

VIEWS OR POINTS OF VIEW ON IMAGES

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Images like other multimedia data need to be described as it is difficult to grasp their semantics from the raw data. With the emergence of standards like MPEG-7, multimedia data will be increasingly produced together with some semantic descriptors. But a description of a multimedia data is just an interpretation, a point of view on the data and different interpretations can exist for the same multimedia data. In this paper we explore the use of view techniques to define and manage different points of view on images. Views have been widely used in relational database management systems to extend modeling capabilities, and to provide logical data independence. Since our image model is defined on an object-oriented model, we will first propose a powerful object-oriented mechanism based on the distinction between class and type. The object view is used in the image view definition. The image view mechanism exploits the separation of the physical representation in an image of a real world object from the real object itself to allow different interpretations of an image region. Finally we will discuss the implementation of the image view mechanisms on the existing object models.

Keywords: Image Database; Image Semantics; Views on Objects; Views on Images.

1. Introduction

Views are used in relational database management systems to extend modeling capabilities and to provide data independence. Relational views can be seen as formulae defining virtual relations that are not produced until the formulae are applied to real relations (view materialization is an implementation/optimization technique). View mechanisms are useful in other newly emerging application areas of database technology. In this paper, we will discuss a view mechanism for one of those areas, image databases. This work is conducted within the context of the DISIMA (DIStributed Image database MAnagement system).¹ Since DISIMA uses object-oriented technology, we will deal with object-oriented views.

Due to the volume and the complexity of the image data, image databases are commonly built on top of object or object-relational database systems. Image databases, in particular, can benefit from a view mechanism. Specifically, an image can have several interpretations that a view mechanism can model. The DISIMA system² defines a model that is capable of handling an image and all the meta-data associated with it, including syntactic characterization (shape, color and texture) of *salient objects* (region of interest) contained in the image. The level at which the syntactic features are organized and stored is called the physical salient object level. Each physical salient object can then be given a meaning at the logical salient object level. In general, salient object detection and annotation is a semi-automatic or a manual process.

Given the fact that we can manually or automatically extract meta-data information from images, how do we organize this information so that an image can be interpreted with regard to a context? That is, if the context of an image changes, the understanding of the image may change as well. Let us take a picture of the Smith family as an example. The family has four members: Mr. and Mrs. Smith, their young son John, and their daughter Jane. If John is describing this picture to his friends, he will say: “This is me, these are my mother and father and over there is my sister.” The description of Mrs. Smith for the same picture will be: “This is me, my husband, my young son John, and my daughter Jane.” The example illustrates different descriptions of the same image. They are all valid within the right context. As another example, consider an electronic commerce system with a catalog containing photographs of people modeling clothes and shoes. From the customer’s point of view, interesting objects in this catalog are shirts, shorts, dresses, etc. But the company may want to keep track of the models as well as clothes and shoes. Assume that the models come from different modeling agencies. Each of the agencies may be interested in finding only pictures in which their models appear. All these users of the same database (i.e. the catalog) have different interpretations of the content of the same set of images. In the first example, the same regions of interest in the image are given different semantics depending on the interpretation context and the second example the objects of interest are different from one context to another.

Defining an image content with regard to a context helps capture more semantics, enhances image modeling capabilities, and allows the sharing of images among several user groups with different points of view on the image. Hence, the user should not notice any difference in the way derived images and base images are handled. However, in existing object view mechanisms,^{3–6} derived objects are not first class citizens. To that end, we first defined an object view mechanism on top of which we built the image views.

An earlier version⁷ of this work was published in the IFIP DS-8 proceedings. Section 2 discusses the existing image metadata and models. Section 3 describes the image view mechanism and the object view on top of which it is built, Sec. 4 presents the image view definition language and discusses the implementation issues. Section 5 discusses object and image view issues, and Sec. 6 concludes.

2. Image, Annotation, Metadata and Semantics

A common problem in multimedia data is their interpretation and understanding. Raw multimedia data are sequences of bytes that cannot be directly used to answer queries; they have to be preprocessed. The preprocessing represents the multimedia data in a more digest form for a computer. The data obtained from the preprocessing are metadata (data about data) for the raw multimedia data. We are more interested in the content metadata that can be divided into two categories: syntactic metadata and semantic metadata. For an image, the syntactic metadata can be feature vectors representing visual features such as color, texture, and geometric shapes found in the image. The semantic metadata describes the image content in terms of real-world objects, actions and feelings.

Hence an image database can be seen as composed of two distinct parts (Fig. 1): The database component in charge of data organization and querying, and the image preprocessing component that extracts features from the images, detects and, sometimes, recognizes objects in the images. The preprocessing is, most of the time, semi-automatic as automatic recognition of objects in an image of arbitrary scene is still a research problem in Computer Vision.⁸

2.1. Annotation

Well-known image processing techniques⁹ can be used to segment an image into homogeneous regions (syntactic objects) and extract visual features (color, texture and shape). Identifying the objects and interpreting the image is more challenging and, in most cases, requires human intervention. Annotation is the process that provides interpretations (semantics metadata) to an image. In the IKONA system¹⁰ a relatively small database is manually annotated by associated keywords

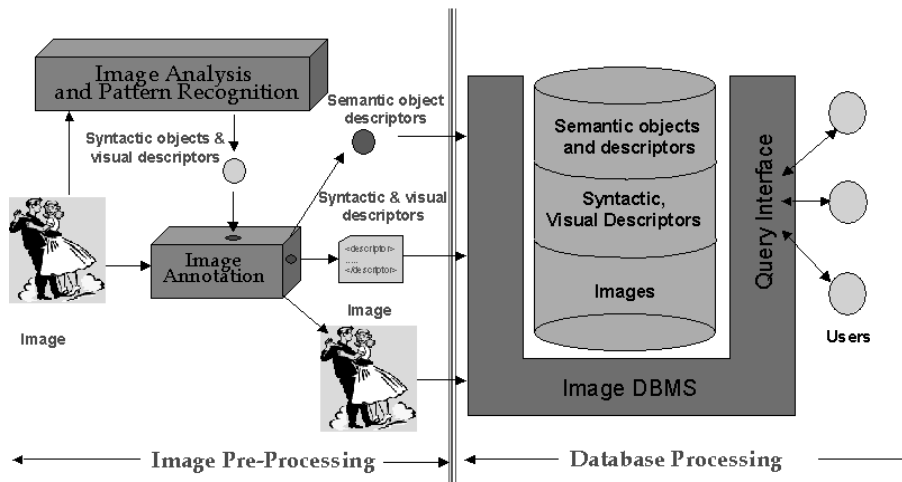


Fig. 1. Image preprocessing and database.

to the images. The system then finds correlations between the keywords and the image visual features in order to propagate the keywords and annotate a larger image database. The same approach can be used to associate keywords with regions of an image. Semi-automatic annotation techniques based on the WorldNet electronic dictionary¹¹ have also been proposed.¹² Semantic classes defined as semantic labels and visual descriptors are used to annotate the images. In the DISIMA system,¹ the link between physical salient objects (regions of an image) and the corresponding logical salient objects (meanings) are done manually.

