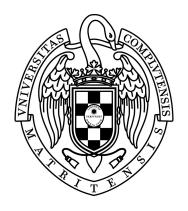
Desarrollo de herramientas bioinformáticas aplicadas a proteómica shotgun y proteómica dirigida



TESIS DOCTORAL

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Desarrollo de herramientas bioinformáticas aplicadas a proteómica shotgun y proteómica dirigida

Memoria que presenta para optar al título de Doctor Vital Vialás Fernandez

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Resumen

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Índice

Re	esumen	V
Ι	Introducción	1
In	troducción	3
	Proteómica. Conceptos generales	3
	Espectrometría de masas	4
	Proteómica en gel	5
	Huella Peptídica	6
	Proteómica shotgun	6
	Separación multidimensional de péptidos	6
	Espectrometría de masas en Tandem. MS/MS	7
	Asignación Péptido-Espectro	7
	Búsqueda basada en secuencia	7
	Búsqueda basada en bibliotecas de espectros	7
	Identificación por secuenciación de novo	7
	Inferencia de proteínas a partir de péptidos	7
	Evaluación estadística de los resultados	7
	Proteomica dirigida. SRM/MRM	7
	Candida albicans como organismo modelo	7
	Repositorios publicos de proteómica shotgun y dirigida	7
II te	Desarrollo de una aplicacion web para datos de pro- ómica shotgun de $Candida\ albicans$	9
1.	Proteopathogen, a protein database for studying $Candida$ $albicans$ - host interaction	11
2.	Proteopathogen 2, adaptación al formato estándar de identificaciones .mzIdentML	19

Índice	ÍNDIO
2.1	
III Creación de un Peptidez	Atlas de Candida albicans
3. A Candida albicans Peptide.	f At las
IV Desarrollo de una base teómica Dirigida (MRM)	de datos para datos de Pro-
Bibliografía	,

Índice de figuras

1. La complejidad del proteoma aumenta exponencialmente y bla

Índice de Tablas



Introducción

Nada en Biología tiene sentido si no es bajo la luz de la Evolución Theodosius Dobzhansky

Tradicionalmente, el gen se ha concebido como la unidad fundamental -el átomo- de la vida, sometida a la acción de la selección natural. Así definió Richard Dawkins en el Gen Egoísta al gen, la unidad indivisible auto-replicante, mientras que los individuos y sus conductas eran meras máquinas de supervivencia. Sin embargo, es el fenotipo y no el genotipo lo que interactúa con el ambiente y con otros organismos. Las proteínas, los ladrillos con que se construye la vida, sí son visibles, a diferencia de los genes, a la selección natural. Por otra parte, el clásico dogma central de la Biología Molecular caducó hace ya tiempo y hoy lo recordamos, más bien, como una sobresimplificación. Actualmente, el emergente campo de la Proteogenómica da cuenta de la intricada red de procesos regulatorios de transferencia de información entre el gen y la proteína. La Proteómica por su parte, le debe a la Genómica el reconocimiento y agradecimiento de haber abierto camino en la Biotecnología moderna. La Bioinformática, en este panorama, tiene un papel integrador. Al igual que la Proteómica, se sirve de diferentes tecnologías que avanzan y se retroalimentan sinérgicamente. Así la Proteómica se beneficia de los avances en Espectrometría de Masas, y estos instrumentos progresan en función de la demanda en investigación. De la misma manera, la Proteómica Computacional, la parte de la Bioinformática mas cercana a la Proteómica, evoluciona para facilitar el análisis de los datos que los investigadores requieren, pero también se beneficia de la incesante, creciente capacidad de procesamiento en las computadoras actuales.

Proteómica. Conceptos generales

El concepto de Proteoma, acuñado originalmente por Marc Wilkins en 1994, ha evolucionado, desde esta definición inicial entendida como el conjunto total de proteínas expresadas por los genes de una célula, tejido u organismo, hasta definiciones más específicas que dependen del momento,

orgánulo celular y otros niveles adicionales de complejidad como por ejemplo modificaciones post-traduccionales. Así, podemos hablar del proteoma (o fosfo-proteoma) de un orgánulo celular, como la mitocondria, en un tejido concreto, en unas condiciones ambientales definidas por los nutrientes disponibles, posiblemente sometida a condiciones de estrés, etc... Es importante, además, tener en cuenta que el grado de complejidad biológica desde la unidad de información, es decir, el gen, hasta la unidad funcional, la proteína, aumenta exponencialmente.

Espectrometría de masas

El desarrollo de las técnicas de ionización suave de macromoléculas biológicas a finales de los años 80, además de valer el Nobel a John Fenn y Koichi Tanaka, permitió sentar las bases de la Espectrometría de Masas aplicada a la Proteómica. Las técnicas de ionización ESI (Ionización por ElectroSpray) (Fenn et al., 1989) y SLD (Desorción Suave por Láser) (Tanaka et al., 1988) permitieron que las grandes y frágiles moléculas biológicas como las proteínas pudieran ser ionizadas y volatilizadas para ser posteriormente introducidas en los espectrómetros de masas.

Como ocurre en muchas otras ocasiones en la ciencia, de forma paralela e independientemente habían surgido en distintas partes del mundo ideas muy similares. Así, el desarrollo de SLD que valió el Nobel a K. Tanaka, tuvo un precedente unos años antes. Franz Hillenkamp y Michael Karas en Frankfurt, Alemania (éstos discutiblemente no galardonados) habían ideado una técnica similar que, en este caso, denominaron MALDI (Desorción/Ionización Láser Asistida por Matriz) (Karas y Hillenkamp, 1988) Aunque MALDI no fue aplicado a la ionización de proteínas hasta la publicación del trabajo de Tanaka, actualmente éste es el acrónimo que se ha impuesto y es, de hecho, una técnica muy extendida en laboratorios de espectrometría de masas.

