

Real-Time Knowledge-Based Integration of Freeway Surveillance Data

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An advanced processing capability based on the use of real-time knowledge-based expert system (KBES) technology to integrate diverse types of traffic surveillance data for freeway monitoring and control purposes, particularly as part of future "smart roads" projects, is described. One of the major functions of the prototype system is the acquisition and processing of input data drawn from sensors and processes in the real world. The real-time nature of these processes, and the associated need for decision-making information and recommendations, places particular importance on the efficient handling of data to avoid unnecessary overloading of the expert system. The relevant types of data include traffic occupancies and volumes from loop detectors in the pavement, information on traffic conditions from closed-circuit television cameras, field reports from police officers and other official personnel, and cellular and emergency telephone calls from motorists. Emphasis is placed on the way in which the data are acquired, processed, and integrated within a prototype KBES framework to achieve the objectives of the incident management tasks, specifically those of incident detection and verification. Examples are given of the implementation of these features.

As part of an ongoing research effort to investigate application of artificial intelligence techniques for advanced traffic management, a prototype real-time knowledge-based expert system (KBES) has been developed for managing nonrecurring congestion on urban freeways. The objective of this system, called "freeway real-time expert system demonstration", (FRED) is to provide decision support in a future traffic operations center (TOC) to traffic control room operators responsible for the monitoring and control of so-called "smart roads" or intelligent vehicle-highway systems (1). The FRED system is described in a companion paper by Ritchie and Prosser in this Record. The only comparable expert system currently known is one being developed by INRETS, the French transportation agency, for monitoring and control of arterial street intersections (2). The G2 expert system shell used in the INRETS project is the same as the one used in the FRED prototype system.

More specifically, FRED provides decision support to TOC operators in the field of incident management. Incidents are events that lead to nonrecurrent congestion. Examples of such events are accidents, spilled loads, stalled or disabled vehicles, temporary maintenance and construction activities, signal and detector malfunctions, and special events. Incident management involves detecting and responding to incidents in order to alleviate the resultant delay to motorists. In future smart road TOCs, it will become increasingly difficult, if not impossible, for human operators to function effectively without

automated assistance because of the increased amount of incoming TOC data and the complexity both of the networks and of incident management and response functions.

The monitoring aspect of FRED involves synthesizing input data from a number of sources. The freeway system is monitored, in the first instance, to detect when and where an incident has occurred. Once an incident has been detected, more information is required about the state of the system to formulate appropriate responses. One of the major functions of the system is the acquisition and processing of data drawn from sensors and processes in the real world. The real-time nature of these processes, and the associated need for decision-making information and recommendations, place particular importance on the efficient handling of data to avoid unnecessary overloading of the expert system.

Focusing on the initial incident detection and verification phases, which represent the foundation of incident management, a description is provided of an advanced processing capability based on the use of real-time KBES technology to integrate diverse types of traffic surveillance data for freeway monitoring and control purposes. The relevant types of data include traffic occupancies and volumes from loop detectors in the pavement, information on traffic conditions from closed-circuit television (CCTV) cameras, field reports from police officers and other official personnel, and in-vehicle cellular and roadside call-box telephone calls from motorists. Emphasis is placed on the way in which the data are acquired, processed, and integrated to achieve the objectives of the incident management tasks, specifically those of incident detection and verification. Examples are given of the implementation of these features in FRED.

SYSTEM OVERVIEW

Figure 1 shows the different stages of incident management (3). In the companion paper in this Record, the development of FRED—a component prototype real-time expert system for managing nonrecurring congestion on urban freeways—is discussed, and the application of FRED to a section of the Riverside Freeway in southern California is presented as a case study to illustrate the current capabilities of the system. As mentioned, the focus here is on the way in which data are acquired and processed to achieve the objectives of the incident management tasks. The incident detection stage relies heavily on data acquisition and processing and incorporates most of the features of interest in the following sections.

