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**PREDIÇÃO DE ALVOS QUIMIOTERÁPICOS E *DESIGN* DE VACINA MULTI-
EPÍTOPO EM *Corynebacterium ulcerans* E *C. silvaticum***

BELO HORIZONTE
2022

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Orientador: Prof. Dr. Vasco Ariston de Carvalho Azevedo

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RESUMO

Corynebacterium é um gênero de bactérias gram-positivas que contém espécies de importância médica, veterinária e biotecnológica. Entre as bactérias patogênicas existe um clado contendo seis espécies que podem ser lisogenizadas por corynefagos *tox* e produzir a toxina diftérica (DT), denominado complexo da *C. diphtheriae*. Entre as espécies do complexo,

C. ulcerans é uma bactéria zoonótica que infecta diversos mamíferos e pode causar diferentes doenças. Atualmente linhagens produtoras de DT são a maior causa de difteria, que vem aumentando em países subdesenvolvidos devido a vacinação negligenciada. Pessoas vacinadas podem ser infectadas e transmitir o patógeno. Além disso, a vacina não protege contra linhagens não produtoras de DT, que são capazes de causar outras doenças. *C. silvaticum* é uma espécie recentemente descrita de importância veterinária que causa linfadenite caseosa em porcos e corças, e possui potencial zoonótico. Pouco se sabe sobre sua distribuição geográfica e espectro de hospedeiros. Neste contexto, o desenvolvimento de uma nova vacina capaz de proteger contra linhagens produtoras e não produtoras de DT das duas espécies teria relevância média veterinária. Além disso, para os casos de tratamento da infecção, a identificação de novos quimioterápicos seria uma alternativa em caso de resistência aos atuais. Neste trabalho utilizamos estratégias *in silico* para identificar, para as duas espécies, novas proteínas alvo de quimioterápicos e desenvolver novas vacinas. Foram analisados 72 genomas de *C. ulcerans* e 36 de *C. silvaticum*. Por meio da genômica subtrativa foi possível identificar quatro proteínas alvo de quimioterápicos. Utilizando-se *docking* molecular e uma biblioteca de 5.008 compostos naturais, foram identificados três possíveis quimioterápicos para cada uma das proteínas alvo. Por meio da vacinologia reversa foram identificadas nove prováveis proteínas alvo. Uma análise imunoinformática foi realizada para desenhar vacinas multi-epitopos a partir de 25 epítomos em comum entre oito das nove proteínas identificadas anteriormente. Quatro vacinas multi-epítomos foram desenhadas com diferentes adjuvantes, capazes de induzir respostas imunes inatas e adaptativas. As informações geradas poderão reduzir o custo e o tempo para a descoberta de novos antibióticos e para o desenvolvimento de uma vacina para as duas espécies, capaz de proteger contra linhagens não produtoras de DT.

Palavras-chave: alvos de drogas, *Corynebacterium*, doença zoonótica, imunoinformática, patógeno, vacinologia reversa.

ABSTRACT

Corynebacterium is a genus of gram-positive bacteria that contains species of medical, veterinary and biotechnological importance. Among the pathogenic bacteria there is a clade containing six species that can be lysed by corynephages tox+ and produce diphtheria toxin (DT), called the *C. diphtheriae* complex. Among the species of the complex, *C. ulcerans* is a zoonotic bacterium that infects several mammals and can cause different diseases. Currently DT-producing strains are the major cause of diphtheria, which is increasing in underdeveloped countries due to neglected vaccination. Vaccinated people can become infected and transmit the pathogen. In addition, the vaccine does not protect against non-DT-producing strains, which are capable of causing other diseases. *C. silvaticum* is a recently described species of veterinary importance that causes caseous lymphadenitis in pigs and roe deer, and has zoonotic potential. Little is known about its geographic distribution and host spectrum. In this context, the development of a new vaccine capable of protecting against DT-producing and non-DT-producing strains of the two species would have medium and veterinary relevance. In addition, for cases of infection treatment, the identification of new chemotherapeutics would be an alternative in case of resistance to the current ones. In this work we use *in silico* strategies to identify, for the two species, new chemotherapeutic target proteins and to develop new vaccines. The 72 *C. ulcerans* and 36 *C. silvaticum* genomes were analyzed. By means of subtractive genomics it was possible to identify four target proteins of chemotherapeutic agents. Using molecular docking and a library of 5,008 natural compounds, three possible chemotherapeutics were identified for each of the target proteins. Through reverse vaccinology, nine probable target proteins were identified. An immunoinformatic analysis was performed to design multi-epitope vaccines from 25 epitopes in common among eight of the nine previously identified proteins. Four multi-epitope vaccines were designed with different adjuvants, capable of inducing innate and adaptive immune responses. The information generated could reduce the cost and time for the discovery of new antibiotics and for the development of a vaccine for both species, capable of protecting against non-DT-producing strains.