Un espectrómetro de masas es, en esencia, una balanza molecular capaz de medir, hasta un determinado límite de sensiblidad, la masa (en relación a la carga) de moléculas (ionizadas). Consta básicamente de tres partes o secciones:

- 1. Fuente de iones Las macromoléculas biológicas, como proteínas y péptidos, no son volátiles
 - En ESI, el analito se encuentra en fase líquida en un solvente orgánico volátil como metanol o acetonitrilo. Esta solución es conducida a través de un capilar sometido a un campo eléctrico de forma que las micro-gotas en el ápice del capilar, una vez que la carga supera un límite, adquieren una forma cónica y forman un aerosol. Se produce entonces la desolvatación por evaporación del

solvente. Así, las micro-gotas del aerosol disminuyen su tamaño, reagrupándose en gotas más estables y pequeñas en un proceso reiterativo, hasta el punto en que las moléculas de analito se repelen con la fuerza suficiente para superar la tensión superficial y liberarse del solvente (explosión de Coulomb) quedando iones de analito en suspensión que son introducidos en un sistema vacío hacia el espectrómetro.

- MALDI consiste en embeber la muestra en una matriz líquida con alta capacidad de absorber luz UV sobre la que inciden pulsos de luz láser, lo que permite que las moléculas sean ionizadas y volatilizadas.
- 2. Analizador de masas
- 3. Detector

Proteómica en gel

La separación bidimensional de proteínas en geles de poli-acrilamida (PAGE, Polyacrilamide Gel Electrophoresis) consiste en separar proteinas presentes en una muestra inicial en base a sus propiedades fisico-químicas diferenciadoras como su tamaño y/o su punto isoeléctrico. En función del número de estas propiedades que se aprovechan para separar, en mayor o menor grado, las proteinas de una muestra se distinguen básicamente dos tipos de PAGE

- Geles monodimensionales, 1D-PAGE. En este tipo de geles las proteinas se separan en función de su peso molecular. La electroforesis hace que las proteínas mas pequeñas, de menor peso molecular, se desplacen mas lejos en el gel, sometido a una diferencia de potencial.
- Geles bidimensionales. 2D-PAGE. En este caso, las proteínas se separan en una primera dimensión en función de su punto isoeléctrico. Las proteinas se desplazan sobre una tira con un gradiente de pH hasta situarse en un punto donde su carga neta se equilibra con la de su entorno. A continuación la tira se coloca en la cabecera de un gel y se aplica la segunda dimensión, de modo que se las proteínas se separan más, en este caso por peso molecular, al igual que en un gel 1D.

La proteómica en gel ha sido (y continúa siendo) una técnica muy empleada en laboratorios de todo el mundo. Tiene algunas limitaciones, como el hecho de que proteínas de bajo peso molecular no son fácilmente observables, o que el número de proteínas identificables a partir de un gel difícilmente pueda superar el millar. Pero sin embargo, este tipo de estudios sigue teniendo un nicho en la Proteómica actual (Roepstorff)

Huella Peptídica

La Huella Peptídica de una proteína se refiere al hecho de que el patrón de fragmentación de una proteína en sus péptidos constituyentes utilizando una enzima proteolítica determinada, es muy específico de la proteína original y el espectro que generan puede ser utilizado para identificarla. Sin embargo existe una altísima variedad de posibles péptidos generados a partir de las proteínas y pueden tener masas muy similares. Por ese motivo, esta técnica requiere que la proteína se encuentre previamente aislada, generalmente a partir de una mancha o spot proteico de 2D-PAGE

Digestión de proteínas

El primer paso, esencial, como se explicará también en la proteómica de alto rendimiento o shotgun, es la digestión o rotura de proteínas en péptidos por acción de una enzima proteolítica. Tradicionalmente se ha utilizado para esto tripsina, que rompe la secuencia aminoacídica a continuación, en el lado carboxilo-, de Arginina(R) o Lisina (K) a menos que exista una Prolina (P) adyacente. Los péptidos generados por acción de la tripsina, llamados péptidos trípticos, tienen un tamaño adecuado, dada la frecuencia media de R y K, para el análisis por espectrometría de masas lo que explica la popularidad de esta proteasa.

Proteómica shotgun

La proteómica shotgun es la técnica de elección para la mayoría de estudios proteómicos a gran escala. El nombre shotgun proviene de una analogía con las técnicas clásicas de secuenciación genómica donde el ADN es fragmentado en secuencias más pequeñas que posteriormente son ensambladas. En la proteómica shotgun las proteínas son fragmentadas en péptidos a partir de los cuales se infiere finalmente la proteína original. Implica varios pasos que se detallan a continuación

Separación multidimensional de péptidos

A diferencia de la técnica de la Huella Peptídica donde cada proteína se encuentra (en mayor o menor grado) aislada, en *shotgun*, se parte de una mezcla más compleja de proteínas. Este hecho es importante porque, si contamos con que a partir de cada proteína presente en la muestra, se generan múltiples péptidos trípticos, el grado de complejidad de la muestra aumenta enormemente tras la digestión. Por este motivo, para evitar que la mezcla de péptidos sea demasiado compleja (para la resolución en el análisis MS), opcionalmente se puede realizar una separación previa de las proteinas que reduzca el grado de complejidad. Esta separación puede realizarse de varias

maneras, por ejemplo la separación puede compenzar a nivel de proteína por electroforesis en un gel 1D-PAGE, por fraccionamientos sub-celulares correspondientes a distintos orgánulos, etc... Pero además de este separación

a nivel de proteína se hace también un fraccionamiento a nivel de péptido en este caso, generalmente, por cromatografía líquida. Al igual que en las técnicas basadas en gel, en cromatografía se habla de dimensiones de separación en función de las propiedades fisico-químicas que se aprovechan para el fraccionamiento. Además frecuentemente son combinables y pueden emplearse secuencialmente. Algunas de los tipos de LC más utilizados se describen a continuación. Cromatografía líquida en fase reversa (RP-LC): Cromatografía de intercambio catiónico:

(o alternativamente colocando los péptidos en solución en gotas sobre una placa de MALDI.)