2.2. Metadata for images

Since it is difficult to directly use raw image data in queries, the acquisition, organization and maintenance of the metadata for images are becoming key issues in image archiving. Several organizations are defining standard metadata structures for images. Among these standards are the Dublin Core metadata, MPEG-7 and RDF.

2.2.1. The Dublin Core metadata initiative

The Dublin Core (DC) is an international initiative whose goal is to define metadata for digital resources. The initial metadata format published in 1995¹³ was extended in 1996 to support images. The extension consisted the addition of two more elements to the thirteen originally defined and the adaptation of the descriptions of some existing elements. The resulting simple Dublin Core¹⁴ is composed of the following 15 elements: Title [Name given to the resource by the “Creator” or Publisher], Author or Creator [person(s) or organization(s) primarily responsible for the intellectual content of the resource], Subject and Keywords [the topic of the resource (keywords, index terms or subject classification)], Description [textual description of the resource], Publisher [the entity responsible for making the resource available], Other Contributor [person(s) or organization(s) not included under “Creator” who have made significant (but secondary) intellectual contribution to the resource], Date [date the resource was made available in its present form], Resource type [The category of the resource, e.g. “image”], Format [the data representation of the resource, e.g. “text/html”, “JPEG”], Resource Identifier [string or number used to uniquely identify the resource, e.g. “URI”, “URL”], Source [Information on the work from which the resource is derived], Language [languages of the intellectual content of the resource], Relation [relationship of the resource with other resources], Coverage [spatial locations or temporal duration characteristic of the resource], Rights Management [a link to a rights management statement].

2.2.2. The MPEG-7 metadata initiative

MPEG-7, formally known as Multimedia Content Description Interface mainly concentrates on the description of multimedia content. The objective of MPEG-7 is to

provide a rich set of organized tools to describe multimedia content and represent information about the content at different levels. MPEG-7 is an ISO standard and addresses image, text, video, audio, speech, and graphics individually along with their combinations.

The core elements of MPEG-7 standards are the Descriptors (D) to define the syntax and semantics of each multimedia feature, Description schemes (DS) to specify the structure and semantics of the relationships between their components. The Description Definition Language (DDL) is used to define the syntax of MPEG-7 and structure of descriptors. It allows the creation of new description schemes and descriptors. The DDL¹⁵ consists of (1) XML Schema structural components, (2) XML Scheme data types, and (3) MPEG-7 specific extensions.

2.2.3. *The resource description framework (RDF)*

RDF¹⁶ is defined by the World Wide Web Consortium (W3C) in order to provide meaning to resources available on the Web (among which images). Meaning is represented in RDF by sets of triples, each triple containing a subject, a verb and an object of an elementary sentence. In RDF, a document makes assertions that particular objects have properties, with certain values. This structure turns out to be a natural way to describe the vast majority of the data processed by machines. Subject and Object are each identified by a Universal Resource Identifier (URI). The simplest form of a URI is the URL. The verbs are also defined by the URIs. A property is a specific aspect, attribute or relationship, used to describe a resource. Each property has a specific meaning, defines its possible values, the types of resources it can describe, and the relationships with other properties.

2.3. *Image database models*

The main objective of the standards or proprietary metadata structures is to provide audio-visual and textual content descriptors for individual images. In the presence of a large number of images, it is crucial to find a suitable representation for images, their metadata and the operations to be applied to them in order to answer users queries. This level corresponds to the conceptual level in the ANSI relational database architecture¹⁷ where algebraic optimizations and algorithm selections are performed. Physical optimizations at the physical level (data files and indexes) consist of selecting the indexes and access methods to be used in the query processing. The image models are built on top of existing database models, mainly object-relational and object-oriented.

2.3.1. *An example of object-relational image data model*

In Ref. 18, an image is stored in a table, $T(h : Integer, x_1 : X_1, \dots, x_n : X_n)$, where h is the image identifier and x_i is an image feature attribute of domain (or type) X_i (note that classical attributes can be added to this minimal schema). The tuple

corresponding the image k is indicated by $T[k]$. Each tuple is assigned a score (ζ) which is a real such that $T[k] \cdot \zeta$ is a distance between the image k and the current query image. The value of ζ is assigned by a scoring operator $\Sigma_T(s)$ given a scoring function $s : (\Sigma_T(s))[k] \cdot \zeta = s(T[k] \cdot x_1, \dots, T[k] \cdot x_n)$.

Since many image queries are based on distance measures, a set of distance functions ($d : X \rightarrow X \rightarrow [0, 1]$) are defined for each feature type X . Given an element $x : X$ and a distance function d defined on X , the scoring function s assigns $d(x)$, a distance from x to every element of X . In addition, a set of score combination operators $\diamond : [0, 1] \times [0, 1] \rightarrow [0, 1]$ are defined.

New selection and join operators on the image table augmented with the scores are defined to allow selection of the n images with the lowest scores, which are the images whose scores are less than a given score value ρ and the images from a table T_1 that match images from a Table T_2 based on score combination functions.

2.3.2. The DISIMA model: an object-oriented approach

The DISIMA model is presented in more detail as it is the base of the view mechanism. In the DISIMA model, as shown in Fig. 2, an image is composed of physical salient objects (regions of the image) whose semantics are given by logical salient objects that represent real world objects. The DISIMA model introduces three new types: *Image*, *Physical Salient Objects*, *Logical Salient Objects* and operators to manipulate them.