Keywords: drug targets, *Corynebacterium*, zoonotic disease, immunoinformatics, pathogen, reverse vaccinology.

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LISTA DE ABREVIACOES

- ANI *Average Nucleotide Identity*
- BLAST *Basic Local Alignment Search tool*
- CAMP Christie–Atkins–Munch–Peterson
- CDS *Coding Sequence*
- CWSS *Cell Wall Sorting Signal*
- GO *Gene Ontology*
- DT Toxina diftrica
- DTPa Difteria, Ttano e Pertussis
- EBS Ensaio Bactericida de Soro
- GBDP *Genome Blast Distance Phylogeny*
- KEGG *Kyoto Encyclopedia of Genes and Genomes*
- LGCM Laboratrio de Gentica Celular e Molecular
- LCA Linfadenite caseosa
- MALDI-TOF *Matrix-Assisted Laser Desorption Ionization-Time of Flight*
- MHC Major Histocompatibility Complex
- MLST *Multilocus Sequence Typing*
- MS *Mass Spectrometry*
- NTTB *Non-Toxigenic but Tox gene-Bearing*
- PATRIC *Pathosystems Resource Integration Center*
- PBIT *Pipeline Builder for Identification of Targets*
- PLD *Phospholipase D (Fosfolipase D)*
- RASTtk *Rapid Annotation using Subsystems Technology tool kit*
- SNP *Single Nucleotide Polymorphism*
- ST *Sequence Type*
- TLR Receptor *Toll-Like*
- TYGS *Type (Strain) Genome Server*
- VFDB *Virulence Factors Database*
- VR Vacinologia Reversa
- WGS *Whole Genome Sequencing (Sequenciamento de Genoma Completo)*

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I. INTRODUÇÃO

I.1 Gênero *Corynebacterium*

O gênero *Corynebacterium* se caracteriza por bactérias aeróbicas bastonetes Gram-positivos e corineformes (claviforme), pertencente a classe Actinobacteria. As corinebactérias possuem alto conteúdo de guanina e citosina (G+C) em seus genomas, a parede celular contém arabinose e galactose e a maioria das espécies contém ácido micólico. *Corynebacterium* possui bactérias fermentadoras e não fermentadoras, as espécies de importância clínica são catalase positivas e imóveis (BERNARD; FUNKE, 2015). A espécie tipo do gênero é *Corynebacterium diphtheriae* (Kruse 1886) Lehmann e Neumann 1896, 350 (“*Bacillus diphtheria*” Kruse in Flüge 1886, 225). O gênero possui 136 espécies descritas e publicadas de acordo com o *List of Prokaryotic names with Standing in Nomenclature* (<https://lpsn.dsmz.de/genus/corynebacterium>, acesso em junho de 2022).

Cinquenta e quatro espécies possuem relevância médica, e 14 destas são patógenos zoonóticos transmitidos de diversos mamíferos de produção e domésticos (BERNARD, 2012). Seis espécies patogênicas de interesse médico e veterinário fazem parte de um clado que pode ser lisogenizado por corinefagos *tox*⁺ e produzir a toxina diftérica (DT): *C. diphtheriae*, *C. belfantii*, *C. rouxii*, *C. pseudotuberculosis*, *C. ulcerans* e *C. silvaticum* (BADELL et al., 2020; BERNARD; FUNKE, 2015; DANGEL et al., 2020). Este clado foi denominado de complexo *C. diphtheriae* (BADELL et al., 2020).

