

A multi-image approach to CADx of breast cancer with integration into PACS

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ABSTRACT

While screening mammography is accepted as the most adequate technique for the early detection of breast cancer, its low positive predictive value leads to many breast biopsies performed on benign lesions. Therefore, we have previously developed a knowledge-based system for computer-aided diagnosis (CADx) of mammographic lesions. It supports the radiologist in the discrimination of benign and malignant lesions. So far, our approach operates on the lesion level and employs the paradigm of content-based image retrieval (CBIR). Similar lesions with known diagnosis are retrieved automatically from a library of references. However, radiologists base their diagnostic decisions on additional resources, such as related mammographic projections, other modalities (e.g. ultrasound, MRI), and clinical data. Nonetheless, most CADx systems disregard the relation between the craniocaudal (CC) and mediolateral-oblique (MLO) views of conventional mammography. Therefore, we extend our approach to the full case level: (i) Multi-frame features are developed that jointly describe a lesion in different views of mammography. Taking into account the geometric relation between different images, these features can also be extracted from multi-modal data; (ii) the CADx system architecture is extended appropriately; (iii) the CADx system is integrated into the radiology information system (RIS) and the picture archiving and communication system (PACS). Here, the framework for image retrieval in medical applications (IRMA) is used to support access to the patient's health care record. Of particular interest is the application of the proposed CADx system to digital breast tomosynthesis (DBT), which has the potential to succeed digital mammography as the standard technique for breast cancer screening. The proposed system is a natural extension of CADx approaches that integrate only two modalities. However, we are still collecting a large enough database of breast lesions with images from multiple modalities to evaluate the benefits of the proposed approach on.

Keywords: Breast cancer, CADx, PACS, RIS, mammography

1. INTRODUCTION

To support a radiologist with the diagnosis of mammographic lesions, we have recently developed a knowledge-based system for the computer-assisted diagnosis (CADx) of mammographic masses.¹⁻³ Within this CADx-system, content-based image retrieval (CBIR) methods as well as case-based reasoning (CBR) techniques⁴ are applied to retrieve similar lesions with known diagnoses from a library, an adequately organized picture archiving

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system or from an annotated image data base of reference lesions.⁵ Based on the histological validated ground truth diagnoses of the retrieved lesions, a diagnosis proposal for the unknown query lesion is generated and displayed to the radiologist. Furthermore, to make the reasoning process more transparent to the operator, the retrieved lesions, their corresponding diagnoses and additional annotations⁶ are presented to the radiologist in an adequate way together with the generated diagnosis proposal.

While our CADx system currently works only on the lesion level for the retrieval of related images and information, the diagnosis of a suspicious mammographic lesion is usually based on further information. Such additional information consists of demographic and anamnestic data available from the patients health care record, the appearance and location of the lesions in corresponding mammographic views (e.g. the craniocaudal (CC) and mediolateral-oblique (MLO) views), and the appearance of the lesions in additional imaging modalities, such as ultrasound or MRI. In other words, to discriminate benign and malignant mammographic lesions, the experienced radiologist does not operate on the lesion level (as most CADx-systems do) but on the full case level, and thus integrates all available sources of information into the process of decision making. Several groups have already proposed CADx approaches that integrate at least some of the modalities and other information sources mentioned above. Veldkamp et al.⁷ integrate features extracted from two mammographic views for the discrimination of benign and malignant mammographic microcalcifications. Gupta et al.⁸ integrate texture features from two mammographic views for the classification of mammographic masses. The work of Drukker et al.⁹ is one of the few examples of CADx approaches that integrate lesion features from multiple modalities. More precisely, they integrate features extracted from ultrasound images and multiple mammographic views to classify benign and malignant breast lesions. Although their case database is relatively small (100 lesions), they conclude that computerized classification of cancer significantly improves when lesion features from ultrasound and mammography are combined.

