomplished in this project is the expert system development environment IDEA and its testing in a grinding chattering diagnosis rule base. IDEA is a PC-based program that has the following features:

- 1. *Integrity*. IDEA is a highly integrated expert system environment. It includes a rule base builder, a rule base verification facility, and an inference engine, which are incorporated in a single user interface.
- 2. Error avoiding and detecting. IDEA provides user with facilities to avoid syntax errors, checks for typing mistakes, illegal O-A-V triples and other inconsistency and incompleteness problems.
- 3. Ask-part auto-generation. After rules are written, the ask-part of the rule base is automatically generated.
- 4. Rule tree. IDEA uses rule tree to graphically represent the rule base. This tree is a multi-functioned tool that can be utilized by knowledge engineer, end user, and possibly domain expert.
- 5. *Prolog-based.* IDEA is a program built in Prolog. The benefits of using Prolog are the efficiency in coding and the relative small size of program.

Because of the limitation of time, IDEA is still in its prototype stage. Many further improvements can be easily identified. Examples are:

- 1. The duplicated O-A pairs should be detected.
- 2. New functions should be added to O-A-V file editor. This editor should allow the user to modify or to delete an O-A-V triple.
- 3. The inference engine should allow the user to undo an input or to restart from a specified point.
- 4. The verification facility should have the ability to deal with confidence factors.

In the long run, the author would like to point out some more interesting future research topics:

- 1. Knowledge representation issue. The trends of knowledge representation are arguably featured by the popularity of hybrid formalism and object-oriented formalism. For example, KEDE, a knowledge-based system building environment, reported by Zheng and Li, 1993, integrated five different formalisms. It would be a promising research topic to develop a knowledge base editor, a verification tool or an inference engine that uses different types of knowledge representations.
- 2. Tree-based algorithm. As mentioned in section 1, one of the disadvantages for production rules is inefficiency. This is because the knowledge in the rule base is not well organized. But after the rule base is verified by the verification facility of IDEA, the rule IDs are added into the rule tree, and the knowledge becomes more organized. It is possible to use the rule tree, the by-product of verification, to develop a new algorithm for the inference engine. The new algorithm will store and retrieve a rule in a tree structure, so it may considerably reduce the goal matching work discussed in Section 3.4.

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APPENDIX A -- A SAMPLE OF RULE BASE

problem. goal

if not engine turn_over rule 1

and auto battery bad

then problem auto battery.

if auto high_beam weak rule 2

then auto battery bad cf 90.

if auto low_beam weak rule 3

then auto battery bad of 80.

if auto radio weak rule 4

then auto battery bad cf 80.

if engine turn_over rule 5

and smell gas

then problem gas flooded cf 80.

if engine turn_over rule 6

and gas gauge empty

then problem gas runout cf 99.

if engine turn_over rule 7

and gas gauge low

then problem gas runout cf 80.

engine turn_over ask

menu_ yes no

prompt_ 'engine turn_over ?'.

auto high_beam ask

menu_ weak strong

prompt_ 'auto high_beam ?'.

auto low_beam ask

menu_ weak strong

prompt_ 'auto low_beam ?'.

ask auto radio

menu_ weak strong

prompt_ 'auto radio ?'.

ask smell gas

menu_ yes no

prompt_ 'smell gas ?'.

ask gas gauge

menu_ empty low full

prompt_ 'gas gauge ?'.

eof.

APPENDIX B - GRINDING CHATTERING DIAGNOSIS RULE BASE

goal problem.

rule 1 if chatter occurs

and chatter_mark frequency random

then random vibration cf 90.

rule 2 if chatter occurs

and chatter_mark frequency regular and not vibration_exist without_grinding and chatter_mark amplitude.going_larger

then regenerative chatter cf 90.

rule 3 if chatter occurs

and chatter_mark frequency regular and not vibration_exist without_grinding

and not chatter_occurence dependent_on_spindle_speed

then regenerative chatter cf 90.

rule 4 if chatter occurs

and chatter_mark frequency regular

and chatter_occurence dependent_on_spindle_speed

then forced vibration cf 90.

rule 5 if chatter occurs

and chatter_mark frequency regular and vibration_exist without_grinding

then forced vibration of 90.

rule 6 if regenerative chatter

and workpiece stiffness weak

then problem workpiece weak.

rule 7 if regenerative chatter

and wheel dressing improperly

then problem wheel dressing.

rule 8 if dressing diamond dull

then wheel dressing improperly cf 90.

rule 9 if diamond holder loose

then wheel dressing improperly cf 90.

rule 10 if wheel_dressing interval too_long

then wheel dressing improperly cf 90.