Espectrometría de masas en Tandem. MS/MS

Asignación Péptido-Espectro

Búsqueda basada en secuencia

Búsqueda basada en bibliotecas de espectros

Identificación por secuenciación de novo

Inferencia de proteínas a partir de péptidos

Evaluación estadística de los resultados

Proteomica dirigida. SRM/MRM

Candida albicans como organismo modelo

Repositorios publicos de proteómica shotgun y dirigida

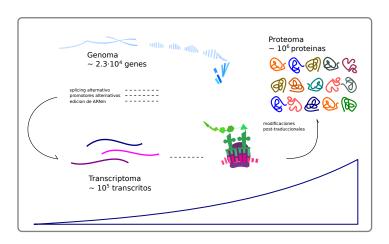


Figura 1: La complejidad del proteoma aumenta exponencialmente y bla

Desarrollo de una aplicacion web para datos de proteómica shotgun de *Candida albicans*

Capítulo 1

Proteopathogen, a protein database for studying *Candida* albicans - host interaction

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TECHNICAL BRIEF

Proteopathogen, a protein database for studying Candida albicans – host interaction

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There exist, at present, public web repositories for management and storage of proteomic data and also fungi-specific databases. None of them, however, is focused to the specific research area of fungal pathogens and their interactions with the host, and contains proteomics experimental data. In this context, we present Proteopathogen, a database intended to compile proteomics experimental data and to facilitate storage and access to a range of data which spans proteomics workflows from description of the experimental approaches leading to sample preparation to MS settings and peptides supporting protein identification. Proteopathogen is currently focused on *Candida albicans* and its interaction with macrophages; however, data from experiments concerning different pathogenic fungi species and other mammalian cells may also be found suitable for inclusion into the database. Proteopathogen is publicly available at http://proteopathogen.dacya.ucm.es

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Candida albicans / Database / Host / MS / Microbiology / Pathogen

Candida albicans is an opportunistic pathogenic fungus, which can be found as a component of the usual flora in human mucoses. Although it does not normally cause disease in immunocompetent colonized hosts, in the case of immunosuppressed patients Candida cells can overproliferate and become pathogenic. Cells in yeast form (oval cells) may produce hyphae, penetrate tissues and eventually cause invasive candidiasis. At present, the frequency of this fatal opportunistic mycosis continues to be distressing and, unfortunately, solution is hindered by the reduced effectiveness and serious side effects of the few available drugs,

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Abbreviations: CGD, Candida Genome Database; GO, Gene Ontology; KEGG, Kyoto Encyclopedia of Genes and Genomes; PDB. Protein Data Bank

the appearance of antifungal-drug resistance, and the lack of accurate and prompt diagnostic procedures [1].

Addressing proteomic studies involving the way *Candida* interacts with immune cells is thus essential in order to improve our comprehension of the process of infection and represents the primary step of investigation that could lead to future development of diagnosis methods, vaccines and antifungal drugs [2–5].

Experimental techniques in proteomics have quickly evolved in such a way that nowadays we have to deal with vast amounts of complex data originated by the combination of multi-dimensional separation techniques and MS analysis together with the bioinformatics software reports [6]. Existing public repositories for management and storage of proteomic data such as World 2-D PAGE [7], the Proteome Database System for Microbial Research 2-D PAGE [8], or PRIDE [9]; and fungi-specific databases such as BioBase MycoPathPD [10], Candida Genome Database (CGD) [11] or Candida DB [12] are very popular and useful tools. However, none of them deals with proteomic experimental



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Proteomics 2009, 9, 4664–4668 4665

data related to the specific research area of fungal pathogens and their interaction with the host. In this context, we present Proteopathogen, a protein database, currently focused on the *C. albicans* – macrophage interaction model – which enables a framework for the access and submission of proteomic workflow data, from description of the experimental approaches leading to sample preparation to MS settings and identification-supporting peptides. Through its interface web site, the database can easily be queried to allow an efficient browsing through all the stored data, improving the quality of eventual analysis of MS results

Regarding the compilation of information used to populate the database, data from three different studies were considered suitable to be present in Proteopathogen. The first two correspond to publish works relating to proteomics of the *Candida* – macrophage interaction [2, 3], where the former reports 66 different *C. albicans* identified proteins and the latter, 38 murine macrophage proteins. The third study represents an analysis of the *C. albicans* plasma membrane proteome [13]. It compiles a set of experiments aimed at extraction and identification of membrane proteins and a set of experiments intended to obtain enrichment in glycosylphosphatidylinositol-anchored surface proteins, which have been reported to be involved in cell wall biogenesis, cell–cell adhesion and interaction with the host [14].

In all cases, protein identifications lists are collected together with the pertinent experimental context specified by descriptions of the experimental approaches, MS settings and peptides supporting identification for each of the proteins (Table 1).

Along with the experimental information, and in order to provide a deeper view of the data, complementary information is retrieved from public web repositories. In the case of *C. albicans* proteins, identifiers, synonyms, aminoacid sequence of the translated open reading frame, *Saccharomyces cerevisiae* orthologs, *Gene Ontology* (GO) annotation, pathway annotations and scientific literature references were obtained from CGD [11], whereas in the case of murine macrophage proteins, the equivalent information was obtained from UniProt KnowledgeBase [15] and the Mouse Genome Database [16]. Additionally, pathways annotations were retrieved from Kyoto Encyclopedia of Genes and Genomes (KEGG)

Pathway Database [17] and structure information from the Protein Data Bank (PDB) [18].