Digital mammography systems are currently the gold-standard for the acquisition of mammograms. In the past years, they have been established worldwide as the successors of analogue film-based mammography systems. The characterization of mammographic masses based on analogue or digital mammograms is based on a pair of two-dimensional X-ray images of each breast obtained from two different angles (e.g. CC and MLO). Since mammographic masses are often more dense than their surrounding tissue, they usually show up as irregular white areas commonly referred to as shadows. Within two-dimensional X-ray images, such shadows are often hidden behind a dense breast tissue, which leads to a superposition of mass and tissue in the resulting mammogram. Thus, manual or computer assisted reading of such mammograms are often error-prone, as such a superpositions may either result in a false-negative or a false-positive diagnosis. However, digital breast tomosynthesis (DBT) systems, which have been recently introduced as a new imaging modality for mammography have a high potential to succeed digital mammography systems as the standard imaging technique for breast cancer screening in the next decade.¹⁰⁻¹³ With DBT, low-dose mammograms are acquired at a number of different projections over a limited range, and the 3D breast volume can thus be reconstructed.¹⁴ Hence, since DBT allows the reconstruction of a stack of two-dimensional slices, the depicted mammographic lesions are free of tissue superpositions and can thus can be better interpreted by both, man and machine.¹⁵ Even though only prototype DBT systems are available so far, considerable work on the computer aided detection (CADe) of breast lesions in DBT images has already been done.¹⁶⁻²⁰ However, work on the computer aided diagnosis (CADx) based on DBT data is sparse.²¹

Considering these observations and developments, the next generation of CADx systems for breast cancer diagnosis are expected to mimic the process of decision making and diagnosis of radiologists more closely and thus will rely on the full case data instead of isolated images from the lesion level. Furthermore, we expect, that DBT will soon be FDA-approved and thus be available for a broader clinical use. Accordingly this paper describes the required extensions to extend our CADx-system from the lesion to the case level.

2. MATERIALS AND METHODS

From our point of view, three main extensions are required to migrate CADx for mammographic lesions from the lesion level to the full case level:

1. The characteristic features describing a lesion, which are currently extracted from relevant image regions from two-dimensional mammograms, must be extended to be extracted from several sources. These data