C. diphtheriae pode causar difteria em humanos devido a capacidade de algumas linhagens de produzir a DT, sendo o primeiro agente etiológico descrito para esta doença. Além disso, pode causar outras doenças que acometem os sistemas respiratório, nervoso e outros tecidos do sistema circulatório (miocárdio), linfático e urinário (rins). A espécie pode infectar equinos, bovinos e gatos domésticos (GUARALDI; HIRATA; AZEVEDO, 2014). *C. belfantii* (BADELL et al., 2020) e *C. rouxii* (DAZAS et al., 2018) são reclassificações de linhagens previamente consideradas *C. diphtheriae* biovar belfanti. *C. pseudotuberculosis* é o patógeno causador da linfadenite caseosa em caprinos e ovinos, porém infecta outros mamíferos, incluindo humanos, causando diferentes manifestações (GUARALDI; HIRATA; AZEVEDO, 2014). *C. ulcerans* e *C. silvaticum* serão discutidas em seguida.

1.2 *Corynebacterium ulcerans*

C. ulcerans foi isolada de humano com doença respiratória similar à difteria, e descrita previamente por Gilbert e Stewart em 1927 (RIEGEL et al., 1995). As linhagens de *C. ulcerans* caracterizam-se por serem bastonetes Gram-positivos, pleomórficos, catalase positivos, não móveis e sem esporos, podem apresentar arranjos coccobacilares sem grânulos de fosfato, ter em seu DNA alto conteúdo de G+C. As principais moléculas componentes de sua parede celular são os açúcares galactose e arabinose, e ácidos micólicos. As cepas são capazes de fermentar maltose e glicose, produz urease e DNase, cistinase positivo em ágar de Tinsdale e negativo para o teste de PYZ (pirazinamidase). Reações reversas de Christie–Atkins–Munch–Peterson (CAMP) são observadas em ágar sangue quando inoculadas transversalmente com uma cepa de *Staphylococcus aureus* (BERNARD; FUNKE, 2015). Além da DT, *C. ulcerans* pode produzir a exotoxina Fosfolipase D (PLD), que hidroliza a esfingomiéline das membranas celulares de mamíferos. Esta toxina auxilia a disseminação das bactérias do sítio inicial da infecção e causa a morte de macrófagos (TAUCH; BURKOVSKI, 2015). A linhagem tipo desta espécie é a NCTC 7910^T (RIEGEL et al., 1995).

C. ulcerans infecta humanos e diversos animais. Seu principal reservatório são diversos mamíferos de grande e pequeno porte, incluindo roedores, suínos e primatas. Linhagens que não produzem DT podem causar diversas doenças no sistema respiratório como pneumonia, sinusite, amigdalite, faringite, peritonite e outras relacionadas mais comuns. Zoonoses podem ser transmitidas principalmente de animais de produção, mas também por animais domésticos e selvagens (HACKER et al., 2016).

Casos de transmissões e infecções por *C. ulcerans* vêm aumentando, principalmente em países subdesenvolvidos, mesmo em populações vacinadas com a DTPa (difteria, tétano e pertussis). Essa vacina é baseada no toxoide da DT, mas não está sendo suficiente para o controle da difteria e não protege contra outras doenças causadas por linhagens não produtoras da toxina diftérica (HACKER et al., 2016).

Corynebacterium ulcerans fez-se, atualmente, um patógeno emergente em humanos tanto em países industrializados quanto em desenvolvimento. Nesses últimos, tem-se considerado um baixo número de notificações de infecções, em consequência de escassa especialização de diagnóstico. Com isso, faz-se necessário aprimoramento do monitoramento global dessas infecções que envolvem testes em humanos e animais previamente contatados com a bactéria para que possa ser identificada e compreendida a forma como a doença é transmitida. Além do exposto, é requerido maior informações sobre fatores de virulência,

mecanismos de colonização em hospedeiros por esse organismo, já que ainda são insuficientes. Em vista disso, estudos básicos e adicionais sobre essas demandas são fundamentais para o entendimento da patogenicidade e consequente infecção dessa espécie (HACKER et al., 2016).

