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ProstateAnalyzer: web-based medical application for the management of prostate cancer using multiparametric **MR** imaging

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19 Objectives: In this paper, we present ProstateAnalyzer, a new web-based medical tool for prostate cancer diagnosis. ProstateAnalyzer allows the visualization and analysis of 20 magnetic resonance images (MRI) in a single framework. 21

Methods: ProstateAnalyzer recovers the data from a PACS server and displays all the 22 associated MRI images in the same framework, usually consisting of 3D T2-weighted 23 imaging for anatomy, dynamic contrast-enhanced MRI for perfusion, diffusion-weighted 24 imaging in the form of an apparent diffusion coefficient (ADC) map and MR Spectroscopy. ProstateAnalyzer allows annotating regions of interest in a sequence 25 and propagates them to the others.

26 Results: From a representative case, the results using the four visualization platforms 27 are fully detailed, showing the interaction among them. The tool has been implemented 28 as a Java-based applet application to facilitate the portability of the tool to the different 29 computer architectures and software and allowing the possibility to work remotely via 30 the web.

Conclusion: ProstateAnalyzer enables experts to manage prostate cancer patient data 31 set more efficiently. The tool allows delineating annotations by experts and displays all 32 the required information for use in diagnosis. According to the current European 33 Society of Urogenital Radiology guidelines, it also includes the PI-RADS structured 34

reporting scheme. 35

> Keywords Applications, database management system, magnetic resonance imaging, magnetic resonance spectroscopy, medical informatics, prostate cancer

BACKGROUND 40

Prostate cancer (PCa) has become a significant health care burden (1). Early 41 diagnosis and active follow-up allow improved prognosis and prevent life-42 threatening conditions. Once the decision of treatment is taken, having the 43 most complete set of information for treatment and then for follow up is 44 crucial. Among the techniques used to detect and diagnose PCa, magnetic 45

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resonance imaging (MRI) allows the non-invasive analysis of the anatomy and 62 the metabolism in the entire prostate gland. The prostate is composed of 63 peripheral (PZ), central (CZ), transition (TZ), and anterior fibromuscular 64 tissue (AFT) zones (Figure 1). The PZ represents up to 70% of a normal 65 prostate gland and around 75% of prostate cancers originate in this zone. The 66 CZ represents about 25% of a normal healthy prostate gland in a young adult. 67 Even if the frequency of cancers originating here is much lower, they tend to be 68 of the more aggressive type (2). 69

MRI has been established as the best imaging modality for the detection, 70 localization, and staging of PCa on account of its high resolution and excellent 71 spontaneous contrast of soft tissues and the possibilities of multiplanar and 72 multiparameter scanning (3). As such, three MRI techniques (anatomic 3D 73 T2-weighted, diffusion-weighted, and perfusion-weighted imaging) and 3D 74 MR spectroscopy will be illustrated in this paper. A 3D T2-weighted imaging 75 (T2WI) sequence (4) yields good contrast between PZ and CZ tissues. Diffusion-76 weighted imaging (DWI) provides functional information of tissues such as cell 77 organization, density and microstructure, and depends principally on the 78 Brownian motion of water molecules (5). It can be displayed either as native 79 diffusion-weighted images or as the ADC parametric map. The motion of water 80 molecules is more restricted in tissues with a high cellular density and intact 81 cell membranes and very low values are clearly indicative of cancer (6). 82 Perfusion imaging is based on the dynamic contrast enhancement (DCE) of the 83 signal during the first pass of the contrast agent. The theoretic underpinnings 84 of this vascular technique are based on tumor angiogenesis. In fact, there is a 85 relationship between abnormal perfusion and neoangiogenesis in tumors 86 (7-9). Magnetic resonance spectroscopy (MRS) is a technique that allows the 87 study of metabolite concentrations by means of a 3D chemical shift imaging 88 protocol (10). This study is useful since healthy and cancer tissues show 89 different concentration levels. Specifically, prostate cancer tissues show lower 90 levels of citrate and higher levels of choline compared with healthy tissue 91 (11-15) and metabolic data are often presented in the form of concentration 92 ratios, e.g. [Choline+Creatine]/Citrate. The exact ratio can vary with equip-93 ment and settings. For example, ratios at 3 T differ slightly from those at 1.5 T 94 because of differences in the shape of the citrate spectrum (16). However, it is 95 generally accepted that PZ zone voxels, in which the ratio of choline and 96 creatine to citrate is at least two SDs higher than the average ratio in healthy 97 tissues, are considered to represent possible cancer (17). Voxels are considered 98

highly suggestive of cancer if the ratio of choline and creatine to citrate is morethan three SDs higher than the average ratio (18).

The increasing amount of data available to analyze a study has not been 101 accompanied with the development of a single standardized way to report it. 102 Often the reports have been unstructured and in a narrative way (19). 103 However, the European Society of Urogenital Radiology (ESUR) has recently 104 proposed the Prostate Imaging-Reporting and Data System (PI-RADS) as the 105 standard structured reporting scheme for prostate cancer (10). Although other 106 107 schemes exist, such as the Likert score (20), the use of PI-RADS is rapidly extending (21-24). Therefore, we included in our tool the possibility to use this 108 109 reporting scheme.