Figure 2 shows the overall FRED system layout, which in terms of incident detection and verification is consistent with

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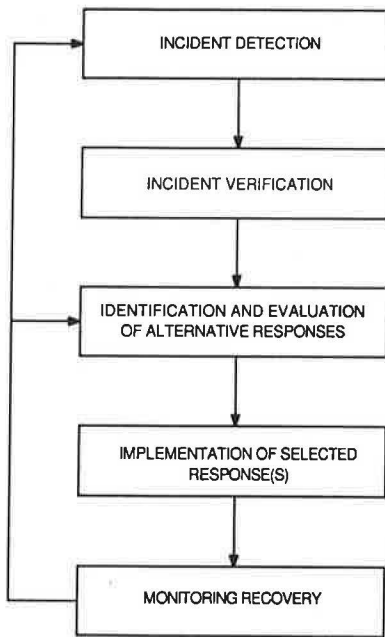


FIGURE 1 Incident management tasks.

the current capabilities of the Los Angeles freeway TOC (operated by the California Department of Transportation) and also with many other TOCs around the country. Of particular note are the external sources of data in the detection phase. Two streams of data, representing the detection and reporting of an incident, respectively, can be recognized. First, freeway loop detectors transfer 30-sec occupancy counts to an incident detection algorithm on a central computer, which in turn notifies FRED of the detection of an incident. Second, a number of agencies and sources may provide on-site accounts of an incident, and a communications center receives and filters their reports before sending high-priority ones to FRED. Thus, the two primary methods of incident detection are (a) an

automated incident detection algorithm, and (b) outside reports.

The incident detection algorithm (4) operates on 30-sec occupancy counts (the percentage of time a loop is occupied by vehicles) obtained from sensors placed on the freeway main line at approximately 1-mi spacing. Each sensor site has a series of loop detectors, one per lane, arranged to form a counting station. Local controllers are responsible for compiling the counts for the detectors in each lane and then averaging them over the counting station. In the FRED prototype system, simulated occupancy counts are read from a data file. The incident detection algorithm examines the average occupancy counts for each station and, by comparing them to those for the station immediately downstream, can detect the presence of a significant disruption to traffic flow between the two stations. The sensitivity of the algorithm can be controlled by varying threshold parameters. Once an incident has been detected by the algorithm, a message is sent to FRED indicating the location of the counting station immediately upstream of the detected disruption.

The FRED system has been developed using a real-time expert system shell known as G2 (5), running under the Unix operating system and X-windows on a Sun SPARCstation 1 workstation. The so-called "California" incident detection algorithms (4) and the report processing algorithm were written as separate C programs capable of running on a different processor. Communication between the FRED expert system and the external C programs is via data files.

In the freeway control environment, an incident management system must operate in real time if it is to detect, verify, and formulate responses to incidents in a sufficiently short period of time for responses to be implemented effectively. For example, one type of response strategy is the provision of advance information to motorists approaching the site of an incident via changeable message signs or in-vehicle navigation systems. Such information, to be of benefit, must be provided soon after the occurrence of an incident. In this context, "real time" probably implies a maximum interval of

