

# **Representation of Sentence Meaning**

# Representing Meaning

- **Meaning Representation:** Capturing the meaning of of linguistic utterances using formal notation.
- **Meaning Representation Languages:** Frameworks that are used to specify the syntax and semantics of these meaning representations.
- **Semantic Analysis:** Mapping the linguistic utterances to these meaning representations.
- Correct meaning representation should be selected for the application.
- For certain language tasks require some form of semantic processing:
  - following a recipe
  - answering an essay question in exam
  - ...

# Meaning Representation Languages

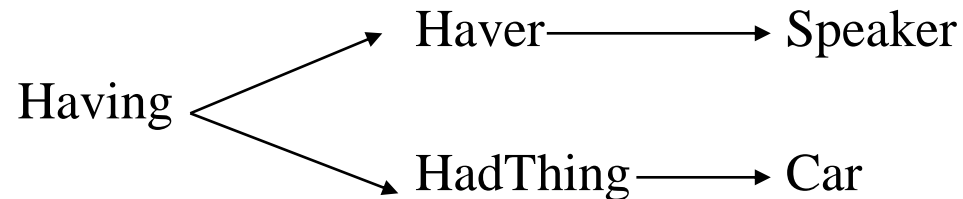
- Let us look at four frequently used meaning representation languages.
  - First Order Predicate Calculus
  - Semantic Network
  - Conceptual Dependency Diagram
  - Frame-Based Representation
- Let us look at the representation of “*I have a car*” in these four formalism.

# Meaning Representation Example - *I have a car*

## *First Order Predicate Calculus:*

$\exists x,y \text{Having}(x) \wedge \text{Haver}(\text{Speaker},x) \wedge \text{HadThing}(y,x) \wedge \text{Car}(y)$

## *Semantic Network:*



## *Conceptual Dependency:*



## *Frame-Based Representation:*

Having  
Haver: Speaker  
HadThing: Car

# What do We Expect from Meaning Representations

- To be computationally effective, we expect certain properties in meaning representations:
  - **Verifiability** -- Ability to determine the truth value of the representation.
  - **Unambiguous Representations** -- A representation must be unambiguous.
  - **Canonical Form** -- Utterances which means the same thing should map to the same meaning representation.
  - **Inference and Variables** -- Ability to draw valid conclusions based on the meaning representations of inputs and the background knowledge.
  - **Expressiveness** -- Ability to express wide range of subject matter.

# Verifiability

- **Verifiability** -- Ability to determine the truth value of the representation by looking at the information available in the knowledge base.
- **Example:**
  - Assume that we have the entry *serve(Subway,VegetarianFood)* in our KB.
  - Question: *Does Subway serve vegetarian food?*
  - The question should be converted into a logical form (a meaning representation).
  - We should be able to verify the truth value of the logical form of the question against our KB.

# Unambiguous Representations

- **Unambiguous Representations** -- A meaning representation must be unambiguous.
- **Example:**
  - Assume that we are looking the representation of “*I want to eat some place near Bilkent*”.
  - There will be different meanings of this sentence, and we will prefer one of them.
  - But that chosen meaning representation **CANNOT** be ambiguous.
- **Vagueness:** Vagueness can make it difficult to determine meaning representation, but it does not cause multiple representations.
  - *I want to eat Turkish food.*
  - Here *Turkish food* is vague, but it does not cause multiple representations.
  - Meaning representations should be able to maintain a certain level of vagueness.

# Canonical Form

- Distinct inputs can map to the same meaning representation.
  - Does Kirac have vegetarian food?
  - Do they have vegetarian food at Kirac?
  - Are vegetarian dishes served at Kirac?
- We shouldn't map these sentences to different meaning representations.
- **Canonical Form** -- The notion that inputs that mean same thing should have the same meaning representation.
- To able to map distinct inputs to the same meaning representation, we should able to know that different phrases mean the same thing such as *vegetarian food* and *vegetarian dishes*.



# Inference and Variables

- **Inference** -- Ability to draw valid conclusions based on the meaning representations of inputs and the background knowledge.
- We should be able to find the truth value of propositions that are not explicitly in KB - *- inference*.
- Example:
  - *I would like to find a restaurant that serves vegetarian food.*
  - This example is complex and we should use **variables** in its representation.
  - *serves(x, VegetarianFood)* -- a part of our meaning representation
  - If there is a restaurant serves vegetarian food, our inference mechanism should be able to find it by binding the variable x to that restaurant.