110 The medical support systems used to assist the diagnosis of prostate lesions are generally focused on prostate segmentation (25–28). They rely on 111 computerized techniques for prostate cancer detection applied to ultrasound, 112 113 magnetic resonance, and computed tomodensitometric images (29). For example, Vos et al. (30) used 3D-T2 imaging to define specific regions of 114 interest (ROI), which were subsequently used on diffusion- and perfusion-115 weighted images to extract relevant features. The purpose was to train and 116 classify the extracted set of features to calculate the likelihood of malignancy. 117 118 Other related surveys have been focused on magnetic resonance spectroscopic data (31). The rapid growth of computer-based clinical exams has seen a 119 significant increase in the number of online medical imaging systems (32,33), 120 some of which being developed as web-based applications (34). The main 121 challenge is the access and the interaction between the hospital database and 122 the distant users. Thus, the aim of our work was to develop a new architecture 123 system allowing including a web-based application connected to a prostate 124 image database. 125

In the field of biomedical informatics, one of the long standing problems is 126 finding a way to share medical data across a variety of media. Inherently, 127 medical data are generated by a multitude of sources (35,36). eXtensible 128 129 Markup Language (XML) has emerged as a leading facilitator. Although the XML is provided with predefined tags, one of its advantages is its extensible 130 use. Over the last few years, a set of standards in the medical domain has been 131 132 developed, called medical markup language (MML) to allow the exchange of medical data between different medical information providers (37,38). 133 Therefore, the XML schema of tags can be defined for each individual case. 134 Finally, the inclusion of the XML databases facilitates the management of 135 XML files by storing them in an efficient way (39). Traditional object relational 136 techniques, based on an XML model, are used to store XML files in an eXist-db 137 138 database (40).

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140 PRIMARY OBJECTIVES

In this paper, we propose a new medical tool in the form of an interactive JAVA applet application called Prostate-Analyzer. The purpose is to facilitate the inclusion of medical findings on existing prostate images using the combination of MRI techniques and MR Spectroscopy. The novelty of our proposed framework is that it includes the use of MRI and MRS in a compact application. Generally, users have different tools to evaluate prostate images

for MRI and MRS. However, it is necessary to compare regions of interest at the same time in order to determine pathologies or prostate lesions. For this reason, it is more efficient for each modality to provide a single tool that can easily manage all the data.

The main objective of the ProstateAnalyzer system is to develop a new tool that encompasses the visualization and analysis of prostate MR images, as well as a new storage system of clinical diagnoses in a single package. The tool is able to characterize the morphological (location, shape, and size) and the imaging features (mean and standard deviation of the signal, area, and center of mass) of a region within the image, defined and annotated by an expert. Furthermore, it calculates the signal-time curve in perfusion studies and

displays the LCModel signal spectrum for spectroscopic analysis (41). Besides,
it allows the analysis of the same image by different experts, an essential
feature in order to obtain a robust evaluation. The final outcome of the analysis
is summarized in terms of the PI-RADS protocol.

With regard to the proposed architecture for storage of medical images, this should guarantee the connection and a secured access to the server, the DICOM database, and the system storage of clinical diagnoses (XML database).

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168 **RESEARCH DESIGN**

170 Materials

Three-dimensional T2-weighted fast spin-echo (TR/TE/ETL: 3600 ms/143 ms/
109, slice thickness: 1.25 mm) images (T2WI) were acquired with submillimetric pixel resolution in an oblique axial plane. DWI (TR/TE: 4200/
101 ms) was performed using a pulsed gradient spin-echo technique with two *b*-values (100 and 800 s/mm²). A parametric image, the ADC map, was directly
generated (on the vendor's MRI consol) from the raw data on a pixel-by-pixel
basis using the formula:

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$$S = S_0 \exp(1 - bD)$$

where S_0 represents signal intensity when b=0 and D is the ADC directly displayed as a gray scale number expressed in units of 10^{-6} mm²/s.

The protocol was completed with DCE-MRI performed using a fat 182 suppressed 3D T1 VIBE sequence (TR/TE/Flip angle: 3.25 ms/1.12 ms/100; 183 184 temporal resolution: 6 s/slab over approximately 5 min). A bolus injection of 185 Gd-DTPA (Dotarem, Guerbet, Roissy, France) was administered at a dose of 0.2 ml Gd-DTPA/kg of body weight. Finally, three-dimensional MR spectro-186 187 scopic data were acquired using a water and lipid-suppressed sequence. Data sets were composed of $16 \times 16 \times 16$ phase-encoded spectral arrays giving 188 189 nominal spectral resolution of 0.14 cm³. Spectral data were processed using 190 a modified version of LCModel (S. Provencher Inc., Asbestos, QC) for 1H 191 spectroscopy of the prostate at 3T. Choline, creatine, and citrate were all 192 quantified individually.

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194 **ProstateAnalyzer application**

ProstateAnalyzer is a network-based database system whose aims are the management and processing of both MRI and MRS data sets in a single package.