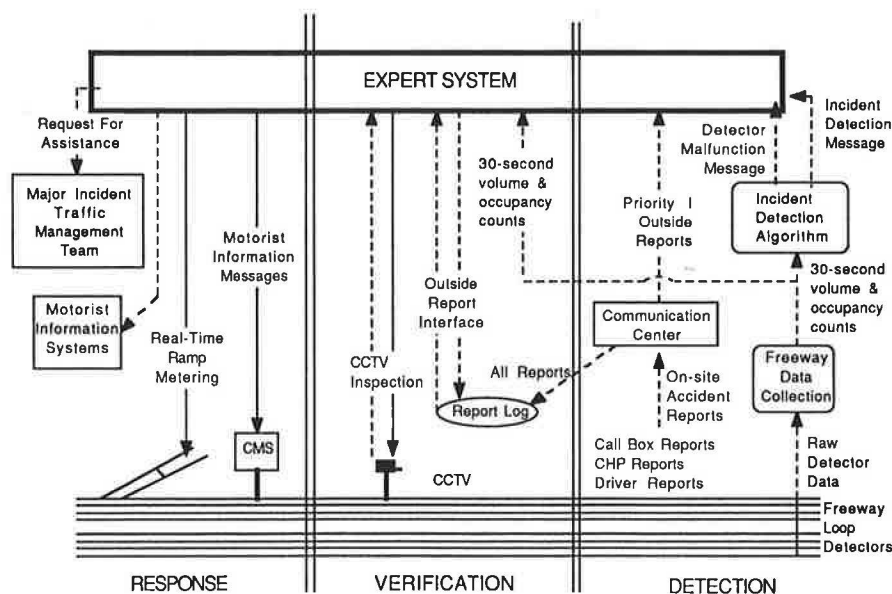


FIGURE 2 Overview of FRED external system.

several minutes between incident occurrence and response implementation. The real-time operating constraint introduces some important issues related to concurrent processes and data transfer, which are discussed in the next section.

DISTRIBUTION OF PROCESSING AND DATA TRANSFER

The workload on the TOC at any one time is directly proportional to the number of current incidents. Effective allocation of the expert system's resources is essential to maintain the speed of the overall system. The expert system should be devoted to those tasks for which it is best suited: the processing of high-level knowledge rather than simple algorithmic tasks. For FRED, the major high-level task is the formulation of responses to incidents. The detection of further incidents should proceed concurrent and external to the expert system. When another incident is detected, the expert system can be interrupted to include the new incident in its set of current incidents. Currently, FRED only handles one incident at a time, but the system structure was developed to facilitate the future treatment of multiple incidents.

Given that different tasks are allocated to separate and concurrent processes, the transfer of data between the processes becomes a major issue. An advantage of a knowledge-based approach is that the expert system has the ability to decide what data it requires at any particular stage and is able to send out requests for it. In this way, the system can select the information it needs for a specific task from the vast amount of data available. The decision of whether to transfer data from an external routine to the expert system can also be made by the external routine itself. In this case, the expert system is waiting to be interrupted by the occurrence of an event.

These aspects of the operation of a real-time expert system are discussed in terms of the implementation of FRED. Before doing so, the manner in which data and knowledge are represented in the system is outlined.

DATA AND KNOWLEDGE REPRESENTATION IN FRED

As explained, the FRED system was developed using a real-time expert system shell called G2 (5), which combines a knowledge base with sophisticated data interfacing and screen management capabilities. G2, like most sophisticated expert systems, is hybrid in nature, in that knowledge and data are present separately. The manner in which the knowledge is applied to the data determines the behavior of the overall system.

Data Representation

All data in FRED are represented as a series of objects. Each object contains a number of predetermined attributes that represent the data relevant to that object. Objects are all instances of a particular class, and the class definition specifies the attributes of all objects within the class.

The attributes of objects can receive values from a number of sources. The data can be inferred from rules by the inference engine, determined by an external routine, or simulated by the G2 simulator incorporated into the G2 system. Additionally, the system operator can enter values for data interactively.

The two most important object classes in the incident detection and verification phases are the counting stations and incident classes. Tables 1 and 2 present the attributes of each of these object classes along with their type and source. Variable attributes are either symbolic or quantitative. The counting station attributes either are constant or, for volumes and occupancies, receive their values from an external routine. Incident attributes either are derived from rules and formulas as part of the inference process in the expert system or are explicitly set by the operator or an outside report. Other attributes within these objects, of less importance, are not shown.

An important feature of G2 is the ability to create and delete objects during the running of the system. Such transient objects are particularly suitable for incident objects. When an incident is detected, an instance of the incident object class is created and its attributes set to the parameters of that incident. Once the incident has terminated, the object can be deleted. Thus, at any one time, the number of incident objects equals the number of current incidents. Counting stations, on the other hand, are examples of permanent objects because they represent a fixed element of the freeway system.