From the *Image* type, the user can define new types to classify the images. Fig. 3 depicts a type hierarchy for an electronic commerce application that represents the catalogs as classes. The general *T_Catalog* type is derived from the root

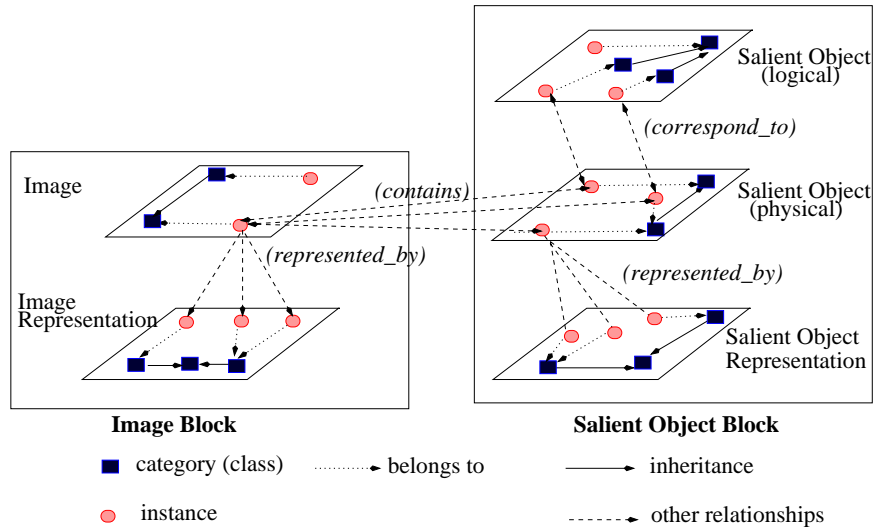


Fig. 2. The DISIMA model overview.

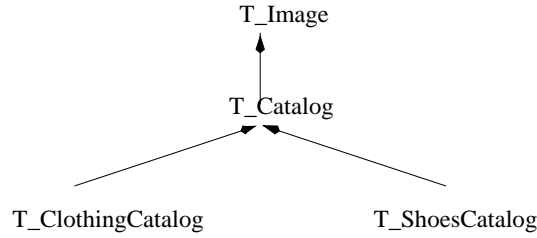


Fig. 3. An example of an image hierarchy.

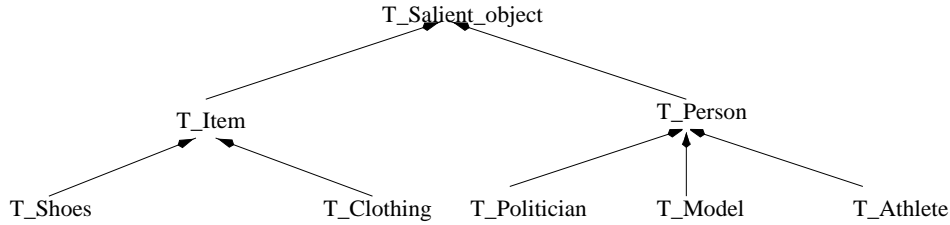


Fig. 4. An example of logical salient object hierarchy.

type `T_Image`, the root image type provided by DISIMA. The type `T_Catalog` is specialized by two types: `T_ClothingCatalog`, and `T_ShoesCatalog`.

An example of salient object hierarchy, corresponding to the image hierarchy defined in Fig. 3, is given in Fig. 4.

A *physical salient object (PSO)* is a region of an image, a geometric object (without any semantics) in a space (defined by an image) with properties like shape, color, and texture. A *logical salient object (LSO)* is the interpretation of a region. It is a meaningful object that is used to give semantics to a physical salient object.

If we denote by \mathcal{L} the set of all logical salient objects and \mathcal{P} the set of all physical salient objects, the content of an image i is defined by a pair $Cont(i) = \langle \mathcal{P}^i, s \rangle$ where:

- $\mathcal{P}^i \subseteq \mathcal{P}$ is a set of physical salient objects,
- $s : \mathcal{P}^i \rightarrow \mathcal{L}$ maps each physical salient object to a logical salient object.

An *image* i is defined by a triple $\langle Rep(i), Cont(i), Desc(i) \rangle$ where:

- $Rep(i)$ is a set of representations of the raw image in a format such as GIF, JPEG, etc;
- $Cont(i)$ is the content of the image i ;
- $Desc(i)$ is a set of descriptive alpha-numeric data associated with i .

Color and texture characterizing the whole image are part of $Desc(i)$.

The new operators introduced for objects of the new types are: *contains* or *semantic join* (to check whether a salient object is found in an image), and the *similarity join* that is used to match two images or two salient objects with respect

to a predefined similarity metric on some low-level features (color, texture, shape, etc.). In addition, DISIMA supports the *spatial join*¹⁹ on physical salient objects.

3. Defining Image Views in DISIMA

The mechanism of image views in DISIMA exploits the distinction between an image and its content represented as physical salient objects, and the distinction between a physical salient object and its meaning represented as a logical salient object. The idea is (1) to be able to define different sets of regions of interest (physical salient objects) from the same image and (2) to be able to assign different logical salient objects to a physical salient object depending on the context.

In the initial DISIMA model, a physical salient object is associated with one and only one logical salient object but to a logical salient object can correspond several physical salient objects. This allows only one context of interpretation for an image.

Definition 1. An *image interpretation context* is defined by the minimal logical salient object sub-schema and the logical salient object instances that give meanings to the physical salient objects contained in the image.

The first extension to the model is to change the relationships between the PSOs and LSOs into a many-to-many relationships. This means that a physical salient object can have different logical salient objects each of which can belong to a different interpretation context. A new interpretation can be derived from an existing one by selecting, aggregating or redefining descriptor elements. These derived interpretation are based on object view mechanism. The second extension is to allow an image to have different contents in terms of the physical salient objects. These two extensions form the image view mechanism based on an object view mechanism that was built on an object model²⁰ we defined earlier.

3.1. The object model

We separate the definition of object characteristics (a *type*) from the mechanism for maintaining instances of a particular type (a *class*) for several well known reasons.²⁰ A *type* defines behaviors (or properties) and encapsulates hidden behavioral implementations (including state) for objects created using the type as template. We use the term behaviors (or properties) to include both public interface functions (methods) and public state (public instance variables). The behaviors defined by a type describe the *interface* for the objects of that class. A *class* ties together the notion of *type* and *object instances*. The entire group of objects of a particular type, including its subtypes is known as the *extent* of the type and is managed by its class. We refer to this as *deep extent* and introduce *shallow extent* to refer only to those objects created directly from the given type without consideration of its subtypes. For consistency reasons all the type names used in this paper start with

T_- . The object model that we use in this study differentiates types from classes. Let \mathcal{C} be the set of class names. If C is a class name, $T(C)$ gives the type of C and $\Gamma(C)$ denotes the extent of the class C . We denote by \mathcal{T} , the graph representing the type hierarchy. The main concepts for the model can be summarized as follows:

- The set of types T with a partial order \leq_t (subtyping). A type T_X can define a set of behaviors ($\{B_alpha_i\}$). It can also bind behaviors to functions.
- The set of classes C . Each class X has an associated type $T_-(X) = T_X$ (type of the objects in the class) and manages a set of objects of type $T_X : [\Gamma(X)]$.
- The set of behaviors B .
- The set of functions F . A function is defined by a signature and a body.
- The set of *objects* O . Each object o belongs to a single class [$X = C(o)$] and each object knows its type [$T_X = T(o)$].