Figure 2. Network-based database system designed to implement the framework and the used hardware. The Zend server allows easy integration of the Apache server with database management tools.

This tool is implemented as a Java-based applet application to facilitate the inclusion of medical findings on existing prostate studies.

220 Architecture

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A system has been designed in order to link the ProstateAnalyzer application to a database of DICOM images and a XML server, which stores the annotated files of the DICOM images. This architecture is supported by a Zend Server in order to allocate the ProstateAnalyzer application and a XML database provided by eXist-db (40). This type of architecture has been used in previous works for supporting other modalities of medical imaging databases such as PACS (42).

The ProstateAnalyzer application (Figure 2) communicates via socket TCP/IP with the Apache Server provided by Zend Server. This server is connected to a DICOM database, containing DICOM prostate images and with the eXist-db database, containing the XML files. We opted for the creation of the XML annotation files in the database in order to preserve the integrity of the original DICOM database (usually stored in a PACS system).

Furthermore, in order to avoid vulnerabilities during the transfer of information, internal security measures have been implemented. Besides using control User ID private keys, the algorithms have systems that protect and encrypt the information during its transmission. Currently, the secure socket layer (SSL) cryptographic protocol is used (43). Finally, the users can download the complete analysis performed by one user in a PDF file such a pre-medical report.

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242 XML database

The ProstateAnalyzer application is related to the eXist-db manager, responsible for the XML database, incorporating its own server for access and management via web interface. We also opted for the creation of the XML



Figure 3. Connections to the eXist XML database within the ProstateAnalyzer and its interaction with the user.

annotation files in the database. Again, all XML files are related to their
 corresponding DICOM images and contain the records provided by experts.

271 Figure 3 shows a scenario of the use of the eXist-db XML database. The 272 database is configured at different levels in order to store XML files associated with studies, collections, status and images. First, the ProstateAnalyzer 273 application connects with the eXist-db to access a study, where each study 274 275 contains a set of XML collections corresponding to 3DT2, DWI (as ADC map), and DCE techniques. Consequently, for each dataset, all the XML files 276 concerning prostate images are allocated. It is important to note that each 277 278 XML file contains the annotations provided by all the experts. In this way, the main structure of the XML file is the same for the images of the same study, 279 thus enabling a faster access to the annotated information. 280

Moreover, at the status level, for each study, a XML file associated with each
 user is saved. Consequently, once the user is logged into the ProstateAnalyzer,
 the corresponding XML file is obtained.

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285 Security and system access

ProstateAnalyzer is designed as a web application; hence the connection between the applet and the server can be both internal and external. The actual login process consists in an authentication measure with a unique username and password. One of the key elements in this application concerns the security of the data and the information across the network. In this sense, internal security measures to avoid vulnerabilities from the transfer of information are already included in the Zend Server.

HTTP was originally designed for the transmission and reproduction of multimedia documents, but these encodings are not part of the HTTP standard. It is up to the applications themselves to break down and reassemble the information in order to transmit and receive it. Fortunately, Java applets are inherently based on the Internet Inter-ORB Protocol (IIOP). This protocol is based on the client/server computing model and a security algorithm is implemented as a security mechanism to ensure the integrity of the medical data being transferred. In the literature, several architectures use the IIOP in their security applications (19,44).

The most widely used regulations with respect to privacy and security are 302 303 the Health Insurance Portability and Accountability Act (HIPAA) and the European Data Protection Directive 95/46/EC. Both regulations mandate 304 health institutions to protect health information against unauthorized use or 305 disclosure (45). In ProstateAnalyzer, the login/password signature allows 306 knowledge of who is dealing with the individual data. Therefore, an audit-log 307 for each study is produced identifying the user who has read the data and 308 309 thereby allowing one to generate audit trails on data access activities for any specific patient (46). 310

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312 ProstateAnalyzer engine

The search engine is the front-end for users once ProstateAnalyzer is accessed. 313 There are two types of patient searches: search for all the patients and an 314 advanced search. In the first option, the search engine returns a list with all 315 the patients recorded in the database. In the second case, the system returns 316 just the patients who share the same regular expression used as a query. To 317 perform a search, it is necessary to create a data set that contains information 318 concerning the four MR modalities described before. For each item, the tool 319 shows different attributes (patient name, comments, and status) and the 320 321 number of images provided by the concerned modality (anatomy, diffusion, perfusion, and spectroscopy). Moreover, each case is represented by a different 322 color depending on the status of the study. When a state is color-coded as 323 yellow, it means that the study has been analyzed partially; a green color-324 325 coding indicates that the study has been completely analyzed and validated. Finally, white represents a study that has not yet been analyzed. 326

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328 Anatomy 3D-T2WI

ProstateAnalyzer offers a new visualization platform to provide annotations 329 for different kinds of tissues (tumor number 1, tumor number 2, PZ, CZ, TZ, or 330 331 others) obtained from T2WI studies. This visualization capability is combined with basic post-processing tasks such as zooming, gamma correction, user-332 specific ROI, surface, and volumetric measurements. In addition, the viewer 333 334 provides information from the DICOM header, the status of the anatomical study (validated, partial, or empty) and the list of the annotations added by 335 336 different users.