# Expressiveness

- **Expressiveness** -- Ability to express wide range of subject matter.
- The ideal situation: a single meaning representation language that could adequately represent the meaning of any sensible natural language utterance.
- Although this ideal situation may not be possible, but the **first order predicate calculus (FOPC)** is expressive enough to handle a lot of things.
- In fact, it is claimed that anything can be representable with other three representation language, it can be also representable with FOPC.
- We will concentrate on FOPC, but other representation languages are also used.
  - For example, Text Meaning Representation (TMR) used in the machine translation system of NMSU is a frame based representation.

# Predicate-Argument Structure

- All natural languages have a form of **predicate-argument** arrangement at the core of their semantic structure.
- Specific relations hold among the constituent words and phrases of the sentence. (predicate and its arguments)
- Our meaning representation should support the predicate-argument structure induced by the language.
- In fact, there is a relation between syntactic frames and semantic frames. We will try to find these relations between syntactic frames and semantic frames.
- Example:
  - *Want(somebody,something)* -- Want is predicate with two arguments

# Predicate-Argument Structure (cont.)

- Syntactic Structures:
  - *I want Turkish food.* NP want NP
  - *I want to spend less than five dollars.* NP want InfVP
  - *I want it to be close by here.* NP want NP InfVP
- Verb sub-categorization rules allow the linking of the arguments of syntactic structures with the **semantic roles** of these arguments in the semantic representation of that sentence.
  - The study of semantic roles associated with verbs is known as **thematic role**.
- In syntactic structures, there are restrictions on the categories of their arguments.
- Similarly, there are also **semantic restrictions** on the arguments of the predicates.
- The **selectional restrictions** specify semantic restrictions on the arguments of verbs.

# Predicate-Argument Structure (cont.)

- Other objects (other than verbs) in natural languages may have predicate-argument structure.

*A Turkish restaurant under fifteen dollars.*

## **Under(TurkishRestaurant,\$15)**

- meaning representation is associated with the preposition *under*.
- The preposition under can be characterized by a two-argument predicate.

*Make a reservation for this evening for a table for two persons at 8.*

## **Reservation(Hearer,Today,8PM,2)**

- meaning representation is associated with the noun reservation (not with make).
- Our meaning representation should support :
  - variable arity predicate-argument structures
  - the semantic labeling of arguments to predicates
  - semantic constraints on the fillers of argument roles.

# First Order Predicate Calculus (FOPC)

- **First Order Predicate Calculus (FOPC)** is a *flexible, well-understood*, and *computationally tractable* approach.
- So, FOPC satisfies the most of the things that we expect from a meaning representation language.
- FOPC provides a sound computational basis for *verifiability, inference*, and *expressiveness* requirements.
- The most attractive feature of FOPC is that it makes very few specific commitments for how things should be represented.

# Structure of FOPC

*Formula*  $\rightarrow$  *AtomicFormula* | *Formula Connective Formula* |  
*Quantifier Variable,... Formula* |  $\neg$ *Formula* | (*Formula*)

*AtomicFormula*  $\rightarrow$  *Predicate(Term,...)*

*Term*  $\rightarrow$  *Function(Term,...)* | *Constant* | *Variable*

*Connective*  $\rightarrow$   $\wedge$  |  $\vee$  |  $\Rightarrow$

*Quantifier*  $\rightarrow$   $\forall$  |  $\exists$

*Constant*  $\rightarrow$  A | VegetarianFood | TurkishRestuarant | ...

*Variable*  $\rightarrow$  x | y | ...

*Predicate*  $\rightarrow$  Serves | Want | Under | ...

*Function*  $\rightarrow$  LocationOf | CuisineOf | ...