Concerning the architecture of the software, the back-end layer consists of a MySQL database managed by the web application development framework Ruby on Rails that sets up structure and relations of data, handles queries to the database and displays the user web-based interface.

The experimental context is addressed in Proteopathogen in a hierarchical manner, where a main general approach, which may correspond to a published article, is characterized by a description or title, authors, target species and Pubmed identifier when available; and experiments within it, are in they turn, characterized by the description of the particular experiment, the date when it was performed and number of identified proteins.

Information on one particular protein is split into several sections in Proteopathogen. Protein Basic Information displays the UniProt accession number, description, species, evidence for the existence, standard gene name, organismspecific database identifiers, yeast orthologs for Candida proteins and human orthologs for mouse proteins and sequence. The Section 2 lists experiments in which the particular protein has been identified. Where available, one or more of the following sections will be displayed as well: the table entitled GO showing GO annotations along with the pertinent scientific references, the KEGG Pathways and CGD Pathways tables rendering annotations from KEGG and CGD respectively, and PDB, a table specifying structural information. Where no PDB identifiers are found for C. albicans proteins, S. cerevisiae orthologs are used instead, and similarly, when a PDB identifier cannot be found for mouse proteins, the human ortholog

In all cases, proteins are unambiguously related to their corresponding experiment, thus enabling a relation to the data concerning experimental parameters of identification and identification-supporting peptides. This data comprise, on the one hand, common MS settings for all proteins identified in the particular experiment, including search database, MS type, analysis software, digestion enzyme, fixed aminoacid modifications, variable modifications and maximum allowed number of miscleavages; and on the other hand, particular parameters and peptides list for each protein, including number of matched peptides, score,

Table 1. Overview of the stored data in Proteopathogen as well as their published evidences

References	Description of experimental approach	Species	#Protein identifications
[2]	C. albicans differentially expressed proteins after 3 h interaction with RAW 264.7 murine macrophages. 2-D silver-stained gel. MS/MS (MALDI/TOF-TOF)	C. albicans	66
[3]	Proteins identified from cytoplasmic extracts of RAW 264.7 cells after 45 min interaction with <i>C. albicans</i>	Mus. musculus	38
[13]	Identification of Glycosyl phosphatidil inositol (GPI)-anchored membrane proteins Identification of membrane proteins	C. albicans	292 1273

4666 V. Vialás et al.

Proteomics 2009, 9, 4664-4668

observed peptide mass, calculated peptide mass, start and end coordinates, number of missed cleavages and the sequence of the peptide.

The web interface to Proteopathogen offers multiple ways to query the database. Through the *Browse Experiments* search option, a list containing all sets of experimental approaches is displayed. In its turn, one particular experiment can be browsed through all the proteins identified in it.

The Search form may be used in different manners. Queries for one particular protein can be performed by supplying one of the multiple supported identifiers, namely standard gene names, Candida feature name, Candida DB identifiers, CGD identifiers, MGI identifiers and UniProt accession numbers. Free text queries can be performed as

well, which will retrieve a list of proteins showing coincidences in the description field of the Proteopathogen protein entry. As an additional feature, peptide sequences can also be searched for retrieving in this case, proteins in any experiment having the searched sequence in any of the identification-supporting peptides. Wild characters ("*") and Boolean operators are supported for free text queries and for peptide sequence queries.

In order to enhance interactivity and collaboration with users, a submission form is included in the web interface to allow the upload of more proteomic experimental approaches as long as they concern the topics addressed in Proteopathogen. Sequential steps request from the user the following information: a description of the experimental context, a related protein list, MS parameters

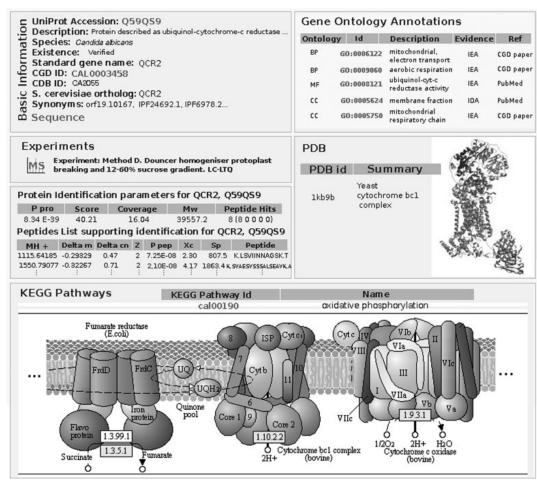


Figure 1. Use case: Search for *C. albicans* ubiquinol-cytochrome-*c* reductase QCR2. The different sections in the result comprise information on protein description and identifiers, experiments in which it has been identified, GO annotation, KEGG and CGD pathway annotation and structural information from PDB.

Proteomics 2009, 9, 4664–4668 4667

and identification-supporting peptides lists. These data are subject to revision prior to eventual insertion into Proteopathogen by the database curators. Besides, the whole relational database and the MS data reports are available for download at the web site.

All the information that is retrievable from Proteopathogen when queried for one particular protein is shown in Fig. 1 for the specific case of ubiquinol-cytochrome-*c* reductase QCR2 of *C. albicans* which has been reported to show antigenic properties in human [19].

The *Protein Basic Information* section displays the Uniprot accession number, a brief description of the protein as stated at CGD, evidence for its existence, standard gene name, feature name, CGD and Candida Database identifiers, yeast ortholog gene name, synonyms and sequence.

The Section 2 lists all the experiments in which QCR2 has been identified. All of them belong to the same general approach aimed at purification of membrane proteins. In every case, the corresponding links to the MS identification parameters and supporting peptides are displayed as well. This experimental data are shown in Fig. 1 for identification of QCR2 in the experiment described as "Method D. Douncer homogenizer protoplast breaking and 12–60% sucrose gradient. LC-LTQ".