Figure 4(a) shows an example of visualization of the T2WI in prostate cancer analysis. When an overlay is manually drawn, a label is automatically defined with the user's specified color. This label is defined using the first letter of the user's first name and surname followed by the number of annotations (in the figure, the first region of interest is identified by letters "AL0" in green). A table shows a list of color-coded overlays, specifying the username and the tissue type (indicated by the user). It also allows us to



COLOR Online / B&W in Print visualize or to delete each overlay, although this latter option is only available

³⁹⁴ to the expert who created it. Selecting an overlay of the table allows displaying

its related information (the annotations filled in a pop-up window). Finally,

396 after pressing the confirmation button of the pop-up window, the overlay is

added to the table, displayed on the image and stored in the XML database.

399 **DWI**

ProstateAnalyzer also provides a support tool for the DWI studies with similar basic post-processing tasks as presented previously for T2WI. Figure 4(b) shows an example of an ADC map with annotations provided by different users. In the example, a multiple user-specific region of interest (ROI) study is presented.

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⁴⁰⁶ Perfusion imaging analysis (DCE)

⁴⁰⁷ Dynamic contrast enhanced (DCE) MRI is also catered for in the
⁴⁰⁸ ProstateAnalyzer. The tool also contains basic post-processing tasks using
⁴⁰⁹ the same features applied in T2WI and DWI. The most important difference
⁴¹⁰ with respect to the other techniques is that the perfusion images are displayed
⁴¹¹ in time-space partitions.

Furthermore, this viewer uses a mean signal-time curve to display the signal enhancement during the arrival of the contrast agent into the tissue. Figure 4(c) shows an example of the ROI analysis in perfusion MR images. For each overlay, the mean signal from each ROI is calculated and displayed as the corresponding signal-time curve: the mean-curve is denoted with a different color according to its associated area.

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419 **MRS**

ProstateAnalyzer offers individual spectrum visualization from a 3D spectro-420 scopic grid. Figure 4(d) presents an example of a MRS study with a LC Model-421 422 processed signal spectrum. This viewer displays a set of 3D-T2 images allowing 423 the spatial location of the spectroscopy study. For each spectroscopic grid, 424 there is a corresponding image. In order to obtain a specific spectrum, the user 425 must click on the corresponding voxel (highlighted in red) within the grid. The LC model is widely used for processing clinical single and multi-voxel 426 427 spectroscopic data (47). It allows individual and batch analyses of the main 428 metabolites within the spectra of the prostate: citrate, choline, creatine, and 429 spermine.

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431 Interaction between spectroscopy and anatomy

432 One of the most important features of ProstateAnalyzer is the interaction of the imaging modalities among themselves. Figure 5 shows an example of the 433 interaction between MRS and T2WI. The aim is to display the annotations, 434 obtained from the T2WI study panel, on the spectrum illustrated within the 435 graphical window. The way to link the processing is provided by the slice 436 location (it is shown in the DICOM information panel of the MRS image, and 437 boarded in yellow). In the MRS study panel, when the expert selects a voxel 438 within the spectroscopic grid, the spectrum is displayed. The next step is to 439 determine whether the selected voxel contains any annotations from the 440 anatomical 3D-T2 panel. If the information exists, it is displayed in the 441



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COLOR Online / B&W in Print 491 spectroscopic graphical area along with the spectrum. The example shows the 492 selected voxel corresponding to an area (in the anatomic analysis) with two 493 denominations for one tissue (named as Tum1 and CG in the figure), and is 494 also represented on the annotation table. This interaction is very important, 495 because one can visualize the information of both techniques in only one 496 representation.

⁴⁹⁷ ₄₉₈ Interaction between anatomy and perfusion

In this case, the anatomic and perfusion studies are not using the same reference image due to the differences in spatial resolution. Consequently, the solution is to locate the image from a set of corresponding perfusion DCE-MRI images, as close as possible, to the slice location of the anatomical 3D-T2 image.

Figure 6 shows an example of the analysis on a perfusion-weighted study 504 using the ROI drawn on the corresponding 3D-T2 image. On the interface of 505 the perfusion image analysis, when the "anatomy" button is pressed all the 506 annotations concerning the anatomy study are displayed. ProstateAnalyzer 507 adapts the annotated regions into the correct position in the perfusion image, 508 according to the spatial resolution and the pixel spacing. In order to 509 distinguish the overlays between perfusion and anatomy, the viewer displays 510 all the anatomy overlays and adds the prefix "A" before the label name. 511

512 **Reporting**

Once the study is analyzed, the findings need to be reported, if possible in a standardized way. ProstateAnalyzer includes the PI-RADS structured reporting scheme, according to ESUR prostate MR guidelines in 2012 (10), which is being established as the common protocol in European countries (21,22). The use of a standardized graphic reporting scheme facilitates the communication with referring colleagues, and it increases the quality and diagnostic value of prostate analyses.