rule 11 if regenerative chatter

and chatter_mark length long and chatter_mark space wide

and workpiece discolored then problem wheel too_hard cf 90. if regenerative chatter rule 12 and chatter_mark length long and chatter_mark space wide and wheel glazed_or_loaded then problem wheel too_hard cf 90. if random vibration rule 13 and thrust bearings work bad then problem thrust bearings. rule 14 if thrust bearings out_of_round then thrust bearings work_bad. if thrust bearings worn_out rule 15 then thrust bearings work_bad. if thrust bearings loose rule 16 then thrust bearings work_bad. if random vibration rule 17 and work center_or_work_rest work_bad then problem work_center_or_work_rest. if not work_center_or_work_rest aligned rule 18 then work_center_or_work_rest work_bad cf 90. if work_center_or_work_rest lubrication inadequate rule 19 then work center_or_work_rest work_bad cf 80. if random vibration rule 20 and not thrust bearings work_bad and not work_center_or_work_rest work_bad then problem random_vibration outside_of_machine cf 60. if forced vibration rule 21 and thrust bearings work_bad then problem thrust bearings. rule 22 if forced vibration and spindle belt with_metal_lacing then problem spindle belt of 90. if forced vibration rule 23 and chatter_mark length short and chatter_mark space close then problem wheel_spindle bearings cf 80.

and chatter_mark length medium

if forced vibration

rule 24

and chatter_mark space medium then problem wheel_spindle sprung_or_out_of_round cf 80.

rule 25 if forced vibration

> and chatter_mark length medium and chatter mark space wide

then problem wheel_spindle sprung_or_out_of_round cf 80.

rule 26 if forced vibration

and chatter_mark synchronous_with motor

then problem motor imbalance cf 90.

if forced vibration rule 27

and chatter_mark synchronous_with electricity

then problem electricity imbalance of 90.

rule 28 if forced vibration

and chatter_mark synchronous_with spindle

and workpiece irregular

then problem workpiece imbalance of 90.

rule 29 if forced vibration

> and chatter_mark length short and chatter_mark space wide

and flat_belt uneven

then problem flat_belt uneven cf 90.

if forced vibration rule 30

> and chatter_mark length medium and chatter_mark space wide

and flat belt uneven

then problem flat_belt uneven cf 90.

rule 31 if flat belt thickness ununiform

then flat_belt uneven.

if flat_belt width ununiform rule 32

then flat_belt uneven.

rule 33 if forced vibration

> and chatter_mark length short and chatter_mark space wide

and not flat_belt uneven

then problem idlers loose_or_out_of_balance cf 80.

rule 34 if forced vibration

> and chatter_mark length medium and chatter_mark space wide

and not flat_belt uneven

then problem idlers loose_or_out_of_balance cf 80.

rule 35 if forced vibration and chatter_mark length long and chatter_mark pattern checkerboard then problem wheel out_of_balance_or_round cf 80.

rule 36 if forced vibration

and chatter_mark synchronous_with wheel then problem wheel out_of_balance_or_round cf 90.

rule 37 if forced vibration

and chatter_mark length long and chatter_mark space wide then problem drive_gear back_lash cf 80.

ask chatter occurs

menu_ yes no

prompt_ 'chatter occurs ?'.

ask chatter_mark frequency

menu_ random regular

prompt_ 'chatter_mark frequency ?'.

ask vibration_exist without_grinding

menu_ yes no

prompt_ 'vibration_exist without_grinding ?'.

ask chatter_mark amplitude

menu_ going_larger other

prompt_ 'chatter_mark amplitude ?'.

ask chatter_occurence dependent_on_spindle_speed

menu_ yes no

prompt_ 'chatter_occurence dependent_on_spindle_speed ?'.

ask workpiece stiffness

menu_ weak other

prompt_ 'workpiece stiffness?'.

ask dressing diamond

menu_ dull other

prompt_ 'dressing diamond?'.

ask diamond holder

menu_ loose other

prompt_ 'diamond holder ?'.

ask wheel_dressing interval

menu too long other

prompt_ 'wheel_dressing interval?'.

ask workpiece discolored

menu_ yes no

prompt_ 'workpiece discolored ?'.

ask wheel glazed_or_loaded

menu_ yes no

prompt_ 'wheel glazed_or_loaded ?'.

ask thrust bearings

menu_ out_of_round worn_out loose other

prompt_ 'thrust bearings?'.

ask work_center_or_work_rest aligned

menu_ yes no

prompt_ 'work_center_or_work_rest aligned ?'.

ask work_center_or_work_rest lubrication

menu_ inadequate other

prompt_ 'work_center_or_work_rest lubrication ?'.

ask spindle belt

menu_ with_metal_lacing other

prompt_ 'spindle belt ?'.

ask chatter_mark length

menu_ long short medium

prompt_ 'chatter_mark length?'.

ask chatter_mark space

menu_ wide close medium

prompt_ 'chatter_mark space ?'.

ask workpiece irregular

menu_ yes no

prompt_ 'workpiece irregular ?'.

ask flat_belt thickness

menu_ ununiform other

prompt_ 'flat_belt thickness ?'.

ask flat_belt width

menu_ ununiform other prompt_ 'flat_belt width ?'.

ask chatter_mark pattern

menu_ checkerboard other

prompt_ 'chatter_mark pattern ?'.

ask chatter_mark synchronous_with

menu_ motor electricity spindle wheel other

prompt_ 'chatter_mark synchronous_with ?'.

eof.