3.1.1. Derived classes

We consider two types of derived classes: simple derived classes (derived from a single class called *the parent class*) and composed derived classes (derived from two or more parent classes). We will use the term *base class* to refer to a non-derived class. In the same way, a base object refers to an object of a base class. The derivation relationship is different from the specialization/generalization one in the sense that the objects and properties introduced are obtained from data previously stored in the database.

Simple Derived Class: A simple derived class is a virtual class derived from a single parent class.

Definition 2. A **derived class** C_d is defined by (C, Φ, Ψ) where:

- C is the parent class.
- Φ , the filter, is a formula that selects the valid objects from C for the extent of C_d .
- Ψ , the interface function, defines the type of C_d by combining the functions A : *Augment* and H : *Hide* such that $\Psi = A \circ H$, where A maps a set of objects of a particular type to a set of corresponding objects in a type with some additional properties. Similarly H hides some properties.
- $\Gamma(C_d) = \Psi(\Phi(\Gamma(C)))$.

As defined, Ψ , A and H have to be applied to sets of objects of a certain type to return sets of objects of another type. To avoid introducing news terms, we will extend their applications to types.

If $\alpha, \beta, \gamma, \delta$ are properties defined in T_C , $H(T_C, \{\alpha, \beta\})$ will create a new type (let us call it $T_Restricted_C$) in which only the properties γ, δ are defined. Hence $T_Restricted_C$ is a supertype of T_C .

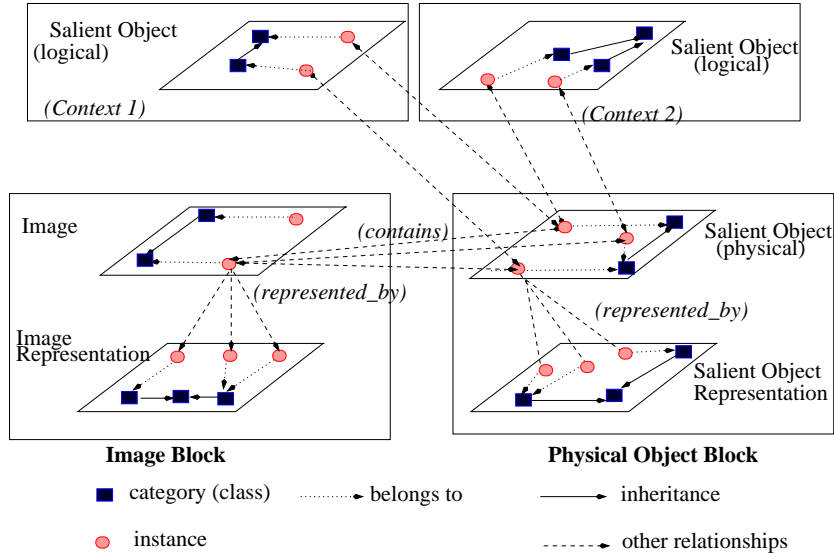


Fig. 5. Image views in DISIMA.

$A(T_C, \{(\mu : f_1), (\nu : f_2)\})$ will create a type ($T_Augmented_C$) with the additional properties μ and ν , where f_1 and f_2 are functions that implement them. $T_Augmented_C$ is a subtype of T_C .

$A(H(T_C, \{\alpha, \beta\}), \{(\mu : f_1), (\nu : f_2)\})$ defines the type T_C_d for a class C_d derived from a class C with the properties α, β of T_C hidden and μ, ν as new properties.

In general, the type T_C_d of a class C_d derived from the class C , is a sibling of $T(C)$. However, if no properties are hidden, $T(C_d) \leq_t T(C)$ (\leq_t stands for the subtyping relationship). Alternatively, if no properties are added, $T(C_d) \geq_t T(C)$ (\geq_t stands for the supertyping relationship). The notion of sibling generalizes the notion of subtyping and supertyping. The most general case where some properties are removed and new ones are added is illustrated by Fig. 6. In this example, we assume that the following properties are defined for the different types:

- T_Person (SIN: int, LastName: string, FirstName: string, Sex: char, DateOfBirth: date).
- $T_Restricted_Person$ (SIN: int, LastName: string, FirstName: string, Sex: char).
- $T_Augmented_Restricted_Person$ (SIN: int, LastName: string, FirstName: string, Sex: char, Age: int).

In Fig. 6, the extent of $Augmented_Restricted_Person$ is a subset of the extent of $Person$ with a different interface defined by the type $T_Augmented_Restricted_Person$.

Composed Derived Class: Assume that the type T_Person has two subtypes $T_Student$ and $T_Faculty$. Some of the students teach and some faculty do

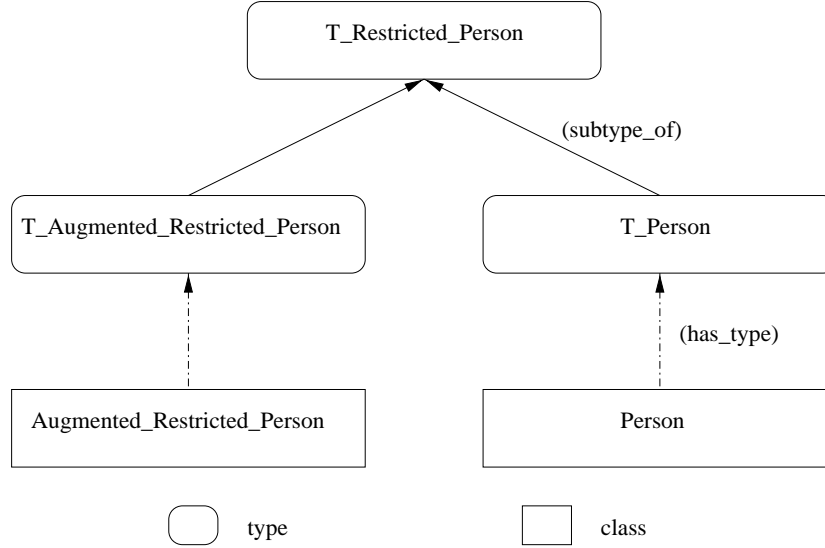


Fig. 6. Example of a derived class and its type.

only research. The Type **T_Student** has the properties (*Year: int*) and (*Teach: boolean*) while the properties (*HiringDate: date*) and (*Teach: boolean*) are defined for **Faculty**. We would like to derive a class **Teacher** of all the persons who teach with the property (*TimeServed: int*) obtained either from *HiringDate* or from *Year* depending on the type of the base object. The class **Teacher** cannot be directly derived from the class **Person** since the useful properties are not defined in **T_Person**. In the following, we propose a way (composed derived class) to solve this problem.