The inclusion of the PI-RADS is carried out using a form with a drop-down list that includes the different answers of the PI-RADS question. Selecting the option, the corresponding number is assigned. The report is saved in an independent XML file and loaded into the XML database. The use of an independent XML file allows a faster search when looking just for similar cases in terms of PI-RADS scores. Besides, there is the option of downloading a PDF file with that information.

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MAIN OUTCOME

In this section, a complete example of use of the ProstateAnalyzer application is presented. In order to display the environment of the application, the obtained results using the four visualizations platform (T2WI, DCE, DWI, and MRS) are fully detailed. The main purpose is to demonstrate the usability of the ProstateAnalyzer application in the localization and the analysis of a tumor, if present.

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537 Clinical diagnosis

The upper image of Figure 7 shows an example of a localized tumor displayed on T2WI. Prostate cancer usually shows low signal intensity on T2WI that is 540



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Figure 7. Interaction between T2WI and DCE.

well defined with respect to normal PZ zone tissue. However, it is not always 618 619 easy to localize a pathologic area. One of the advantages of the ProstateAnalyzer application is to analyze the same study using other 620 techniques. Once a sequence of the T2WI is analyzed, the application offers 621 an efficient solution to compare the same ROI in DCE, DWI, and MRS. On the 622 DCE, the peak of the curve corresponds to the first pass of the contrast agent 623 and indicates the rapid uptake of the contrast agent, typically observed in 624 cancer tissue. 625

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We compare the mean curve of a representative tumor tissue with that of 626 healthy tissue in Figure 8. The first curve (in yellow) is from tumor tissue and 627 628 the second from healthy tissue (in red). In this case, the first annotation corresponds to a tumor region labeled as "PW0" which is made in the perfusion 629 panel. The second, which represents healthy tissue, is annotated in the 630 anatomical panel and displayed in the perfusion panel labeled as "A-PWO". In 631 the diffusion panel, the analysis also demonstrates the presence of a tumor 632 (Figure 9). It is important to notice that the user draw a different ROI as the 633 one in Figure 7. Indeed, the tumor will correspond to low signal (therefore, low 634 diffusion coefficient) on the ACD map. 635

On the MRS panel (Figure 10), the example shows a selected voxel corresponding to an area of cancer tissue (Tum1). The LC Model-processed







Figure 9. Example of a localized tumor is represented using diffusion-weighted imaging of the same prostate study.

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Figure 10. Example of the interaction between T2WI and MRS studies showing the
 spectra of the cancer Voxel.



Figure 11. A window pop-up allows to report the case according to the PI-RADS protocol.

signal spectrum demonstrates how the signal is modeled in red from the 760 noisy signal in black. Hence the different peaks of the spectrum are easily 761 762 extractable. Illustrating graphically the levels of these metabolites is very important, since cancerous tissue is characterized within the spectrum as 763 reduced citrate and elevated choline peaks. In the figure, the first two 764 consecutive peaks correspond to choline and creatine, while the third one 765 corresponds to citrate. It is clear that the relative levels of choline and creatine 766 are very different to those observed in normal healthy tissue (Figure 4d), 767 768 where higher levels of citrate and lower levels choline are observed.

Once the clinical diagnosis has been performed, users have the option to create a MR reporting, as depicted in Figure 11. For each MR technique, a "PI-RADS" button is available in the option panel. When it is pressed, a window pop-up appears where the user can report the findings for each technique according to the PI-RADS protocol.

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775 Design effort benchmarks

ProstateAnalyzer is currently being used in the MR department of the 776 University Hospital of Dijon (France). The present working database is 777 composed of more than 1600 patient datasets. The time to access to the Java 778 Applet application (ProstateAnalyzer) by users is around 2–3s in an intranet 779 environment. It has been tested with the following common internet web 780 browsers: Mozilla Firefox, Internet Explorer, Google Chrome, and Opera. It is 781 preferable to use the Java Deployment Toolkit 6.0.160.1.1 or superior and to 782 use the JavaTMPlatform SE 6 U16 6.0.160.1. The most time-consuming part 783 corresponds to loading a prostate study. A typical study consists of a set of 784

64 anatomical images, 14 ADC images, 640 perfusion images, and a set of spectroscopic data containing up to 1000 files. It is important to note that the number of spectroscopic data files is very variable and depends on the prostate size. The average of the time consuming is around 60–70 s to load and display a typical prostate study with the data presented previously.