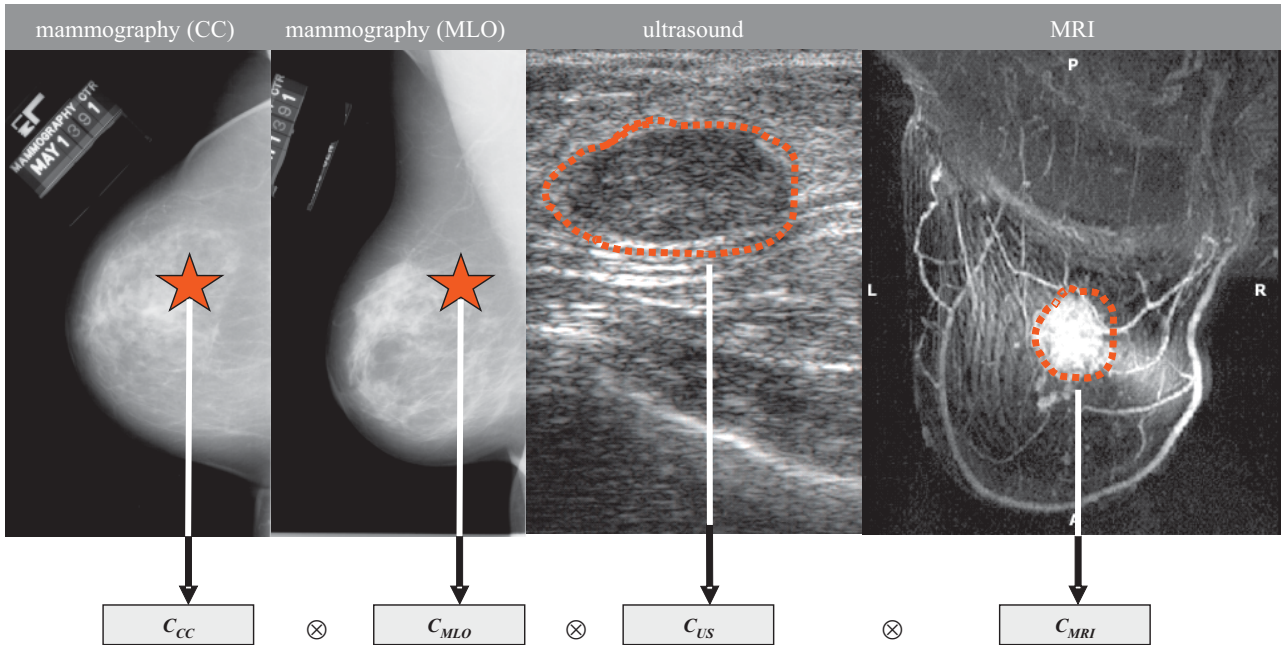


Figure 1. Generation of an extended feature vector characterizing a mass lesion from multiple modalities, including the craniocaudal (CC) and mediolateral-oblique (MLO) projections of mammograms, ultrasound and MRI.

sources feeding a highly dimensional feature vector include the description of the same lesion from multiple 2D mammographic projections (CC and MLO), multiple modalities (mammography, ultrasound, MRI) as well as multiple frames of the 3D image stacks from DBT and MRI. Furthermore, available demographic and anamnestic patient data must be considered as an essential part of the extended feature vector.

2. The architecture of the CADx system must be extended in such a way as to allow the fusion of characteristic attributes from multi-medial sources related to the mammographic case. These sources include annotated and validated reference images, alphanumeric text as well as class indices taken from ontologies or standardized vocabularies such as the Breast Imaging - Reporting And Data System (BI-RADSTM) standard of the American College of Radiology (ACR).²² The work of Drukker et al.⁹ suggests that classification performance strongly depends on the specific method for combining features from multiple images of a lesion.
3. Finally, in order to access and retrieve the additional information sources and thus support the radiologist during the process of clinical decision making, the case-based CADx approach must be integrated into the infrastructure of the radiology information system (RIS), as well as the picture archiving and communication system (PACS). It should also support the integration with health care enterprise (IHE) profiles.

2.1 Multi-image features

The first migration stage of the proposed extended CADx system will allow to extract and combine characteristic lesion attributes from corresponding regions-of-interest (ROIs) depicting the lesion in different projections in two-dimensional mammograms, such as the CC and MLO views. This approach mimics the idea that a radiologist will always use both projections of a breast for the diagnosis of a mammographic mass to partially resolve superpositions of masses and tissue. This stage of course requires linking regions of interest in different mammographic views that correspond to the same lesion. We have implemented an approach based on the method

proposed by Engeland et al.²³ to accomplish this task. Furthermore, optional features from the same mass lesion obtained from additional breast imaging modalities will be able to be added to the feature vector on demand (cp. Figure 1). Ultrasound and MRI are two common imaging modalities, which are routinely used by radiologists to provide further information about suspicious breast lesions previously detected in mammograms.²⁴ As the superposition of masses and surrounding breast tissue is especially problematic for an image-based computer-assisted diagnosis of mammographic masses, it is important that characteristic features obtained from a stack of superposition-free two-dimensional slices from DBT images can be integrated and added to the extended feature vector. Even though we will provide this capability to our system, we expect that a broad distribution, availability, and acceptance of DBT will still take a couple of years. Furthermore, all available demographic (sex, age, race, etc.) and anamnestic data (smoker, children, inheritance, etc.) about the patient must be described as categorical data and be considered as an essential part of the extended feature vector. Previous experiments have shown that clinical decision making based on such patient data can significantly improve the diagnosis suggestions of a CADx system.²⁵

2.2 CADx system architecture

In order to handle image features calculated from multiple views and multiple modalities as described in Section 2.1, the current architecture of the CADx system must be revised. The original *image-to-image* relation for the comparison and retrieval of similar images in the reference database on the *lesion level* must be extended towards a *set of images (SOI)-to-SOI* relation, where the described multi-view and multi-modality features can be embedded. Especially the extended comparison function between such SOI's must be newly defined, allowing the needed possibility to compare only partially filled set of images. This special case must be considered, since a SOI may not always be complete. Most often, only mammographic images are available for a SOI, as ultrasound and MRI modalities are only applied to a patient if there is found indication for a mass lesion. Accordingly, the integration of features obtained from DBT slices must also be supported, since ROIs depicting a mass in different DBT slices must be treated similar to ROIs depicting the same mass in different mammographic projections. Additionally, image registration and geometric standardization for DBT images is required as pre-processing steps.²⁶