Definition 3. Let $(C_1, C_2) \in \mathcal{C}^2$ be a pair of classes. Then:

- $C_d = C_1 * C_2$ with a filter Φ and an interface Ψ is a composed derived class with $\Gamma(C_d) = \Psi(\Phi(\Gamma(C_1) \cap \Gamma(C_2)))$,
- $C_d = C_1 + C_2$ with a filter Φ and an interface Ψ is a composed derived class with $\Gamma(C_d) = \Psi(\Phi(\Gamma(C_1) \cup \Gamma(C_2)))$,
- $C_d = C_1 - C_2$ with a filter Φ and an interface Ψ is a composed derived class with $\Gamma(C_d) = \Psi(\Phi(\Gamma(C_1) - \Gamma(C_2)))$,

with $T(C_d)$ as a sibling of $\text{Anc}(T(C_1), T(C_2))$ where $\text{Anc}(X, Y)$ is a function that returns the first common ancestor of (X, Y) in the type hierarchy \mathcal{T} .

The semantics of the constructive operations $\{*, +, -\}$ are respectively based on the basic set operations \cap, \cup and $-$. As defined, $\{*, +, -\}$ are binary operations but the formulae obtained can be seen as terms and be combined for more complex ones. Note that C_1 and C_2 can be derived classes as previously defined. The ancestor function *Anc* works fine when \mathcal{T} is rooted. When this is not the case, a common

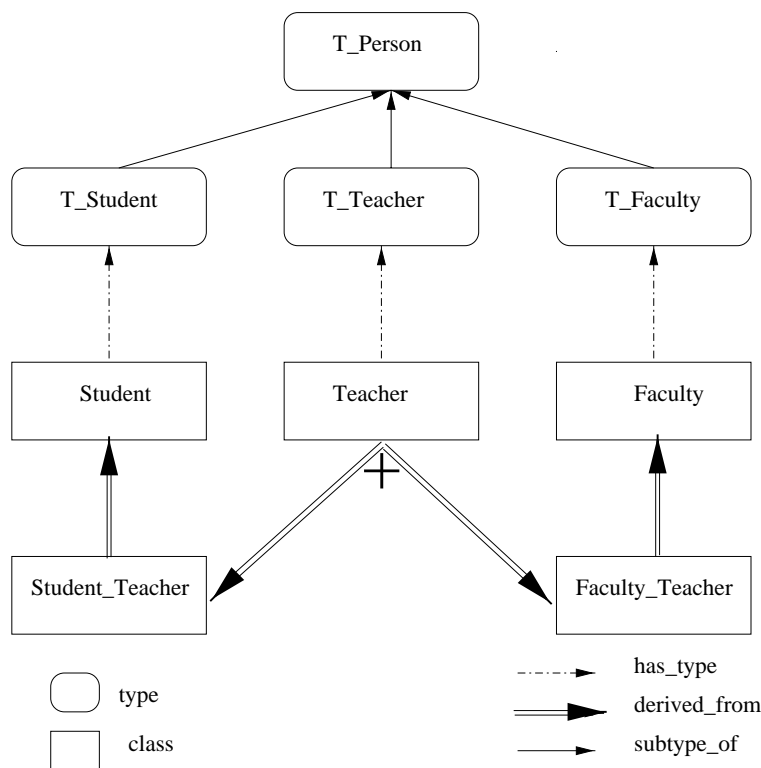


Fig. 7. An example of a composed derived class and its type.

supertype T_C is created for $T(C_1), T(C_2)$. In the worst case, T_C will not have any properties in it.

The problem of deriving a class **Teacher** can be solved by defining a simple derived class **Student_Teacher** whose extent is a subset of all the students. In the same way, we derive the class **Faculty_Teacher** from **Faculty**. **Teacher** is then defined as $Teacher = Student_Teacher + Faculty_Teacher$. The type **T_Teacher** is a subtype of **T_Person** which is the common ancestor (Fig. 7).

3.1.2. Identifying and updating a derived object

A derived object is always derived from one and only one base object although its properties can be totally different from the properties of the base object. This happens when all the properties of the base class are hidden and new ones are defined for the derived class. Hence, a derived object can be seen as a base object viewed from another angle (the interface function of the derived class). Both the derived object and its corresponding base object can be identified by the OID of the base object ($BASE_OID$). If we redefine the notion of OID as follows: $OID = \langle class_name, BASE_OID \rangle$ then the base object can be differentiated from

the derived one. This OID defines a logical identifier for any object including the derived ones independently from any view implementation technique. In the case of view materialization with incremental maintenance, an active research area,^{21–24} the derived object OID is a key candidate and can be directly used as identifier.

A derived object knows its base object. Therefore, updating a property inherited from the base type can easily be propagated to the base object. Creating new objects for a derived class should first create the objects in the base class with some possible unknown property values.

3.1.3. *Inserting the type of a derived class into the type system*

The type of a derived class has to be inserted at the right place in the type system. The insertion is simple when the associated type of a derived class is a subtype of the type of the parent class (or an ancestor of the parent classes). This happens when the definition of a derived class uses only the function *augment*. In this case, the new type is simply inserted as a subtype of the type of the parent class. The problem of inserting the new type into the type system is more complex when the function *hide* is used in the definition of a derived class, especially when the type associated to the parent class inherits some properties from another type. For example, all the behaviors inherited by `T_RestrictedPerson` in Fig. 6 and Fig. 7 are defined in `T_Person`. Therefore `T_RestrictedPerson` should be defined as a supertype of `T_Person`, and `T_Person` should have locally defined only the behaviors not in `T_RestrictedPerson`. The type system is automatically updated when classes are derived.

3.1.4. *Derived classes and subclassing*

As indicated earlier, the class derivation relationship is different from the subclass/superclass relationship. However X_d derived from a class X can be defined as a subclass of X ($X_d \leq_c X$), if the following two conditions are satisfied:

- $T_{X_d} \leq_t T_X$.
- $\Gamma(X_d) \subseteq \Gamma(X)$.

These two constraints ensure that we keep the same semantics of subclass/superclass relationships as traditional object models. That is, an object of a subclass is also an object of the superclass, and any behavior defined for objects of the superclass can be applied to objects of the subclass.