The section entitled *GO annotations* shows terms related to the electron transport chain, but more interestingly, it also shows an *inferred from direct assay* (IDA) annotation to the term *membrane fraction* [20], which fits to the fact that the protein is identified in five of the methods aimed at purification of membrane proteins.

KEGG Pathways table provides a link to the KEGG Pathway entry for Oxidative phosphorylation, and provides the feature to show in place the image corresponding to the map from KEGG. CGD Pathways displays an analogous link to the pathway entry at CGD that, in this case, is named aerobic respiration (cyanide sensitive)—electron donors.

Finally, in the *PDB* section, there are four structure images available along with links to the *PDB* entries, corresponding to a cytochrome bc1 complex from *S. cerevisiae*. Orthologs were used since no structure could be found for the *Candida* protein.

In conclusion, Proteopathogen represents, up to date, the first public web-based repository for proteomics data related to studies involving *C. albicans* pathogenicity and its interaction with immune system cells in the host. Moreover, it enables a framework for public access and submission of this type of data and it is intended to be more actively populated in the near future, including data from different pathogenic fungi and mammalian cells, becoming a reference database in its field. Unlike other protein identification databases, Proteopathogen is focused to a specific topic but, at the same time, includes a wide range of data including descriptions of the experimental contexts, information on proteins such as GO and pathway annotations, structural information and detailed MS parameters. Therefore, Proteopathogen will contribute to save time and facilitate

analysis of proteomic workflow reports for researchers interested in this area.

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The authors have declared no conflict of interest.

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4668 V. Vialás et al.

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Capítulo 2

Proteopathogen 2, adaptación al formato estándar de identificaciones .mzIdentML

2.1.

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Creación de un PeptideAtlas de *Candida albicans*

Capítulo 3

$\begin{array}{ccc} A & Candida & albicans \\ \textbf{PeptideAtlas} \end{array}$

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A Candida albicans PeptideAtlas

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ABSTRACT

Candida albicans public proteomic datasets, though growing steadily in the last few years, still have a very limited presence in online repositories. We report here the creation of a C. albicans PeptideAtlas comprising near 22,000 distinct peptides at a 0.24% False Discovery Rate (FDR) that account for over 2500 canonical proteins at a 1.2% FDR. Based on data from 16 experiments, we attained coverage of 41% of the C. albicans open reading frame sequences (ORFs) in the database used for the searches. This PeptideAtlas provides several useful features, including comprehensive protein and peptide-centered search capabilities and visualization tools that establish a solid basis for the study of basic biological mechanisms key to virulence and pathogenesis such as dimorphism, adherence, and apoptosis. Further, it is a valuable resource for the selection of candidate proteotypic peptides for targeted proteomic experiments via Selected Reaction Monitoring (SRM) or SWATH-MS.

Biological significance

This *C. albicans* PeptideAtlas resolves the previous absence of fungal pathogens in the PeptideAtlas project. It represents the most extensive characterization of the proteome of this fungus that exists up to the current date, including evidence for *uncharacterized* ORFs. Through its web interface, PeptideAtlas supports the study of interesting proteins related to basic biological mechanisms key to virulence such as apoptosis, dimorphism and adherence. It also provides a valuable resource to select candidate proteotypic peptides for future (SRM) targeted proteomic experiments.

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Abbreviations: SRM, Selected Reaction Monitoring; CGD, Candida Genome Database; FDR, False Discovery Rate; PSM, Peptide–Spectrum Match; PRIDE, Protein Identifications Database; PSS, Predicted Suitability Score; ESS, Empirical Suitability Score

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1. Introduction

Candida albicans is a fungus of great clinical importance. In addition to asymptomatically colonizing mucous membranes as a commensal in a large percentage of the population, it may cause severe opportunistic infections in specific cases such as patients with weakened immune defenses, a common circumstance in cancer and AIDS patients. C. albicans infections are also a threat to patients in post-surgical situations and intensive care unit stays. In this respect, invasive candidiasis remains nowadays one of the major types of nosocomial infections and a challenge in terms of economical and health costs [1–3]. From the perspective of proteomics, recent studies have provided new insights into the C. albicans biology and suggested new clinical biomarker candidates for diagnosis and prognosis of invasive candidiasis [4–7].

However, the clinical relevance of this organism is not reflected in the number of large-scale publicly available proteomics resources. Up to the current date, the PRIDE [8] database includes only 15 experiments accounting for 1786 identified proteins. The more C. albicans-focused Proteopathogen database [9] comprises several hundred protein identifications including data from gel based proteomics, and other major proteomic online resources such as the Global Proteome Machine Database (GPMDB [10]) or Tranche [11] contain no C. albicans data whatsoever.

As for the genomic data, according to Candida Genome Database (CGD), currently the most comprehensively annotated C. albicans sequence repository [12], the C. albicans genome contains 6215 ORFs (as of May 28, 2013), out of which 1497 are annotated as verified, i.e. representing genes for which there is empirical evidence that the ORF actually encodes a functionally characterized protein. In contrast, 4566 ORFs are termed uncharacterized, indicating that there exists no conclusive evidence for the existence of a protein product. This data implies that most part of the predicted proteome, over 70% of the ORFs, is still unknown or has not been properly annotated yet. An extensive characterization of the C. albicans proteome will therefore be of great value to increase our knowledge in proteins involved in mechanisms of virulence and infection and, thus

serves as a basis to design strategies for diagnosis, vaccination and treatment of invasive candidiasis.