Knowledge Representation

Knowledge is encoded into the FRED system as a set of rules forming a knowledge base. These rules consist of antecedent and consequent parts. The antecedents specify conditions that

TABLE 1 INCIDENT OBJECT ATTRIBUTES

ATTRIBUTE NAME	TYPE	SOURCE
<i>Status</i>	Symbol	Inference Process
<i>Upstream_station</i>	Symbol	Inference Process
<i>Downstream_station</i>	Symbol	Inference Process
<i>Milepost</i>	Quantity	Operator/Report
<i>Direction</i>	Symbol	Inference Process
<i>Type</i>	Symbol	Operator/Report
<i>Duration</i>	Quantity	Operator/Report
<i>Lanes_blocked</i>	Quantity	Operator/Report
<i>Side_blocked</i>	Symbol	Operator/Report
<i>No_injuries</i>	Quantity	Operator/Report
<i>No_fatalities</i>	Quantity	Operator/Report
<i>Load_description</i>	Symbol	Operator/Report
<i>Load_weight</i>	Quantity	Operator/Report
<i>Total_lanes</i>	Quantity	Inference Process
<i>Capacity</i>	Quantity	Inference Process

TABLE 2 COUNTING STATION OBJECT ATTRIBUTES

ATTRIBUTE NAME	TYPE	SOURCE
<i>Incident_status</i>	Quantity	External
<i>Occupancy</i>	Quantity	External
<i>Volume</i>	Quantity	External
<i>Downstream_station</i>	Symbol	Constant
<i>Milepost</i>	Quantity	Constant
<i>Direction</i>	Symbol	Constant

must be satisfied before actions in the consequents can be executed. When this occurs, the rule is said to have fired. Essentially, it is the execution of the actions within the consequents of the rules that drives the system.

The most important action command is the conclusion action because it drives the inference process. Conclusions about the state of the system are drawn, or inferred, from the conditions in the antecedent of the rule, and linking conclusions to antecedents achieves a chaining of individual rules via the inference process. The state of the system is embodied solely within the objects and their attributes. Rules make reference to the attributes of objects in a number of ways. For example, they can refer to the attribute of a particular instance of an object. The following rule refers to the occupancy attribute of the object WB5, which is an instance of the counting station class:

```
if the Occupancy of WB5 > 30
then. . .
```

It is more likely, however, that rules will be generic and will refer to object classes rather than instances of that class. The following rule applies to any instance of the counting station class and is the rule that initiates the incident management process (*Incident_status* refers to the attribute of the counting station class that contains a flag indicating whether or not an incident has been detected):

```
if the Incident_status of any count_station = 1
then activate the subworkspace of inc-detection
```

Such generic rules are well suited to the encoding of knowledge.

The chaining of rules can be done in two ways: forward and backward. Forward chaining attempts to match facts about the current state of the system to conditions embodied in the antecedents of all rules in the knowledge base. Rules that are fired may then provide further information, via conclusions, about the state of the system, which can fire more rules, and so on.

Backward chaining operates in reverse. The system attempts to infer a hypothesis about the state of the system by firing only the rules that are relevant to that hypothesis. An attempt to fire the relevant rules is made by examining the conditions contained in the antecedent. To satisfy these conditions, more rules may become relevant.

Thus, forward chaining is a data-driven process, whereas backward chaining is a goal-driven process. FRED is primarily data driven but uses backward chaining when necessary. For example, the major event that drives the incident management process is the detection of an incident. To respond to a particular incident, rules may require further information. Backward chaining will occur to furnish this information.

DATA ACQUISITION AND INTEGRATION IN FRED

This section examines the way in which data are transferred from external routines and sensors to the FRED expert system. In the current implementation of FRED, there are no sensors in the physical sense but, rather, a series of external routines that simulate these sensors. Separate programs model the provision of counting station data to the incident detection algorithm and the processing of outside reports. Even in real applications, the expert system would receive data from controller routines that would operate in a manner similar to that used in FRED.

Figure 3 shows the system components from the data interfacing perspective. FRED, and the external routines interfacing with it, are all processes running concurrently and sharing the resources of the central processing unit. Communication between these processes is via data files. The external routines write any data that need to be passed to FRED to the data files, which are accessed by a separate G2 standard interface (GSI). Data read from these files are passed to objects within FRED that have been designated as having external data sources. Application-specific bridge routines must be written within the GSI module to perform the interrogation of data files and returning of data to FRED.