3.2. *Derived image classes and derived salient object classes*

3.2.1. *Derived image classes*

A derived image class, in addition to defining a new type, selects the needed physical salient objects from the base image possibly with new semantics.

Definition 4. A **derived image class** is a class derived from an image class that specifies the valid logical salient objects for images in its extent. If i_d is an image derived from an image i , then the set of physical salient objects contained in i_d is a subset of the set contained in i . The physical salient objects in i_d are those for which the corresponding logical salient objects belong to one of the valid logical salient objects.

In addition to redefining the type, a derived image class redefines the content of the images it contains. For example, from the *ClothingCatalog* class defined in Fig. 3, we can derive two different catalogs giving different interpretations of the images in the *ClothingCatalog* image class: the customer catalog class (*CustomerCatalog*) and the clothing company catalog (*CompanyCatalog*). The customers are interested in finding clothing from the catalog. Therefore, the valid logical salient object class is *Clothing*. In addition to the clothing, the company may be interested in keeping some information about the models.

A composed derived image class can also be created. For example, from *ClothingCatalog* we can derive the class *FemaleClothingCatalog*. We can also derive *FemaleShoesCatalog* from *ShoesCatalog*. *FemaleClothingCatalog* and *FemaleShoesCatalog* can be combined using the + operator to derive a class *FemaleApparelCatalog*. The common ancestor of *FemaleClothingCatalog*, and *FemaleShoesCatalog*

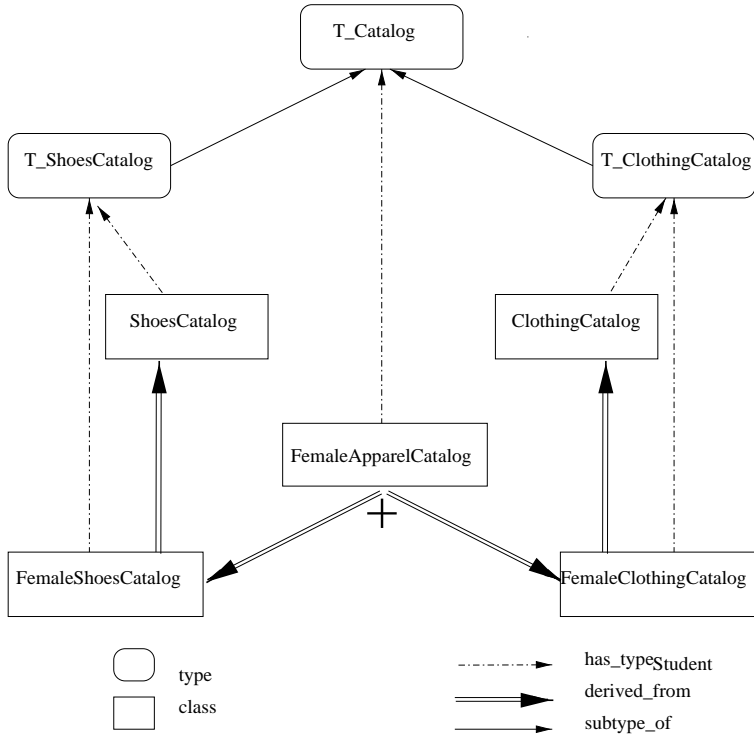


Fig. 8. An example of a composed derived image class.

is *Catalog*. Therefore the type of *FemaleApparelCatalog* has to be a sibling of the type of *Catalog* (Fig. 8).

3.2.2. Derived logical salient object classes

Definition 5 defines the content of an image i as a pair $Cont(i) = \langle \mathcal{P}^i, s \rangle$ where \mathcal{P}^i represents the physical salient objects and the function s maps each physical salient object to a logical salient object. An image i_d can be derived from i and $Cont(i_d) = \langle \mathcal{P}^i, s_d \rangle$. Assume we derived a logical salient object class L_1 from the logical salient object class L and that all the physical salient objects in \mathcal{P}^i are mapped to objects of L . If we note by f the interface function that transforms an object of L to an object of L_1 , and we define $s_d = s \circ f$, then i_d is a derived image that contains L_1 objects.

For example, the classes **FemaleClothing** and **FemaleShoes** can be respectively derived from *Clothing* and *Shoes* (Fig. 4). A composed derived class *FemaleApparel* can be derived from the two previously derived classes and the derived image class **FemaleApparelCatalog** can be defined as images containing female apparels. Of course, **T_FemaleClothing** and **T_FemaleShoes** can respectively be different from **T_Clothing** and **T_Shoes**. **T_FemaleApparel** is then, a sibling of **T_Apparel**.

Definition 5. An **image view** is defined by:

- a derived image class.
- an interpretation context defined as a set of selected logical salient object classes (base or derived).

4. The Image View Definition Language

The view definition language allows the definition of derived image and logical salient classes. Queries in the view definition are expressed in MOQL (Multimedia Object Query Language),²⁵ the query language defined for DISIMA. MOQL extends the standard object query language, OQL²⁶ with predicates and functions to capture temporal and spatial relationships. Most of the extensions have been introduced in the *where* clause in the form of new predicates as the new operators introduced by the DISIMA model are joins. The convention used in the language definition is: $[]$ for optional, $\{ \}$ (different from $\{ \}$ which are part of the language) for 0 to n times, and $|$ for one and only one among different possibilities.

- Create a derived class

```

derive { <derived class name> from <class definition>
    augment { <virtual behavior name> as <query> |
                <function name> ; }
    hide <behavior list>
    { cast <behavior — type> into < new type > }

```

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```

    [content ⟨interpretation context⟩]
    extent ⟨extent name⟩ [as ⟨query⟩]
  };
  ⟨class definition⟩ := ⟨class name⟩|
  (⟨class definition⟩ union | intersect | minus ⟨class definition⟩).

```

- Delete a derived class

```
delete ⟨derived class name⟩.
```

- Create an Image View

```

create image view ⟨image view name⟩ as
{ {⟨derived image class definitions⟩};
  {⟨derived logical salient object class definitions⟩}
}

```

- Select an Image View