Since its inception, the PeptideAtlas project [13] has encouraged mass spectrometry data submission by the community and has thus grown to a large compilation of atlases of different species including human tissue and body fluid specific builds (brain, plasma [14] and urine), microbial builds (Halobacterium [15], Mycobacterium tuberculosis [16], Streptococcus [17], Leptospira, Plasmodium [18], Saccharomyces [19] and Schizosaccharomyces [20]); invertebrate builds (Caenorhabditis elegans, Drosophila [21] and Apis mellifera [22]); and a pig and a bovine milk [23] builds. The PeptideAtlas project, as a multi-species compendium of proteomes, is continuously increasing its biological diversity. The recent Schizosaccharomyces pombe atlas [23] attains a large coverage of its proteome by ad hoc extensive fractionation and high-resolution LC-MS/MS, and contributes in the sense that some of the fission yeast biological processes have a high degree of conservation with the corresponding pathways in mammalian cells. The incorporation of C. albicans resolves the previous absence of fungal pathogens in the PeptideAtlas and their under representation in any public proteomic data repository.

Furthermore, the proven utility of PeptideAtlas as a resource for selecting proteotypic peptides for Selected Reaction Monitoring (SRM) [24] or SWATH-MS [25] will enable a starting point for future targeted proteomics workflows in *C. albicans*.

2. Material and methods

2.1. Empirical data compilation

Large amounts of mass spectrometry data corresponding to many and diverse measurements of the *C. albicans* proteome initially intended for different purposes were assembled in order to build the PeptideAtlas. A range of proteomic methods, protocols and different biological conditions were used to generate the data as shown in Table 1. These include membrane protein extractions [26], morphological yeast to hypha transition experiments [27] and phosphoprotein enrichment treatments. The combination of these diverse datasets resulted in an

# experiment	Sample (as named in the web interface)	Labeling/treatment	Instrument type	# raw files
1	Calb_acidic_subproteome	-	LTQ	3
2	Calb_memb	-	LTQ	8
3	SILAC_phos_OrbitrapVelos_1	SILAC. IMAC + TiO2	Orbitrap Velos	3
4	SILAC_phos_OrbitrapVelos_2	SILAC. IMAC + TiO2	Orbitrap Velos	3
5	SILAC_phos_OrbitrapVelos_3	SILAC. IMAC + TiO2	Orbitrap Velos	3
6	SILAC_phos_OrbitrapVelos_4	SILAC. IMAC + TiO2	Orbitrap Velos	3
7	SILAC_phos_OrbitrapXL_1A	SILAC. IMAC	Orbitrap XL	11
8	SILAC_phos_OrbitrapXL_1A_TiO2	SILAC. IMAC + TiO2	Orbitrap XL	5
9	SILAC_phos_OrbitrapXL_1B	SILAC. IMAC	Orbitrap XL	6
10	SILAC_phos_OrbitrapXL_1B_TiO2	SILAC. IMAC + TiO2	Orbitrap XL	6
11	SILAC_phos_OrbitrapXL_2	SILAC. IMAC	Orbitrap XL	6
12	SILAC_phos_OrbitrapXL_3	SILAC. IMAC	Orbitrap XL	6
13	SILAC_phos_OrbitrapXL_4	SILAC. IMAC	Orbitrap XL	5
14	Calb_extract_3TOF	-	Triple TOF	2
15	Hyphal_extract_OrbitrapVelos	-	Orbitrap Velos	4
16	Yeast_extract_OrbitrapVelos	_	Orbitrap Velos	4

unprecedented overall coverage of the *C. albicans* proteome. Protein samples were obtained as previously described in [27]. Briefly, cells of the clinical isolate SC5314 were grown in YPD medium for standard growth, whereas hyphal form growth was induced using either Lee medium pH 6.7 or heat-inactivated fetal bovine serum. Protein extracts were then obtained by mechanical cell disruption using either glass beads in the MSK cell homogenizer or the Fast-Prep cell breaker. Protein digests were obtained by trypsinization and separated via HPLC. All spectra acquisition runs were performed by LC–MS/MS in a data-dependent manner in different instruments and setups. Table 1 provides an overview of the experiments along with the instruments used for the mass spectrometry and the corresponding number of raw spectra data files that were acquired.

In addition, raw MS data from unpublished, SILAC labeled and phosphoprotein enriched samples generated from studies focused on *Candida* interaction with host immune cells and from experiments studying the hyphal and yeast-form proteomes, were added to the collection.

2.2. Peptide and protein identification

PeptideAtlas ensures consistency and quality of the stored data by processing the raw spectra sets by the Trans-Proteomic Pipeline (TPP) [28], a suite of software tools for processing shotgun proteomic datasets. The TPP tools are run in a wellestablished sequential pipeline spanning steps from creating appropriate standard files to be used as input by the search engine to statistical validation of protein inference and calculation of the False Discovery Rate (FDR).

The collected raw spectra files in different proprietary file formats were converted to the standard format for mass spectrometry output data mzML [29], searched using X!Tandem [30] with the K-score algorithm plug-in [31] and the output search results were converted to the search engine-independent pepXML format [32].

The target fasta sequence file used for the search was obtained from the Candida Genome Database (CGD) [12] at: http://www.candidagenome.org/download/sequence/C_albicans_SC5314/Assembly21/.

Common contaminants from the common Repository of Adventitious Proteins (cRAP) were appended. Then for each of these sequences, counterpart reversed decoy sequences were appended.

PeptideProphet [33] was then run on the search results to model the distributions of correctly and incorrectly assigned Peptide-to-Spectrum Matches (PSMs). It then assigns probabilities of being correct for each PSM, yielding a sensitive and flexible approach to report results in a comparable manner. Next, iProphet [34] was used to combine additional sources of evidence including multiple identifications of the same peptide across spectra, experiments, and charge and modification states, allowing a more precise integration of evidence supporting the identification of each unique peptide sequence. ProteinProphet [35] was then run to refine iProphet probabilities by adding the information at the protein level, like the number of sibling peptides within a protein and to compute final protein level probabilities. The prophet tools together combine multiple layers of evidence and refine the model iteratively to achieve an optimal analysis of the data. Finally MAYU [36] estimated FDR at different levels for each contributing experiment and for the entire dataset based on the PSMs to decoy proteins.