The search engine provided by ProstateAnalyzer is tested in order to obtain 790 the complete study. The timing response to retrieve a prostate study was 791 around 1 min including the corresponding XML-associated files. The time 792 necessary to record annotations on the database is around 3–5 s for anatomical 793 and diffusion images and around 12s for perfusion images. The computational 794 cost is higher for the latter because of the need to calculate the signal-time 795 curve. In order to obtain a spectrum, the cost is around 5s to display it in the 796 graphical window. Once the analysis is finished by the user, ProstateAnalyzer 797 takes around 4s to save the XML file into the database. 798

799 With the new architecture, ProstateAnalyzer can also be accessed using regular external network connections outside the intranet connection of the 800 hospital. We tested the connection from Girona in Spain (several hundreds of 801 kilometers from Dijon). Experiments were carried out by loading a large study 802 consisting of a set of 64 anatomical images, 15 ADC images, 720 perfusion 803 804 images, and a set of spectroscopic data containing up to 1200 files (corresponding to around 150 MB of transferred data). We loaded the case several 805 times during a week, and the total elapsed time response varied between 100 806 and 130 s. When comparing the times provided by the internal and external 807 connections using the same study, the difference was found to be around 808 40–70 s. The difference was due to two main reasons. On one hand, it depends 809 on the instantaneous network load and on the other hand to the number of 810 users being connected to the ProstateAnalizer server. The main conflict, when 811 multiple users are connected, is in the downloading of the cases, since the 812 server must split the downloading process between the connected users. 813 However, once the study is loaded, the usability, visualization, and annotation 814 815 times are independent with respect to the number of users – they just depend on their own machine, not on the server). 816

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DISCUSSION

ProstateAnalyzer is a web-based application for the analysis of prostate images using four MR modalities: T2WI, DWI, DCE, and MRS. One of the most evident advantages of ProstateAnalyzer is that it allows simultaneous analysis of a prostate study using the different modalities. It also provides an interaction among them and MR spectroscopy. Since the application has been designed to work with DICOM files, any equipment that operates under such image standard can potentially be used with ProstateAnalyzer.

The principal problem encountered in the diagnosis of a prostate study is the localization of a ROI-containing tumor tissue. Normally, experts use different tools to validate the diagnoses using different software and make many annotations in different files. This is not a practical solution to managing abundant medical data. ProstateAnalyzer offers the solution, because it allows experts to analyze prostate ROI on T2WI, DWI, DCE, and MRS panels within the same application.

The development of the proposed interface has been made in such a way 834 that it is simple and intuitive to use for the users. The interface is divided in 835 four panels with the purpose of visualizing a patient study, simultaneously, for 836 the different techniques. The main objective was to provide useful tools for 837 experts to manage examinations with different types of images and data. 838 ProstateAnalyzer allows the annotation of findings provided by different 839 experts in the prostate. All annotations are saved in XML files associated with 840 the prostate images. Although a prostate study can be shared among all the 841 842 users, they can still validate or modify their own diagnosis in individual cases. Finally, our tool offers the possibility to work remotely via the web and 843 represents an improvement in the data management. The access to the 844 ProstateAnalyzer should be provided to multiple users in order to make easier 845 both local and external connections. In order to facilitate this task, the 846 application is implemented as a JAVA applet tool. Thus, it is not necessary to 847 install any program on computers to run our interface unless users have a 848 browser that supports Java technology. Moreover, applets can be executed 849 from any operating system (Windows, Linux, and MAC) because they are 850 running in a Web browser. ProstateAnalyzer can be integrated in a server to 851 manage medical images stored in a prostate database. 852

ProstateAnalyzer still presents some limitations which will be addressed in 853 order to enrich the tool. For instance, in perfusion analysis, the tool uses a 854 mean DCE signal-time curve to display the signal enhancement during the 855 arrival of the contrast agent into the tissue. However, providing quantitative 856 values related to microvascular permeability, $K^{\rm trans}$ or the diffusion space, $v_{\rm e}$ 857 could also be helpful. With regard to the spectroscopy section, the possibility of 858 displaying 2D and 3D metabolite maps will be addressed. For this task, 859 specialized spatial and spectral data processing methods, for which sources are 860 not commonly available, are needed (e.g. morphological analysis and spectral 861 characteristics of the observed metabolites). We are also trying to improve the 862 efficiency of the computational cost. One solution could be to load single frames 863 only, on the basis of their spatial correspondence, instead of the whole dataset. 864 This would also allow the possibility of automatically displaying the four 865 sequences simultaneously. Thus, when scrolling the slices, the information 866 will change simultaneously on the T2WI, DWI, DCE, and MRSI displays. 867

Taking into account how rapidly clinical databases are growing, 868 ProstateAnalyzer should be an important contribution to the management of 869 870 large databases. Indeed, the implementation of ProstateAnalyzer offers the possibility to extend and adapt to other MR modalities. As a proof of concept, 871 this project is an adaptation from previous work based on mammography (42) 872 873 and it demonstrates that our approach is easily portable to other sources of medical images. Finally, as ProstateAnalyzer allows downloading reports in 874 875 PDF format containing the annotations and graphical results (mean signaltime curve and signal spectra), it should also be useful as a pre-report. 876

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CONCLUSION

In this paper, we have presented ProstateAnalyzer, a new medical tool that allows the evaluation of the prostate cancer in an effective way. ProstateAnalyzer visualizes the different MRI techniques (anatomy, diffusion, and perfusion) together with MR spectroscopy, and
automatically places annotations, made in one of the images, onto the
others. In addition, ProstateAnalyzer also includes the PI-RADS reporting
protocol, thereby offering the possibility to fully report the prostate study in a
standardized way.