Furthermore, the data structures capturing medical meta information (e.g., previously annotated ground truth based on histological findings) will be extended. While the knowledge base of our current CADx system consists of a structured collection of interactively annotated ROIs together with histology-based ground-truth diagnosis, the knowledge-base needed for the proposed extended CADx system must be capable to handle all information sources associated with a given mammography case. The image and information sources making up a typical case of a breast mass lesion consist in the different related images (mammography, ultrasound, DBT, MRI) stored on (distributed) RIS and PACS servers as well as standardized classification ontologies the description of tissue type and lesions (e.g. BI-RADSTM) and alphanumeric free text and categorical data (anamnestic and demographic patient data as well as additional patient information) stored on the hospital information systems (HIS) server.

2.3 PACS integration

Besides the extension of the feature vector describing a mammographic lesion and the thus related system architecture as described above, furthermore, the CADx system must be integrated with PACS, radiological (RIS) and hospital information systems (HIS). In previous work, we have already proposed a possibly framework for the integration of content-based image retrieval in medical applications (IRMA) with PACS and RIS²⁷ which is currently extended to support the integration with health care enterprise (IHE) profiles.²⁸ Generic communication modules that communicate with the RIS, HIS and PACS servers over standard protocols such as Digital Imaging and Communications in Medicine (DICOM) and Health Layer 7 (HL7) have been used for this integration. The application programming interface (API) of the viewing software used by the radiologist when reading mammograms has been used to interconnect the web-based IRMA interfaces providing extended query refinement and relevance feedback.²⁹