```
set image view to ⟨image view class name⟩
```

The *derive* command is used to define a derived logical salient object class, as well as derived image classes. The classes that the *derived classes* are derived from one or more base or derived classes. The query in the *extent* clause defines the derived class extent and must return a unique subset of the combination of the parent class extents. The *augment* clause is used to define new properties. A query can invoke an existing property. In this case, the keyword *this* is used to refer to the current object. If $(\alpha : T(C))$ is a property and C_d is a class derived from C , then the clause *cast* can be used to cast the type of α into $T(C_d)$. The *content* clause allows to define the logical salient object classes that will form the interpretation context. This clause is used only for derived image classes and is used to select the interpretation context. It is complemented by the *cast* clause. If a logical salient object class mentioned in the view is a derived class, then the salient objects from parent image should be casted to the derived type. Assume a salient object class S_d is derived from class S and an image i (element of the image class I) contains a salient object of type $T(S)$. If we derive an image class I_d from I with the clause *content* S_d , image i_d derived from i will contain a salient object of type $T(S_d)$ instead of $T(S)$. For example, in the image view `CustomerClothing` that follows, an image of `CustomerCatalog` contains elements of `CustomerClothing`, rather than `Clothing` as salient objects. Derived classes can dynamically be deleted.

An image view defines the interpretation context for one or more image classes. An image view is created using the command *create image view* and contains derived image and salient object classes. The salient objects classes do not have to be derived. They can be base classes introduced to provide complementary interpretations. In this case the image view just selects the the logical salient objects

to form the interpretation context. The *Set Image View to* command is used to set the current image interpretation for querying purpose.

4.1. Examples of Image Views

In the following, we give some examples of image views derived from the catalog database. The corresponding schema expressed in the ODMG object model²⁶ is given in the Appendix. The schema given in the Appendix can be seen as the most general view, the view of the company owner of the catalog: each image contains models and clothes. The example views correspond to the Customer View, the Female Clothing Catalog View, and the Female Apparel Catalog view.

```

Create Image View CustomerCatalog
{
  derive    {CustomerClothing from Clothing
  augment  inStock as this.inStock();
            avgPriceForType as
            avg(Select c.price
            From Clothes c
            Where c.type = this.type);
  hide     stock, lastOrderDate, lastArrivalDate, nextArrivalDate
  extent   CustomerClothes};

  derive    {CustomerCatalog from ClothingCatalog
  hide     photographer, date, time, place
  cast     accessories into Set<Ref<CustomerCatalogs>>
  cast     Clothing into CustomerClothing
  extent   CustomerCatalogs
  content  CustomerClothing}
}

```

The derived class `CustomerClothes` redefined `Clothes` for the customers' use. Attributes `stock`, `lastOrderDate`, `lastArrivalDate`, `nextArrivalDate` are hidden and the virtual attribute `avgPriceForType` returns the average price for this type of clothing.

The image view `CustomerCatalog` uses the image class `Catalog` renamed as `CustomerCatalog` with `CustomerCatalogs` as its extent name. All the images are available but their content will be limited to objects of `CustomerClothes` which redefines `Clothing`. Attributes `photographer`, `date`, `time`, `place` are hidden. The attribute `accessories` was defined as a set of images from `Catalog`. Its type has to be changed to set of `CustomerCatalogs` to ensure consistency.

```

Create Image View FemaleClothingCatalog
{
  derive    {FemaleClothing from CustomerClothing

```

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```

extent FemaleClothes as
  Select c
  From CustomerClothes c
  Where c.sex = "female" or c.sex = "unisex";

derive {FemaleClothingCatalog from ClothingCatalog
hide   photographer, date, time, place
cast   accessories into Set<Ref<CustomerCatalogs>>
cast   CustomerClothing into FemaleClothing
extent FemaleClothingCatalogs
content FemaleClothing};
}

```

Only images containing female items are selected from the clothing catalog. The salient objects are restricted to female clothing.

Create Image View FemaleApparelCatalog

```

{
  derive {FemaleShoes from Shoes;
augment inStock as this.inStock();
  avgPriceForType as
  avg(Select s.price
  From Shoes s
  Where s.type = this.type);
hide   stock, lastOrderDate, lastArrivalDate, nextArrivalDate
extent FemaleShoesExtent as
  Select s
  From ShoesExtent c
  Where c.sex = "female" };
derive {FemaleShoesCatalog from ShoesCatalog
hide   photographer, date, time, place
extent FemaleShoesCatalogs
content FemaleShoes};
derive {FemaleApparelCatalog from FemaleClothingCatalog union
  FemaleShoesCatalog
extent FemaleApparelCatalogs};
}

```

The *FemaleApparelCatalog* combines the *FemaleClothingCatalog* and the *FemaleShoesCatalog* into a new derived catalog.

4.2. Implementing Derived Classes in DISIMA

The distinction between types and classes is not supported by most object-oriented languages in current use. DISIMA is being implemented on top of Object-Store²⁷

using C++. DISIMA provides types for image and logical salient objects that can be subtyped by the user. The implementation we describe in this section simulates the idea using C++. We implement all our types as C++ classes. We call these C++ classes *type classes* and their names start with *T_*. For example *T_Person* will be a type class for the class *Person*. Our classes are objects of the C++ class *C_Class*. *C_Class* has a subclass *D_Class* for derived classes. The properties defined for *C_Class* are:

- Name: name of the class.
- Type: type class name.
- SuperclassList: list of the superclasses.
- SubclassList: list of the subclasses.
- ShallowExtent (virtual function): The shallow extent of the class.
- DependentsList: list of classes derived depending on this one.

The properties defined for *D_Class* are:

- Base ClassList: list of the classes it is derived from.
- Filter: filter function.
- ShallowExtent: redefined.
- MaterializationFlag: set when the ShallowExtent is up-to-date.
- Change: function used to unset the MaterializationFlag.

The *DependentsList* in the class *C_Class* contains all the classes derived from that class and also all the derived classes for which an augmented property is computed using objects of that class. Since the type of a derived class can be different from the type of its base class we choose to materialize the derived class extent. An object of *C_Class* represents a user's class and the extent (*ShallowExtent*) property returns objects of the type class (*Type*). The *SubclassList* can be used to recursively compute the deep extent. To simplify the materialization process, we only store one level of base class. That is, the *Base ClassList* of a derived class contains only non-derived classes. A derived class extent is materialized the first time the class is referred to and the materialization flag is set. Each time new objects are created, modified or deleted in a base class, a change message is sent to each of the classes in the *DependentsList* to unset the materialization flag. If the materialization flag is unset when a derived class is accessed, the derived class extent is recomputed and the materialization flag is reset.

When the augmented properties of a derived object are computed from the single base object without any aggregate, the management algorithm for incremental view maintenance can easily be implemented as follows. An object of a derived class contains the OID of the base object it is derived from. The *Change* method passes the OID of the changed base object (new, deleted or updated) to the derived class object where it is kept in the *ChangeList* of the derived class object. The *ChangeList* can then be visited to update or create the derived objects for modified or new base objects and to delete derived objects corresponding to deleted base objects.