The G2 expert system and the GSI module communicate by object indexes. Each object that has an external routine as its data source is assigned a unique object index by GSI when the system is started. In addition to the system-assigned index, each interface object has attributes, specified by the system developer, that can be used to identify the type of data. In FRED each variable with an external source has a data type attribute that is used to determine from which external routine the data originate. For example, the volume variable for each counting station has the same value for the data type attribute. In this way, the interface bridge routine knows when to access the data file containing the current volume values.

Data can be transferred from the external routines to FRED in two ways: as solicited input or unsolicited input.

Solicited Input

As part of the process of backward chaining, the inference engine may require the value of an attribute that has a data source external to the expert system. From the expert system's perspective, the external entity is polled for the current value of that variable; once it returns, the value processing continues. At a lower level, the inference engine passes a request for data to the GSI, specifying the index of the object for which a value is sought and passing the value of the data type

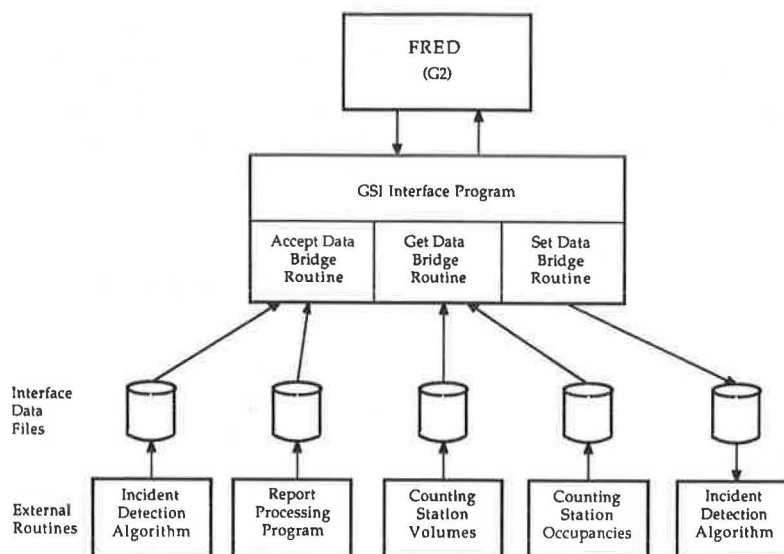


FIGURE 3 FRED data interface structure.

attribute. The “get data” bridge routine determines which data file to interrogate, opens it, and obtains the required value.

For example, when formulating responses to incidents, 30-sec volume counts are often required from the freeway mainline and ramp counting stations. These values are written to a data file every 30 sec by a separate simulation routine. If volume counts are required, a request is sent to GSI, and the “get data” bridge routine is called, passing the object index and the data type attribute. The bridge routine recognizes the data type attribute as being the one for volume counts. It opens the file containing the current volume counts, selects the appropriate value, and returns them to FRED.

In order to ensure an efficient interface, requests for data are usually grouped so that the interface bridge routine is passed an array of object indexes and data type attributes representing requests for a series of objects. In the preceding example, requests for volumes from a series of counting stations are grouped and sent to GSI at the same time. This procedure ensures that the volume data file is opened only once.

Unsolicited Input

Sensors or programs external to the expert system can send input without it being directly requested by G2. In this case, the external routine, not the expert system, is making the decision to transfer data. Such data transfers can act as interrupts to the current operation of the system. An “accept data” bridge routine is called by the interface program every second, and this routine interrogates data files that may contain input from external routines. If the information is found, it is returned along with the appropriate object index.

In FRED, the detection of an incident by the incident detection algorithm leads to unsolicited data transfer from the algorithm to FRED by the GSI interface. The Incident_status attribute of each counting station (see Table 1) has an external source, namely, the incident detection algorithm. When this

algorithm detects an incident, the identification number of the counting station immediately upstream of the incident is written to a data file. This file is examined by the “accept data” bridge routine in GSI every second. If new data are found, the bridge routine reads the identification number of the upstream counting station from the file and returns 1 as the value for the Incident_status attribute of that counting station.