It has been implemented as an interactive JAVA applet application with the purpose of facilitating the inclusion of medical findings on existing prostate images, using the combination of MRI techniques and MR spectroscopic analysis. Furthermore, a new architecture is presented to store the medical records in a XML database which stores a set of annotated files.

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895 896 DECLARATION OF INTEREST

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903 REFERENCES

- Schröder FH, Hugosson J, Roobol MJ, et al. Prostate-cancer mortality at 11 years of follow-up. N Engl J Med 2012;366:981–90.
- 2. Verma S, Rajesh A. A clinically relevant approach to imaging prostate cancer: review. Am J Roentgenol 2011;196:S1-10.
- 3. Chen M, Dang HD, Wang JY, et al. Prostate cancer detection: comparison of T2weighted imaging, diffusion-weighted imaging, proton magnetic resonance spectroscopic imaging, and the three techniques combined. Acta Radiol 2008;49:602-10.
- Bitar R, Leung G, Perng R, et al. MR Pulse sequences: what every radiologist wants to know but is afraid to ask. RadioGraphics 2006;26:513–37.
- 5. Kiliçkesmez O, Cimilli T, Inci E, et al. Diffusion-weighted MRI of urinary bladder
 and prostate cancers. Diagn Interv Radiol 2009;15:104–10.
- 6. Ren J, Huan Y, Wang H, et al. Diffusion-weighted imaging in normal prostate and differential diagnosis of prostate diseases. Abdom Imaging 2008;33:724–8.
- 7. Jackson AS, Reinsberg SA, Sohaib SA, et al. Dynamic contrast-enhanced MRI for prostate cancer localization. Br J Radiol 2009;82:148–56.
- 8. Bard RL. Dynamic contrast-enhanced mri atlas of prostate cancer. New Haven:
 Springer; 2009.
- 919
 9. Alonzi R, Padhani A, Allen C. Dynamic contrast enhanced MRI in prostate cancer.
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- Barentsz JO, Richenberg J, Clements R, et al; European Society of Urogenital Radiology. ESUR prostate MR guidelines 2012. Eur Radiol 2012;22:746-57.
- Villeirs GM, Oosterlinck W, Vanherreweghe E, De Meerleer GO. A qualitative approach to combined magnetic resonance imaging and spectroscopy in the diagnosis of prostate cancer. Eur J Radiol 2010;73:352–6.
- P25 12. Créhange G, Parfait S, Liegard M, et al. Tumor volume and metabolism of prostate
 p26 cancer determined by proton magnetic resonance spectroscopic imaging at 3T
 without endorectal coil reveal potential clinical implications in the context of
 radiation oncology. Int J Radiat Oncol Biol Phys 2011;80:1087–94.
- Particulation oncodegy. Int 5 Radiat Oncol Distribut Phys 2011,0011007 94.
 13. Créhange G, Maingon P, Gauthier M, et al. Early choline levels from 3-tesla MR spectroscopy after exclusive radiation therapy in patients with clinically localized prostate cancer are predictive of plasmatic levels of PSA at 1 year. Int J Radiat Oncol Biol Phys 2011;81:407–13.