5. Object Models and Views

5.1. Object models

In most object systems, as indicated earlier, the terms *type* and *class* are used interchangeably to refer to one or more of the following notions: (1) real-world entity, (2) programmatic interface, (3) implementation of the programmatic interface, (4) internal machine representation, (5) factory for creation of instances, (6) maintainer of the extent (the set of all the instances of a class). For example, in GemStone,²⁸ *class* refers to all of the above notions. In ObjectStore,²⁷ *type* refers to (2) and (4), while *class* refers to (1), (3), and (5). Alternately, the *type* of an object of a particular *class* is implicitly defined by that *class*, so it may also be argued that *class* in ObjectStore actually refers to all notions except (6). In O_2 ,²⁹ *type* refers to (2) and (4), while *class* is a special kind of *type* that refers to (1) through (6). In Iris,³⁰ *type* refers to (1), (4), (5), and (6); (2) and (3) are defined with respect to the *type* but are not a part of the *type*. In VODAK³¹ (which is built on top of ObjectStore but has a different data model), *type* refers to (2), (3), and (4), while *class* refers to (1) and (5). *Type* in VODAK has two components: the *interface*, referring to (2), and the *implementation*, referring to (3). Both parts of the VODAK *type* can define the representation. In relational database systems, *type* refers to (2) and (4). XML³² and DAML + OIL,³³ through XML Schema, offer a variety of types that refer to (2) and (4). But XML has no explicit constructs to define classes. RDF,¹⁶ has only literal (strings) as type (2) and (4) and uses the $\langle rdfs : class \rangle$ to define classes (5).

5.2. Views and objects

Several object view mechanisms have been proposed since the early 1990s.^{3-6,34} In general, the main problems with these views⁶ are (i) expressive power (restrictions on queries defining views); (ii) reusability and modeling accuracy (insertion of the views into the generalization hierarchy); and (iii) consistency (stability in OID generation).

Problems (i) and (ii) are somewhat related. For example, using the view mechanism in Ref. 5, if the user wants the view class to be linked to the generalization hierarchy, the query that generates the view class has to be restricted. In addition, problem (ii) raises a typing problem (how is the type of the virtual class related to the type hierarchy?) and a classification problem (how is the extent of a virtual class related to the existing ones?). Finding an answer to these two questions in an environment where the only relationship is the *is-a* relationship can lead to contradictions. The distinction between the derivation hierarchy and the generalization hierarchy in our proposal, based on the distinction between type and class, provides an elegant solution to problems (i) and (ii). For example, the fact that the associated type of a class is not tied to the class allows a view based on a projection of its parent type behaviors to be created with an associated type defined as a supertype

of the parent type, while the extent is a subset of the extent of the parent class extent. The type of a derived class is computed from the view definition and inserted into the type system. The insertion of the new type can lead to local reorganization of the type hierarchy without any effect on existing classes.

In addition, the object view mechanism proposed in this paper allows us to derive classes from several existing ones. Problem (iii) is solved by the fact that a derived object is seen as a base object with a different interface function. A derived object and its base class share the same OID but are uniquely identified by the pair $\langle class_name, OID \rangle$, which is invariant even if the derived object is recomputed. These problems needed to be solved in order to provide clean image views as for the user there should not be any difference between an image view (a derived context) and a base context. The two levels of salient objects ensure the semantic independence and multi-representation of salient objects.

6. Conclusion

Because of their nature, images are often represented using object or object-relational models. Hence, an image view mechanism should be an extension of object models. The aim of an image view mechanism is to provide a way to define different context of interpretations for the same image without having to duplicate the image data.

The image view is based on an object model. The distinction between type and class in the model used in our view mechanism allows the manipulation of both the type and the extent in a view definition. The distinction also facilitates the integration of the type of a virtual class into the type hierarchy. A virtual class can be derived from one or several classes without any constraint on the query defining it. The DISIMA model separates the objects contained in an image (*physical salient objects*) from their semantics (*logical salient objects*). Using our object view mechanism, we proposed an image view mechanism that allows us to give different semantics to the same image. For example, a derived image class can be defined by deriving new logical salient object classes that give new semantics to the objects contained in an image or by hiding some of the objects using a derived image class.

The main contributions of this paper are the proposal of a powerful object-oriented view mechanism based on the distinction between class and type, a proposal of an image view mechanism based on image semantics and the image view implementation using a language that does not intrinsically support the distinction between class and type.

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Appendix A

```

class Image{
    Set<Ref<Representation>> representations;
    Set<Ref<PhysicalSalientObject>> physicalSalientObjects
        inverse image;
    // Methods
    display();}
class Catalog: Image{
    Person photographer; Date date; Time time; String place;}
class LogicalSalientObject{
    Set<Ref<PhysicalSalientObject>> physicalSalientObjects
        inverse logicalSalientObject;
    //Methods
    Region region(Image m); // salient object's region in image m
    Color color(Image m); // salient object's color in image m
    Texture texture(Image m); // salient object's texture in image m}
class Person: LogicalSalientObject{
    String name; String occupation; Address address;}
class Model: Person{
    String : agency;}
class Apparel: LogicalSalientObject{
    String name; String type; Real price; Set<Real> size;
    Manufacturer manufacturer; Integer stock; String colors;
    Date lastOrderDate; Date lastArrivalDate; Date nextArrivalDate;
    //Methods
    Boolean inStock(); // true if the the clothing is in stock}
class Clothing: Apparel {
    Set<Ref<Catalog> accessories;// images of items that match with the cloth}
class Shoes: Apparel {
    String sole;String upper;}
class PhysicalSalientObject{
    Ref<LogicalSalientObject> logicalSalientObject
        inverse physicalSalientObjects;
    Ref<Image> image
        inverse physicalSalientObjects;
    Region region; Color color; Texture texture}

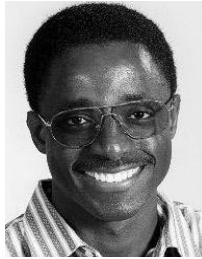
```

Set⟨Ref⟨LogicalSalientObject⟩⟩ LogicalSalientObjects; //all salient objects
 Set⟨Ref⟨Person⟩⟩ Persons; //salient objects of type Person
 Set⟨Ref⟨Model⟩⟩ Models; //salient objects of type Model
 Set⟨Ref⟨Clothing⟩⟩ Clothes; //salient objects of type Clothing
 Set⟨Ref⟨Shoes⟩⟩ ShoesExtent; //salient objects of type shoes
 Set⟨Ref⟨Image⟩⟩ Images; //all images

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