Passing Data from FRED to External Routines

The behavior of the external routines can be controlled by FRED. Object attributes with external sources can be set by rules in the knowledge base. This procedure is the reverse of data retrieval because GSI is passed a value along with an object index. The “set data” bridge routine can then use that value to set some aspect of an external routine. For example, the calibration of the incident detection algorithm can be altered by setting the algorithm number and threshold parameters. Updated values for these parameters are sent to the “set data” bridge routine and then written to a data file by the bridge routine. The incident detection algorithm interrogates this data file at regular intervals; if new values are found, they are used to recalibrate the model.

Validity of Data

As mentioned, data representing the state of the external system are continually changing, and the inference engine must be able to change its conclusions on the basis of new data. This process is termed nonmonotonic reasoning. Conclusions inferred at one stage of the system may become invalid later. Thus, data and conclusions have validity intervals associated with them. For example, counting station volumes are valid for only 30 sec. Any conclusions inferred from these variables are also valid for only 30 sec.

Variables with external data sources have prescribed validity intervals. Variables whose values are inferred from rules

have their validity intervals supplied by the inference engine as the minimum validity interval of the facts used in the inference. For example, in the following rule, if facts *A* and *B* expire in 5 and 10 sec, respectively, then fact *C* will have an expiration time of 5 sec from the time at which the rule fired:

if *A* is true and *B* is true
then conclude that *C* is true

System Robustness

Any system that interfaces with a large number of external processes must be tolerant of malfunctions and breakdowns. A feature built into the inference engine causes a failed data request to be retried after a specified interval, typically 5 sec. If such attempts repeatedly fail, the variable causing the request is said to have timed out. A special rule handles such a situation. For example, the following rule fires when the volume attribute of any counting station times out:

whenever the volume of any count_station *c1* fails to receive a value
then inform the operator that "Request for volume value failed"

Additionally, rules can be written to check for erroneous data, such as negative or excessively high volume counts.

In G2 and FRED, the interface program can send error messages and statuses back to the expert system in the case of erroneous communication with an external routine. A separate body of rules within the expert system can be invoked to handle such situations.

Selective Knowledge Processing

In order to maintain a real-time expert system, consideration must be given to speed of rule processing. In FRED there is a need for control over incident management tasks, particularly when considering multiple incidents. The tasks associated with managing each incident must be organized to use the resources of the inference engine and the TOC operator efficiently.

The FRED system was built in such a manner that the rules associated with each incident management task could be invoked when required. Responsibility for invoking these sub-knowledge bases resides with a set of management rules. Rules in FRED are activated and deactivated during the running of the system, thus reducing the load on the inference engine at any one time. For example, none of the incident management rules are active when no incidents have been detected. Once an incident is detected, the incident detection rules are activated, followed by the incident verification rules, and so on.

Currently, FRED only works on a single incident at a time, but later expansion will incorporate multiple incidents, with some method of ranking the incident management tasks for several incidents. The formulation of incident responses is complicated when considering multiple incidents on the same section of freeway.

System Performance

The case study corridor used for developing the FRED prototype system is relatively small when considering the amount of surveillance data received by the system. Future research should involve on-line testing in a larger freeway environment to evaluate the actual real-time performance of the system. However, the FRED system approach will undoubtedly meet the processing requirements of incident management. The G2 shell used in the system has recently been successfully employed in a number of industrial and aerospace applications, including the U.S. space shuttle, with much more demanding performance criteria. The nature of freeway traffic control is such that the typical response times are several minutes rather than seconds. The most likely performance limitation in FRED will be its ability to handle several simultaneous incidents. Ongoing research is addressing this necessary capability.

DATA INTEGRATION IN INCIDENT DETECTION AND VERIFICATION

In order to illustrate the manner in which data arriving from different sources are integrated within FRED, the basic tasks in incident detection and verification are considered. To simplify the example, it is assumed that there are two ways to establish the existence of an incident: (a) automated incident detection with verification by operator-controlled CCTV and (b) on-site incident reports.