1.3 Corynebacterium silvaticum

Corynebacterium silvaticum foi isolada de abscessos de Linfadenite Caseosa (LCA) de javalis selvagens (*Sus scrofa*), corça (*Capreolus capreolus*), e porcos domésticos (*Sus scrofa domesticus*), e identificada previamente como *C. pseudotuberculosis* ou *C. ulcerans* (CONTZEN et al., 2011; EISENBERG et al., 2014; OLIVEIRA et al., 2014). *C. silvaticum* tem crescimento lento em condições aeróbicas ou microaeróbicas, as colônias são pequenas, cerosas a brancas cremosas, com superfície lisa e discreta β -hemólise após 48 h de incubação em Ensaio Bactericida de Soro (EBS) e crescimento bom em pH 7 a 8. As linhagens são *tox* positivas mas não expressam a DT, portanto são linhagens portadoras de *tox* mas não toxigênicas (*Non-toxigenic tox gene bearing* ou NTTB) (DANGEL et al., 2020). A espécie pode ser identificada pela filogenia dos genes 16S e *rpoB*, filogenômica, Tipagem de Sequência Multilocus (MLST), conteúdo G+C (54.5%), espectrometria de massa *Matrix-Assisted Laser Desorption Ionization-Time of Flight* (MALDI-TOF MS). A linhagem tipo desta espécie é a KL0182^T (DANGEL et al., 2020).

A variedade de hospedeiros da espécie ainda é desconhecida, o que se sabe é que javalis selvagens, porcos domésticos e corças são seus reservatórios (DANGEL et al., 2020). A distribuição geográfica inclui Alemanha (CONTZEN et al., 2011; DANGEL et al., 2019; EISENBERG et al., 2014), Áustria (BUSCH et al., 2019), Portugal (MÖLLER et al., 2020; OLIVEIRA et al., 2014; VIANA et al., 2020) e Suíça (<https://www.ncbi.nlm.nih.gov/biosample/SAMN16955830>). *C. silvaticum* pode ser um patógeno humano devido ao seu potencial zoonótico por análise *in silico* (VIANA et al., 2020) e a confirmação de sua toxicidade para células epiteliais humanas *in vitro* (MÖLLER et al., 2021). Essa espécie já é considerada de importância veterinária, pois causa linfadenite em mamíferos.

1.4 Aplicação de análises genômicas para o controle de patógenos

Como discutido anteriormente, *C. ulcerans* e *C. silvaticum* possuem como reservatórios animais domesticados e selvagens, além de relevância médica e veterinária (HACKER et al.,

2016; MÖLLER et al., 2021). A vacina DTPa foi desenhada para proteger contra linhagens produtoras de DT, não oferecendo proteção contra linhagens não produtoras e que são capazes de causar outras doenças (HACKER et al., 2016). Uma vacina mais eficiente poderia utilizar proteínas alvo compartilhadas entre o maior número de linhagens possível, além de oferecer proteção contra as duas espécies. Neste contexto, a análise de genomas contribui para o entendimento da biologia dos patógenos e para propor estratégias de controle (HARDISON, 2003). Com o auxílio de ferramentas computacionais, pode-se realizar previsões e simulações experimentais (*in silico*) qualitativa e quantitativamente (NOGUEIRA et al., 2021). As próximas seções abordarão métodos e aplicações da análise de genomas para o controle de patógenos.

1.5 Genômica comparativa

A genômica comparativa implica na comparação de genomas de diferentes organismos por meio de suas sequências de nucleotídeos para inferir função, conservação, filogenia, evolução, níveis de expressão gênica, entre outros (HARDISON, 2003). A comparação de genomas bacterianos de uma mesma espécie revelou uma variação no conteúdo gênico e deu origem ao conceito de pangenoma, caracterizado por um genoma central composto de genes compartilhados por todos os genomas e um genoma dispensável composto de genes ausentes em alguns genomas ou linhagem-específicos (MEDINI et al., 2005). O genoma central contém genes que codificam funções básicas da biologia celular, enquanto o genoma dispensável, ou acessório, pode conter genes associados a adaptação à diferentes ambientes. Um pangenoma é considerado aberto quando o sequenciamento de novos genomas aumenta significativamente o número de novos genes descobertos (TETTELIN et al., 2008).