# FOPC Example

*I only have five dollars and I don't have a lot of time.*

$\text{Have}(\text{Speaker}, \text{FiveDollars}) \wedge \neg \text{Have}(\text{Speaker}, \text{LotOfTime})$

*A restaurant that serves Turkish food near Bilkent.*

$\exists x \text{ Restaurant}(x) \wedge \text{Serves}(x, \text{TurkishFood}) \wedge$   
 $\text{Near}(\text{LocationOf}(x), \text{LocationOf}(\text{Bilkent}))$

*All vegetarian restaurants serve vegetarian food.*

$\forall x \text{ VegetarianRestuarant}(x) \Rightarrow \text{Serves}(x, \text{VegetarianFood})$



# Semantics of FOPC

- The truth value of each FOPC formula can be computed using meanings of the elements of FOPC.
  - Truth tables for  $\neg \wedge \vee \Rightarrow$
  - Meanings of  $\forall \exists$
  - Assigned meanings to Predicates, Constant, Functions in an interpretation.
- The truth values of our examples:
  - $\text{Have}(\text{Speaker}, \text{FiveDollars}) \wedge \neg \text{Have}(\text{Speaker}, \text{LotOfTime})$
  - $\exists x \text{Restaurant}(x) \wedge \text{Serves}(x, \text{TurkishFood}) \wedge$   
 $\text{Near}(\text{LocationOf}(x), \text{LocationOf}(\text{Bilkent}))$
  - $\forall x \text{VegetarianRestuarant}(x) \Rightarrow \text{Serves}(x, \text{VegetarianFood})$

# Inference

- Ability to determine the truth value of a formula not explicitly contained in a KB.
- We should have **inference rules** to infer new formulas from formulas available in a KB.
- For example, **modus ponens** is an inference rule.

$$\frac{\alpha \quad \alpha \Rightarrow \beta}{\beta}$$

- Example:

VegetarianRestaurant(Kirac)

$\forall x \text{VegetarianRestaurant}(x) \Rightarrow \text{Serves}(x, \text{VegetarianFood})$

---

$\text{Serves}(\text{Kirac}, \text{VegetarianFood})$

# Inference (cont.)

- We may use **forward chaining** or **backward chaining** in the implementations of inference rules.
- Implementation of certain inference rules for FOPC is not computationally effective.
- **Resolution** is a computationally effective inference rule.
  - **Prolog** uses resolution and backward chaining.
- Inference rules must be **sound** and **complete**.
  - Sound -- If a formula is derivable using inference rules, it must be valid
  - Complete -- If a formula is valid, it must be derivable.

# Inference -- Prolog Example

- Prolog uses resolution and backward chaining.

father(X,Y) :- parent(X,Y), male(X).

parent(john,bill).

parent(mary,bill).

male(john).

female(mary).

?- father(F,bill).

# Representation of Categories

- The semantics of the arguments are expressed in the form of selectional restrictions.
- These selectional restrictions are expressed in the form of **semantically-based categories**.
- The most common way to represent a category is to create a unary predicate.
  - `VegaterianRestraunt(Kirac)`
  - Here categories are relations (not objects), and difficult to make assertions about categories.
  - We cannot use `MostPopular(Kirac,VegetarianRestraunt)` because `VegetarianRestraunt` is not an object.
  - The arguments of formulas must be Terms (Predicates cannot be arguments in FOPC).

# Representation of Categories -- Reification

- Solution is to make each category an object.
  - This technique is known as **reification**.
- Thus we can define relations between objects and categories and relations between categories.
- Membership relation ISA between objects and categories.

ISA(Kirac, VegetarianRestaurant)

- A category inclusion relation AKO between categories.

AKO(VegetarianRestaurant, Restaurant)

# Representations of Events

- The simplest approach to predicate-argument representation of a verb is to have the same number of arguments present in that verb's subcategorization frame.
- But this simple approach may cause some difficulties:
  - determining correct number of arguments.
  - Ensuring soundness and completeness

- Example:

I ate.

Eating1(Speaker)

I ate a turkey sandwich

Eating2(Speaker,TurkeySandwich)

I ate a turkey sandwich at my desk.

Eating3(Speaker,TurkeySandwich,Desk)

I ate at my desk.

Eating4(Speaker,Desk)

I ate lunch.

Eating5(Speaker,Lunch)

I ate a turkey sandwich for lunch.

Eating6(Speaker,TurkeySandwich,Lunch)

I ate a turkey sandwich for lunch at my desk. Eating7(Speaker,TurkeySandwich,Lunch,Desk)

# Representations of Events -- Another Approach

- Using the maximum number of the arguments and the existential quantifiers will not solve the problem.

I ate at my desk.

$\exists x,y \text{ Eating}(\text{Speaker},x,y,\text{Desk})$

I ate lunch.