Automated Incident Detection and CCTV Verification

The initial event in this process is the detection of an incident by the incident detection algorithm. The detection message is sent as unsolicited input to FRED by setting the Incident_status attribute of the counting station that triggered the detection to 1. This counting station, call it *C*, is always the counting station immediately upstream of the incident site. A rule is fired whenever the Incident_status attribute of any counting station is set to 1 and leads to the creation of an incident object. Referring to Table 1, the following attributes of the new incident are set at this stage:

- *Status*. Set to the symbol "possible" at this stage.
- *Upstream_station*. Set to *C*—the name of the counting station that triggered the original detection.
- *Downstream_station*. Set to the name of the counting station object immediately downstream of the incident. Obtained from the value of the Downstream_station attribute of *C*.
- *Milepost*. The approximate milepost position of the incident set at this stage to the average of the Milepost constants of the upstream and downstream counting stations.
- *Direction*. The direction of the carriageway on which the incident was detected, for example, eastbound or westbound, copied from the Direction attribute of upstream counting station *C*.

From the operator's perspective, the next step is to verify the incident (in this simplified example, by consulting a CCTV

camera). From visual inspection of the incident site, the operator is required to enter information for the following incident attributes:

- *Type*. A selection from a prescribed list of incident types, such as overturned truck, three-car accident, and stalled vehicle.
- *Duration*. An estimate of the duration of the incident in hours.
- *Lanes_blocked*. The number of lanes of the carriageway that are blocked by the incident.
- *Side_blocked*. The side of the carriageway, left or right, blocked by the incident.
- *No_injuries*. The number of confirmed injuries resulting from the incident.
- *No_fatalities*. The number of confirmed fatalities resulting from the incident.

For an overturned truck or spilled load, the following attributes need to be set as well:

- *Load_description*. A text string of free format describing the spilled load (e.g., diesel fuel).
- *Load_weight*. An estimate of the weight of the spilled load in tons.

Additionally, at the incident confirmation stage, the operator is able to specify the location of the incident more precisely. This step is done in FRED by moving a screen marker on a freeway schematic, drawn approximately to scale, to the position of the incident as seen from the CCTV camera. Once the location is confirmed by the operator, the Milepost attribute of the incident is updated using the coordinates of the incident marker. Finally, the Status attribute of the incident is set to "confirmed" to denote a confirmed incident.

Certain attributes of an incident have their values inferred from other attribute values by formulas and rules:

- *Total_lanes*. The total number of lanes at the incident site, set by examining a table that lists the number of lanes for different milepost locations.
- *Capacity*. The unrestricted capacity of the freeway at the incident site, again obtained from a table ordered on milepost values.

Outside Reports

The incident attributes outlined previously are set in a different manner when an incident is detected only from outside reports.

A major part of the acquisition of data from outside reports is performed by the report processing program external to FRED, which allows a communications center operator to enter an outside report. The user is prompted for all necessary details. The information is then transferred to FRED as unsolicited input and, as in the previous case, an incident object is created. The following incident attributes, described previously, are derived from the report:

- Milepost,
- Direction,

- Type,
- Duration,
- Lanes_blocked,
- Side_blocked,
- No_injuries,
- No_fatalities,
- Load_description, and
- Load_weight.

The Milepost attribute is used to infer the names of the upstream and downstream counting stations. The remaining attributes are inferred as in the previous case.

Once an incident has been confirmed, the data contained within the attributes are used to formulate responses. For example, information describing the nature of the incident is used to compute the expected capacity of the freeway at the incident site, and FRED then recommends whether to close entrance ramps upstream of the incident.

CONCLUSIONS

The operating requirements of a real-time expert system to aid traffic control operators in incident management place particular emphasis on the efficient acquisition and integration of data from a number of external sensors. The manner in which the FRED system achieves this objective was outlined, with particular reference to the data interfacing between several concurrent processes in the incident detection and verification phases of incident management. The system structure developed serves as a good foundation for the development of a sophisticated and robust real-time system for management of multiple incidents that commonly occur on large freeway systems. In this respect, the refinement, expansion, and evaluation of the FRED system is ongoing.

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