A genômica comparativa é amplamente usada em estudos de doenças infecciosas, contribuindo para a epidemiologia (origem e propagação) molecular. É vantajosa também, para a identificação de genes de resistência à antibióticos e identificação de fatores de virulência, consequentemente podendo estipular a possibilidade de um organismo causar doenças (MOUSTAFA et al., 2020).

Neste contexto, a equipe do LGCM tem publicado nos últimos anos estudos que utilizam análises comparativas para elucidar diversidade genômica, filogenômica, mecanismos de patogenicidade e predição de alvos vacinais e terapêuticos para diferentes espécies do gênero *Corynebacterium* (VIANA et al., 2022; GOMES et al., 2022; JAISWAL et al., 2021; VIANA

et al., 2017, 2020; VILELA RODRIGUES et al., 2022).

I.6.1 Genômica Subtrativa

A Genômica Subtrativa é uma abordagem comumente utilizada na identificação de alvos terapêuticos e no desenvolvimento *in silico* de vacinas. Fundamenta-se na filtragem de sequências proteômicas do patógeno baseada nos principais critérios: (i) fator de virulência, (ii) essencialidade na sobrevivência do organismo e (iii) não homologia com o hospedeiro. No fim, resultam os alvos selecionados de vacinas ou drogas, que posteriormente podem ser submetidos a outras análises (BARH et al., 2011). Em pesquisas com bactérias e vírus, o pangenoma é utilizado na abordagem de genômica subtrativa, pois estes organismos são divididos em linhagens, biovars e serovars. Neste caso, utiliza-se o genoma central no intuito de selecionar alvos em comum em todas as linhagens analisadas (PRADO et al., 2021).

I.6.2 Imunoinformática e Vacinologia Reversa

Imunoinformática é um campo da Bioinformática, cuja interface está entre métodos computacionais e a imunologia experimental. Utiliza-se de ferramentas computacionais para o entendimento de respostas imunes e testar hipóteses experimentais *in silico*. A principal aplicação da Imunoinformática é no *design* de vacinas e modelagem do sistema imunológico (OLI et al., 2020; TOMAR; DE, 2010).

A vacina multi-epítipo é constituída por uma proteína quimérica, resultante da junção de vários epítipos selecionados de proteínas do patógenos, pequenos *linkers* e adjuvantes que são sequências peptídicas para o aumento da imunogenicidade da vacina. Epítipos são pequenos peptídeos presentes em antígenos, com a capacidade de induzir resposta imune pela ligação de anticorpos específicos de seu sítio ativo. A vacina multi-epítipo é testada *in silico* em análises de interação molecular com um Receptor *Toll-Like* (TLR), a fim de checar a adequação da vacina candidata (VILELA RODRIGUES et al., 2022).

As áreas de pesquisa desse campo incluem: predição de epítipos de células B e T, identificação de genes/proteínas virulentos, interação patógeno-hospedeiro, predição de alergia, vacinação *in silico*, *design* de banco de dados imunológicos, *design* e engenharia de terapias e diagnósticos imunológicos, estudo de genes e suas expressões em estados saudáveis e doentes, compreensão de genes relacionados ao sistema imunológico, modelagem de resposta do sistema imunológico e simulação de experimentos de laboratório. Existem pelo menos 14 bancos de

dados imunológicos que permitem as pesquisas nas diversas áreas e aplicações citadas anteriormente (OLI et al., 2020; TOMAR; DE, 2010).