$\exists x,y \text{ Eating}(\text{Speaker},x,\text{Lunch},y)$

I ate lunch at my desk.

$\exists x \text{ Eating}(\text{Speaker},x,\text{Lunch},\text{Desk})$

- If we know that 1st and 2nd formulas represent the same event, they can be combined as 3rd formula. But we cannot do this, because we cannot relate events in this approach.



# Representations of Events -- A Solution

- We employ **reification** to elevate events to objects.

I ate.                                   --  $\exists x \text{ ISA}(x, \text{Eating}) \wedge \text{Eater}(x, \text{Speaker})$

I ate a turkey sandwich   --  $\exists x \text{ ISA}(x, \text{Eating}) \wedge \text{Eater}(x, \text{Speaker}) \wedge \text{Eaten}(x, \text{TurkeySandwich})$

I ate at my desk.                   --  $\exists x \text{ ISA}(x, \text{Eating}) \wedge \text{Eater}(x, \text{Speaker}) \wedge \text{PlaceEaten}(x, \text{Desk})$

I ate lunch.                           --  $\exists x \text{ ISA}(x, \text{Eating}) \wedge \text{Eater}(x, \text{Speaker}) \wedge \text{MealEaten}(x, \text{Lunch})$

- With the reified-event approach:

- There is no need to specify a fixed number of arguments
- Many roles can be glued when they appear in the input.
- We do not need to define relations between different versions of eating (**postulate**)

# Representations of Time

- Time flows forward, and the events are associated with either points or intervals in time.
- An ordering among events can be gotten by putting them on the timeline.
- There can be different schemas for representing this kind of **temporal information**. (the study of *temporal logic*)
- The **tense of a sentence** will correspond to an ordering of events related with that sentence. (the study of *tense logic*)

# Representations of Time -- Example

1. *I arrived in Ankara.*      2. *I am arriving in Ankara.*      3. *I will arrive in Ankara.*

- All three sentences can be represented with the following formula without any temporal information.

$$\exists w \text{ ISA}(w, \text{Arriving}) \wedge \text{Arriver}(w, \text{Speaker}) \wedge \text{Destination}(w, \text{Ankara})$$

- We can add the following representations of temporal information to represent the tenses of these examples.

1.  $\exists w, i, e \text{ ISA}(w, \text{Arriving}) \wedge \text{Arriver}(w, \text{Speaker}) \wedge \text{Destination}(w, \text{Ankara})$   
 $\wedge \text{IntervalOf}(w, i) \wedge \text{EndPoint}(i, e) \wedge \text{Precedes}(e, \text{Now})$

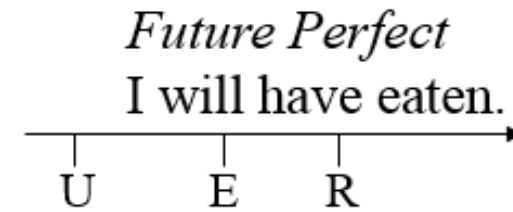
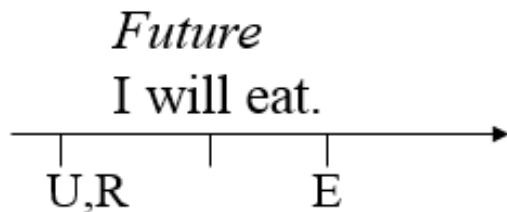
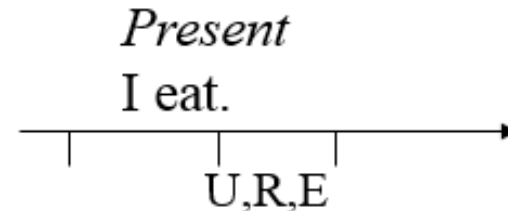
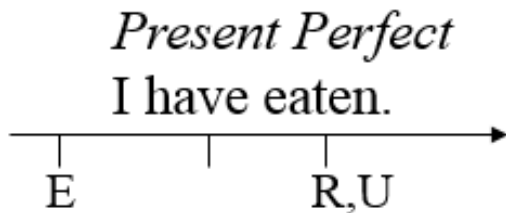
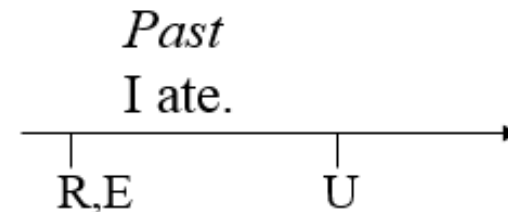
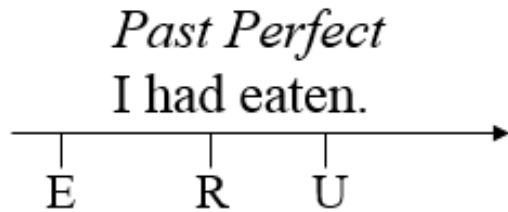
2.  $\exists w, i, e \text{ ISA}(w, \text{Arriving}) \wedge \text{Arriver}(w, \text{Speaker}) \wedge \text{Destination}(w, \text{Ankara})$   
 $\wedge \text{IntervalOf}(w, i) \wedge \text{MemberOf}(i, \text{Now})$