This process followed the pipeline first implemented in the construction of the human plasma PeptideAtlas described in [14] and successfully applied to other builds such as the bovine milk and mammary gland PeptideAtlas [23].

2.3. Construction of the PeptideAtlas

The PeptideAtlas building process calculates the cumulative number of identified peptide and proteins across the experiments, gathers information on protein to genome location mappings and estimates the peptides' Empirical Suitability Score (ESS and PSS). The genomic mappings, since C. albicans is not present in the Ensembl database, which is the default PeptideAtlas uses to that purpose, were extracted from a generic feature file located at the following url: http://www.candidagenome.org/download/gff/C_albicans_SC5314_version_A21-s02-m05-r10_features.gff.

An overview of how the different experiments contribute, in terms of the number of identified spectra and peptides, to the atlas build is depicted in Fig. 1.

Besides, and due to the particularly rich number of identifications in experiments aimed at the detection of phosphory-lated proteins (experiments #3 to #13), a similarly processed version of the PeptideAtlas was created including in this case PTMProphet results which provide, alongside each modified residue, the probability that the post-translational modification is truly detected at that site.

3. Results and discussion

3.1. Assessment of proteome coverage and functional enrichment analysis

The assembled proteomic datasets (Table 1) were subject to uniform data processing in order to build the C. albicans

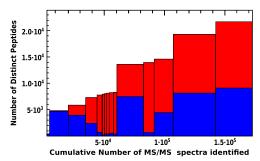


Fig. 1 – Histogram showing the cumulative number of distinct peptides in the *C. albicans* PeptideAtlas. Each bar represents a different experiment that has contributed to the build. Bar width is proportional to the number of high confidence PSMs. Height of the blue section of the bar represents the number of distinct peptides in each experiment and total height of the bar (red plus blue sections) indicates the cumulative number of peptides. The order of experiments is the same as in Table 1.

PeptideAtlas. The PSM assignment and protein inference processes were conducted by means of the consistent and robust pipeline TPP. The prophet tools integrate various levels of information and report identification results in statistical terms so that spectrum assignments, peptide to protein mappings and protein groups are statistically validated, leading to an overall improved sensitivity for a defined FDR level. As a result the generated C. albicans PeptideAtlas comprises 21,938 peptides identified at a 0.24% FDR allocated to 2562 proteins at a 1.2% FDR, that is, a coverage of 41.3% of the 6209 C. albicans translated ORF sequences from the fasta database used for searches. While the presented instance of the C. albicans PeptideAtlas has reached unprecedented coverage, it does not represent a final representation of the respective proteome. Like other PeptideAtlas instances for other species, the C. albicans atlas will be expanded upon submission and processing of new MS data generated in ongoing projects.

To determine the biological functions encompassed by the covered part of the proteome in this PeptideAtlas a Gene Ontology (GO) annotation enrichment analysis was carried out for the list of all detected C. albicans canonical proteins, excluding decoy hits, using the biological process ontology and Genecodis software [37]. Predictably, it generated a diverse array of clusters heterogeneously annotated, among which the largest in number of proteins are associated with the GO terms oxidation-reduction process, cellular response to drug, pathogenesis and hyphal growth respectively (Fig. 2). The enrichment in some very generic GO terms such as oxidation-reduction process, cellular response to drug and translation supports the hypothesis that the diversity of experiments assembled to build the atlas provides a representative, unbiased subset of the C. albicans proteome. In contrast, the more precise groups resulting from the analysis related to pathogenesis, hyphal growth and fungal-type cell wall organization are consistent with the large contribution to the atlas by the experiment aimed at identifying proteins from cells in hyphal form and by the profusion of these sort of annotations in the source database.

As for the set of proteins present in the fasta database used for the searches that are not covered in the PeptideAtlas, they were subject to a similar analysis and were found to be enriched in annotations related to the transmembrane transport GO term (Fig. 2). These proteins are not easily observed by LC-MS/MS techniques as previously reported [20]. Also, we observed enrichment in regulation of transcription, DNA-dependent in the undetected part of the proteome. Given the short life span and low abundance of many transcription factors it is plausible that they were not detected in the collected datasets and their under representation in proteomic data has also been reported in other proteomic studies and in PeptideAtlas instances from other species [20,38,39]. The low number of protein groups significantly associated with GO annotations in the undiscovered set is understandably due to the fact that 2460 out of 3665 of the undetected protein sequences, roughly two thirds, correspond to unnamed ORFs, meaning, that little is known about their biological function.

In addition to the groups of functionally characterized proteins, this PeptideAtlas offers solid empirical evidence for the existence of 1564 proteins, showing a ProteinProphet probability score greater than 0.9, corresponding to uncharacterized ORFs in the CGD database (i.e., one-third of all 4566 uncharacterized ORFs).

3.2. Proteins of interest. Case of use

From the clinical angle, the characterization of the *C albicans* proteome is focused on particular subproteomes, including cell surface constituents, and the set of proteins involved in the yeast-to-hypha transition. The cell wall, as the outermost cell structure represents the contact surface with host cells and therefore gathers many antigens, virulence factors and Pathogen Associated Molecular Patterns (PAMPs) [40]. Proteins

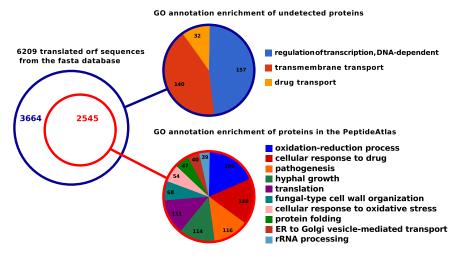


Fig. 2 – Gene Ontology annotation enrichment analysis for both the covered and undetected proteome subsets. All shown GO annotations correspond to the biological process ontology and were found significant for a p-value cut-off below 0.01.

involved in hyphal growth are also relevant in pathogenesis, in the sense that hyphae have been proven as key for invasiveness whereas the switch back to yeast form plays a role in dissemination [41].