- 20 C. Mata et al.
- 932 14. Parfait S, Walker PM, Créhange G, et al. Classification of prostate magnetic
 933 resonance spectra using support vector machine. Biomed Signal Process Control
 934 2012;7:499-508.
 15 The C S big in P. Leli P. et al. Process for the limit of the second s
- 15. Testa C, Schiavina R, Lodi R, et al. Prostate cancer: sextant localization with MR imaging, MR spectroscopy, and 11 C-choline PET/CT. Radiology 2007;244: 797-806.
- 937 16. Scheenen TW, Heijmink SW, Roell SA, et al. Three-dimensional proton MR
 938 spectroscopy of human prostate at 3 T without endorectal coil: feasibility. Radiology
 939 2007;245:507-16.
- 17. Verma S, Rajesh A, Fütterer JJ, et al. Prostate MRI and 3D MR Spectroscopy: How We Do It. Am J Roentgenol 2010;194:1414–26.
- 941
 942 18. Jung JA, Coakley FV, Vigneron DB, et al. Prostate depiction at endorectal MR
 943 spectroscopic imaging: investigation of a standardized evaluation system.
 943 Radiology 2004;233:701-8.
- 944
 95
 945
 946
 19. Dickinson L, Ahmed HU, Allen C, et al. Scoring systems used for the interpretation and reporting of multiparametric MRI for prostate cancer detection, localization, and characterization: could standardization lead to improved utilization of imaging within the diagnostic pathway? J Magn Reson Imaging 2013;37:48–58.
- Villers A, Puech P, Mouton D, et al. Dynamic contrast enhanced, pelvic phased array magnetic resonance imaging of localized prostate cancer for predicting tumor volume: correlation with radical prostatectomy findings. J Urol 2006;176:2432–7.
- 950 21. Portalez D, Mozer P, Cornud F, et al. Validation of the European Society of
 951 Urogenital Radiology scoring system for prostate cancer diagnosis on multi 952 parametric magnetic resonance imaging in a cohort of repeat biopsy patients. Eur
 951 Urol 2012;62:986–96.
- 22. Rosenkrantz AB, Kim S, Lim R, et al. Prostate cancer localization using multiparametric MR imaging: comparison of Prostate Imaging Reporting and Data System (PI-RADS) and Likert scales. Radiology 2013;269:482–92.
- 856 23. Roethke MC, Kuru TH, Schultze S, et al. Evaluation of the ESUR PI-RADS scoring
 857 system for multiparametric MRI of the prostate with targeted MR/TRUS fusion958 guided biopsy at 3.0 Tesla. Eur Radiol 2014;24:344-52.
- Reisæter LA, Fütterer JJ, Halvorsen OJ, et al. 1.5-T multiparametric MRI using
 PI-RADS: a region by region analysis to localize the index-tumor of prostate cancer
 in patients undergoing prostatectomy. Acta Radiol 2014..
- 25. Shah V, Turkbey B, Mani H, et al. Decision support system for localizing prostate
 cancer based on multiparametric magnetic resonance imaging. Med Phys 2012;39:
 4093-103.
- 26. Dickinson L, Ahmed HU, Allen C, et al. Magnetic resonance imaging for the detection, localisation, and characterisation of prostate cancer: recommendations from a European consensus meeting. Eur Urol 2011;59:477-494.
- Pradet V, Kurhanewicz J, Cowan JE, et al. Prostate cancer managed with active
 surveillance: role of anatomic MR imaging and MR spectroscopic imaging.
 Radiology 2010;256:176-83.
- 969
 28. Ghose S, Oliver A, Mitra J, et al. A supervised learning framework of statistical
 970
 971
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- 971 29. Ghose S, Oliver A, Martí R, et al. A survey of prostate segmentation methodologies
 972 in ultrasound, magnetic resonance and computed tomography images. Comput
 973 Methods Prog Biomed 2012;108:262–87.
- 974 30. Vos PC, Hambrock T, Barenstz JO, Huisman HJ. Computer-assisted analysis of
 975 peripheral zone prostate lesions using T2-weighted and dynamic contrast enhanced
 976 T1-weighted MRI. Phys Med Biol 2010;55:1719–34.
- 31. Carroll PR, Coakley FV, Kurhanewicz J. Magnetic resonance imaging and spectroscopy of prostate cancer. Rev Urol 2006;8:S4–10.
- 978
 979
 979
 980
 32. Oppelt A. Systems for medical diagnostics: fundamentals, technical solutions and applications for systems applying ionizing radiation, nuclear magnetic resonance and ultrasound. Germany: Publicis Publishing; 2005.

 981
 33. Madabhushi A, Dowling J, Huisman H, Barratt D, eds. Prostate cancer imaging. 982
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- 34. Kurc T, Janies DA, Johnson AD, et al. An XML-based system for synthesis of data from disparate databases. J Am Med Inform Assoc 2006;13:289–301.
- 35. Marcos E, Acuña CJ, Vela B, et al. A database for medical image management.
 Comput Methods Prog Biomed 2007;86:255-69.
- 36. Tahmoush D, Samet H. A new database for medical images and information.
 In: Proceedings of SPIE medical imaging 2007: image processing 2007, San Diego, CA.
- 37. Guo J, Araki K, Takada K, et al. The latest MML (Medical Markup Language)
 version 2.3-XML-based standard for medical data exchange/storage. J Med Syst
 2003;27:357-66.
- 38. Guo J, Takada A, Tanaka K, et al. The development of MML (Medical Markup Language) version 3.0 as a medical document exchange format for HL7 messages. J Med Syst 2004;28:523-33.
- 39. Lindsay, J. XML databases and biomedical informatics. Connecticut: University of Connecticut; 2008.
- 40. Meier W. eXist: an open source native XML database in web, web-services, and database systems. Germany: Springer; 2002:169–83.
- 998 41. Provencher SW. Estimation of metabolite concentrations from localized in vivo
 999 proton NMR spectra. Magn Reson Med 1993;30:672–9.
- 42. Mata C, Oliver A, Torrent A, Martí J. MammoApplet: an interactive Java applet tool for manual annotation in medical imaging. In: IEEE international conference on bioinformatics and bioengineering, 2012, Larnaca, Cyprus: 34–9.
- 43. Sumathi S, Sivanandam SN. Introduction to data mining and its applications; 29 of studies in computational intelligence. Germany: Springer; 2006.
- 44. Papadakis I, Chrissikopoulos V, Polemi D. Secure medical digital libraries. Int J
 Med Inform 2001;64:1–8.
- 45. Zhou Z, Liu BJ. HIPAA compliant auditing system for medical images. Comput Med Imaging Graph 2005;29:235-41.
- 46. Lee WB, Lee CD, Ho KI. A HIPAA-compliant key management scheme with revocation of authorization. Comput Methods Prog Biomed 2014;113:809–14.
- 47. Provencher W. Automatic quantitation of localized in vivo 1H spectra with
 LCModel. NMR Biomed 2001;14:260-4.
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- 1013 1014