3.  $\exists w, i, e \text{ ISA}(w, \text{Arriving}) \wedge \text{Arriver}(w, \text{Speaker}) \wedge \text{Destination}(w, \text{Ankara})$   
 $\wedge \text{IntervalOf}(w, i) \wedge \text{EndPoint}(i, e) \wedge \text{Precedes}(\text{Now}, e)$

# Representations of Time (cont.)

- The relation between simple verb tenses and points in time is not straightforward.
  - We fly from Ankara to Istanbul.      -- present tense refers to a future event
  - Flight 12 will be at gate an hour now.      -- future tense refers to a past event
- In some formalisms, the tense of a sentence is expressed with the relation among *times of events* in that sentence, *time of a reference point*, and *time of utterance*.

# Reinhenbach's Approach to Representing Tenses



# Representations of Beliefs

- We can represent a belief as follows:
  - I believe that Mary ate Turkish food.
$$\exists u,v \text{ ISA}(u,\text{Believing}) \wedge \text{ISA}(v,\text{Eating}) \wedge \text{Believer}(u,\text{Speaker}) \wedge \text{Believed}(u,v) \wedge \text{Eater}(v,\text{Mary}) \wedge \text{Eaten}(v,\text{TurkishFood})$$
- But from this, we can get the following (which may not be correct).
$$\exists v \text{ ISA}(v,\text{Eating}) \wedge \text{Eater}(v,\text{Mary}) \wedge \text{Eaten}(v,\text{TurkishFood})$$
- We may think that we can represent this as follows, but it will not be a FOPC formula.
$$\text{Believing}(\text{Speaker}, \text{Eating}(\text{Mary}, \text{TurkishFood}))$$
- A solution is to augment FOPC with operators. (**modal logic with modal operators**).
$$\text{Believing}(\text{Speaker}, \exists v \text{ ISA}(v,\text{Eating}) \wedge \text{Eater}(v,\text{Mary}) \wedge \text{Eaten}(v,\text{TurkishFood}))$$
- Inference will be complicated with modal logic.