Within these groups, a selected set of proteins of interest present in the atlas, are the adhesins from the ALS family with a role in invasiveness Als2p and Als3p; those required for cell wall biogenesis and organization glycosidases Phr1p, Phr2p and Utr2p; mannosyltransferases Pmt1p, Pmt4 and Pmt6; those involved in the cell-wall glucan metabolism Mp65p and

Ecm33p, and the hyphal cell wall constituents Hwp1, Csp37p and Rbt1p.

Other relevant proteins in the atlas are the ones related to apoptosis, since those would make an ideal target for the treatment of invasive candidiasis. Among those, the atlas contains Mca1p, Bcy1p, Ras1p and three unnamed ORFs with orthologous in other species showing roles in the apoptotic process (orf19.713, orf19.967 and orf19.7365).

For any particular proteins of interest, the PeptideAtlas web interface provides tools to explore the data. A user can

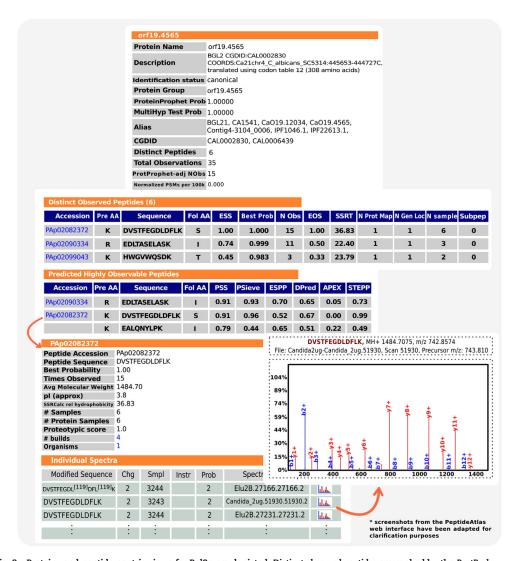


Fig. 3 – Protein- and peptide-centric views for Bgl2p are depicted. Distinct observed peptides are ranked by the BestProb parameter (representing the PeptideProphet probability). Of those, most probably, some will also be present in the following Predicted Highly Observable Peptides table were peptides are ranked by PSS, a combination of different prediction algorithms. For all observed peptides, spectra from the different experiments are also available.

browse through a set of protein and peptide-centric views as illustrated in Fig. 3 for the specific case of Bgl2p, a cell wall glucosyltransferase. Its corresponding observed peptides are highlighted in the protein sequence and sorted by the Empirical Suitability Score (ESS), which represents the proportion of the number of samples in which the peptide is observed with regard to the number of samples in which the original protein is observed. This parameter, in combination with others, such as a number of protein mappings, genome location and amino acid composition will help the user to select candidate proteotypic peptides for a targeted proteomics (SRM, Selected Reaction Monitoring) experiment.

Concerning those cases where a selected protein of interest is not observed in the selected build, the PeptideAtlas also provides the Predicted Suitability Score (PSS), a value resulting from the combination of different observability prediction algorithms based upon physico-chemical properties derived from the amino acid composition and previous training datasets as described in [42].

The build that assembles the phosphoprotein enrichment experiments may be of great potential interest to study biological processes such as signal transduction, since it encompasses a number of kinases and phosphatases. A total of 421 different phosphopeptides were detected and allocated to 210 phosphoroteins. The largest number of phosphorylation sites occurs in S, 410 phosphopeptides contain, at least, one phosphorylation in T; and 10 phosphopeptides contain one phosphorylation in Y.

4. Conclusions

This C albicans PeptideAtlas build provides empirical identification evidence for 21,938 unique peptides including 421 phosphopeptides at a 0.24% peptide-level FDR that account for a high-confidence set (as defined in [14]) of 2562 canonical proteins at a 1.2% protein-level FDR representing thus a significant advance in the proteomic characterization of C. albicans.

Through the web interface, an important set of tools are made available to the scientific community, enabling a solid foundation to study different basic biological processes like dimorphism, signal transduction, apoptosis and the interaction with the human host. Furthermore, its value as a resource for proteotypic peptide selection is of great potential interest for future SRM experiments.

The current version of the PeptideAtlas can be found at:https://db.systemsbiology.net/sbeams/cgi/PeptideAtlas/buildDetails?atlas_build_id=323and the version including PTM results at:https://db.systemsbiology.net/sbeams/cgi/PeptideAtlas/buildDetails?atlas_build_id=324.

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3.1. Introducción

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Desarrollo de una base de datos para datos de Proteómica Dirigida (MRM)

Bibliografía

Y así, del mucho leer y del poco dormir, se le secó el celebro de manera que vino a perder el juicio.

Miguel de Cervantes Saavedra

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-¿ Qué te parece desto, Sancho? - Dijo Don Quijote -Bien podrán los encantadores quitarme la ventura, pero el esfuerzo y el ánimo, será imposible.

> Segunda parte del Ingenioso Caballero Don Quijote de la Mancha Miguel de Cervantes

-Buena está - dijo Sancho -; fírmela vuestra merced. -No es menester firmarla - dijo Don Quijote-, sino solamente poner mi rúbrica.

> Primera parte del Ingenioso Caballero Don Quijote de la Mancha Miguel de Cervantes