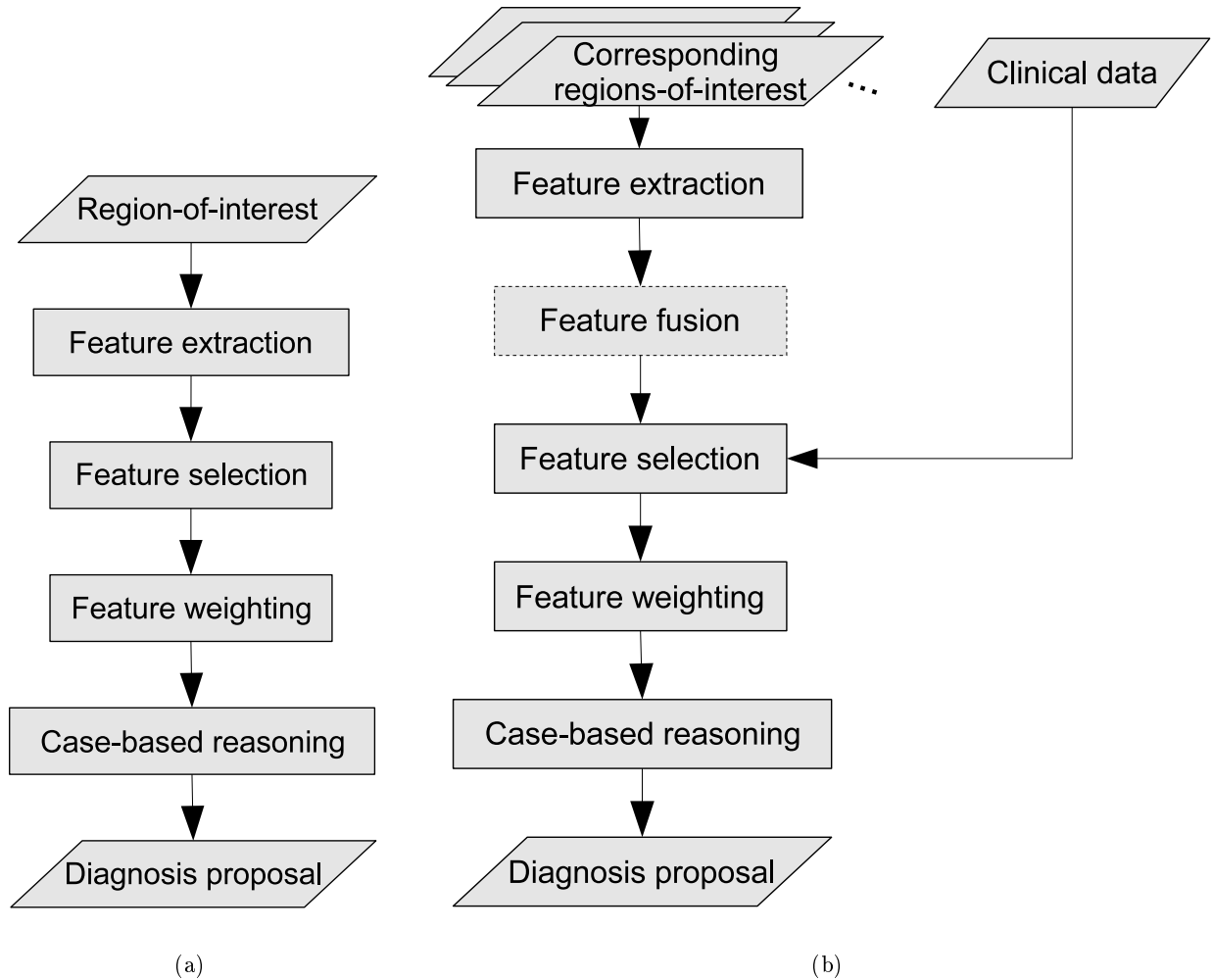


Figure 2. Comparison of the system architecture of the previously proposed CADx system that works on the lesion level (a) and the system proposed in this work that is extended to work on the full case level (b).

3. RESULTS

Figure 2 illustrates the system architectures of the previously proposed CADx system and the proposed extended version. In the extended version the reasoning process has been enhanced by an additional module that allows the fusion of characteristic lesion attributes extracted from corresponding ROIs from different image views, image frames and image modalities. This module is optional as an alternative way of handling lesion attributes from corresponding ROIs is to simply concatenate the attribute vectors. Lesion attributes extracted from textual clinical data such as patient information or histological data can also be added to this attribute vector.

We have previously evaluated the performance of our current CADx approach on the *lesion level* using ROC analysis. The resulting area under the ROC curve was $A_z = 0.86 \pm 0.03$ on the publicly available DDSM mammography database.³⁰ This database contains over 2000 cases acquired in the early 1990s. Each case includes two images of each breast, along with the patient’s age, breast density rating, and BI-RADS standard lexicon compliant descriptions of abnormalities. The database contains normal, benign, and malignant cases. As the DDSM only consist of pure (analogue) mammography cases, the construction and validation of the proposed extension of the CADx system to the *case-level* requires an extended collection of mammographic cases that includes clinical data as well as images acquired using additional modalities such as DBT, MRI and ultrasound.

Currently we are compiling an appropriate multi-modal database and while our experiments with the case-level CADx approach proposed in this work indicate a significant increase in ROC performance, these results so far are based on a very limited dataset. Merging the existing IRMA mammography database³¹ and the collection of mammographic cases from the University Hospital of Erlangen, approximately 11,000 images have been collected so far.

4. CONCLUSIONS

In this work, an extended computer assisted system (CADx) for improved breast cancer diagnosis based on multiple image views, image modalities and image stacks has been proposed, which operates on the medical case level rather than the individual lesion level. For easy access to the required information and image sources such as mammography, ultrasound, DBT, MRI, we propose an integration of the CADx system with PACS, RIS and HIS using the existing IRMA framework. We show that our concept of multi-image features is also applicable to lesion attributes extracted from multiple DBT slices. The system evaluation will be based on a large ground truth of annotated and histologically validated images integrated from different reference databases. While the proposed system is a natural extension of CADx approaches, like the work of Drukker et al.,⁹ which integrate only two modalities, we are still collecting a large enough database of breast lesions with images from multiple modalities to evaluate the benefits of the proposed approach on.

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