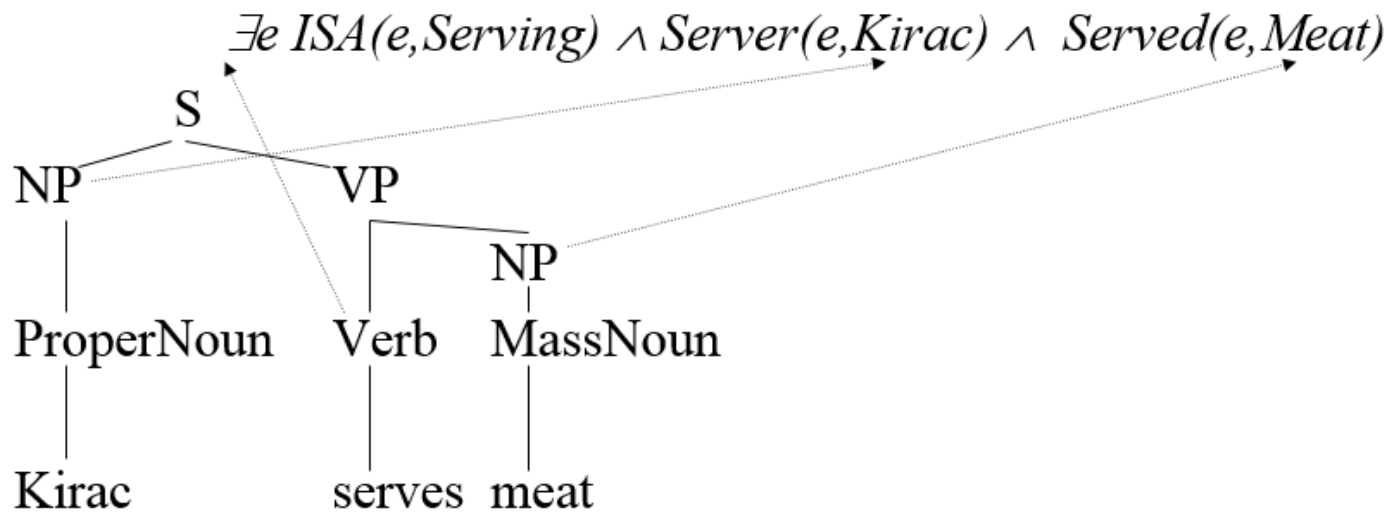
# Semantic Analysis

- **Semantic Analysis** -- Meaning representations are assigned to linguistic inputs.
- We need static knowledge from grammar and lexicon.
- How much semantic analysis do we need?
  - **Deep Analysis** -- Through syntactic and semantic analysis of the text to capture all pertinent information in the text.
  - **Information Extraction** -- does not require complete syntactic and semantic analysis. With a cascade of FSAs to produce a robust semantic analyzer.

# Syntax-Driven Semantic Analysis

- **Principle of Compositionality** -- the meaning of a sentence can be composed of meanings of its parts.
- Ordering and groupings will be important.

Kirac serves meat.





# Semantic Augmentation to CFG Rules

- CFG Rules are attached with **semantic attachments**.
- These semantic attachments specify how to compute the meaning representation of a construction from the meanings of its constituent parts.
- A CFG rule with semantic attachment will be as follows:

$$A \rightarrow \alpha_1, \dots, \alpha_n \quad \{ f(\alpha_j.\text{sem}, \dots, \alpha_k.\text{sem}) \}$$

- The meaning representation of A, **A.sem**, will be calculated by applying function f to the semantic representations of some constituents.

# Naïve Approach

ProperNoun → Kirac	{ Kirac }
MassNoun → meat	{ Meat }
NP → ProperNoun	{ ProperNoun.sem }
NP → MassNoun	{ MassNoun.sem }
Verb → serves	{ $\exists e,x,y \text{ ISA}(e,\text{Serving}) \wedge \text{Server}(e,x) \wedge \text{Served}(e,y)$ }

- But we cannot propagate this representation to upper levels.

# Using Lambda Notations

ProperNoun  $\rightarrow$  Kirac      { Kirac }

MassNoun  $\rightarrow$  meat      { Meat }

NP  $\rightarrow$  ProperNoun      { ProperNoun.sem }

NP  $\rightarrow$  MassNoun { MassNoun.sem }

Verb  $\rightarrow$  serves      {  $\lambda x \lambda y \exists e \text{ ISA}(e, \text{Serving}) \wedge \text{Server}(e, y) \wedge \text{Served}(e, x)$  }

VP  $\rightarrow$  Verb NP      { Verb.sem(NP.sem) }

S  $\rightarrow$  NP VP      { VP.sem(NP.sem) }

*application of lambda expression*

*lambda expression*

# Quasi-Logical Form

- During semantic analysis, we may use quantified expressions as terms. In this case, our formula will not be a FOPC formula.
- We call this form of formulas as **quasi-logical form**.
- A quasi-logical form should be converted into a normal FOPC formula by applying simple syntactic translations.

$\text{Server}(e, \langle \exists x \text{ ISA}(x, \text{Restaurant}) \rangle)$       a quasi-logical formula



$\exists x \text{ ISA}(x, \text{Restaurant}) \wedge \text{Server}(e, x)$       a normal FOPC formula

# Integrating Semantic Analysis into Earley Algorithm

- Modifications required to integrate a semantic analysis into an Earley parser are:
  - The rules of the grammar will have an extra field to hold semantic attachments.
  - The states in the chart will have an extra field to hold the meaning representation of the constituent.
  - The ENQUEUE function will be changed so that when a complete state is entered into the chart its semantics are computed and stored in the state's semantic field.