No desenvolvimento de vacinas, a vacinologia convencional embasa-se no isolamento e cultivo do patógeno e posterior identificação dos elementos essenciais para a imunidade. Existem dois tipos principais de vacinas: as que contêm vírus ou bactérias vivos enfraquecidos (atenuados) e as que apresentam esses organismos mortos (inativados). O tipo inativado pode conter o organismo inteiro ou ser uma vacina de subunidades, que são construídas por componentes antigênicos ou que induzem resposta imunológica como proteínas, polissacarídeos e ácidos ribonucleicos (BAGNOLI et al., 2011).

Com o surgimento e o desenvolvimento da genômica, foi possível o surgimento da abordagem de vacinologia reversa (VR). Esse termo foi mencionado pela primeira vez em 2001, o qual consiste na utilização de dados genômicos e análises computacionais para a identificação dos componentes necessários para a geração de imunidade no hospedeiro (BAGNOLI et al., 2011). Basicamente, através dos genomas pode-se utilizar *pipelines* para prever proteínas dos micro-organismos de interesse, identificar suas propriedades imunogênicas e essenciais para o metabolismo do patógeno, além de analisar e simular a interação de moléculas terapêuticas (existentes em bancos de dados) com essas proteínas.

O primeiro imunizante feito através da vacinologia reversa, foi a vacina meningocócica *Bexsero*^R contra a bactéria *Neisseria meningitidis* (serogrupo B) (SERRUTO et al., 2012). A partir de então, a abordagem foi utilizada para muitas outras bactérias patogênicas, além de vírus e fungos. Desde o primeiro protocolo de VR até os dias atuais, avanços em Sequenciamento de Genomas Completos (WGS) contribuiu para ampliação da aplicabilidade de VR na genômica, imunologia, entre outras áreas afins. Além de identificar alvos quimioterápicos para a produção de drogas e vacinas, pode-se também avaliar a diversidade e evolução de antígenos, juntamente com a interação patógeno-hospedeiro. Concomitantemente, vários programas e ferramentas foram desenvolvidos para patógenos e vêm sendo refinados para a identificação e a classificação de moléculas antigênicas e, também, com dados genômicos e imunológicos sobre os hospedeiros (DALSASS et al., 2019; MOXON et al., 2020).

II. OBJETIVOS

II.1 Objetivo geral

O objetivo deste trabalho foi a predição *in silico* de proteínas alvo candidatas para tratamento com quimioterápicos e a predição de vacina multi-epítipo para o controle de *Corynebacterium ulcerans* e *C. silvaticum*.

II.2 Objetivos específicos

Os objetivos específicos foram:

- a) Montar e anotar os genomas de *C. ulcerans* e *C. silvaticum* os quais apenas os dados brutos estavam disponíveis publicamente
- b) Identificar proteínas alvo e moléculas candidatas a quimioterápicos
- c) Identificar proteínas antigênicas candidatas para o desenvolvimento de vacina
- d) Predizer vacinas multi-epítipo para *C. ulcerans* e *C. Silvaticum*

III. CAPÍTULO I

ARTIGO DE PESQUISA

Neste capítulo foi feita uma pesquisa *in silico* sobre as bactérias *Corynebacterium ulcerans* e *C. silvaticum*, através das abordagens de genômica subtrativa e vacinologia reversa, para a identificação de proteínas alvos de quimioterápicos e proteínas antigênicas que possam ser utilizadas no desenvolvimento de vacinas. Além disso, foram propostos compostos naturais como prováveis quimioterápicos para as proteínas drogáveis. Essa é uma nova investigação feita para ambas as bactérias desse gênero, no intuito de propor estratégias para o controle de infecções causadas por elas.

Candidate common targets (vaccine and drug) and potential drug likemolecules against *Corynebacterium ulcerans* and *C. Silvaticum*

Janaína Canário Cerqueira^a, Marcus Vinicius Canário Viana^b, Arun Kumar Jaiswal^a, Siomar Castro Soares^c, Sandeep Tiwari^a, Alice Rebecca Wattam^d, Artur Silva^b, Khalid J.Alzahrani^e, Murtaza M. Tambuwala^f, Debmalya Barh^{ag}, **Vasco Azevedo^{a*}**

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Running title: Targets against *C. silvaticum* and *C. ulcerans*.

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Abstract:

The availability of complete genomes of pathogens enables *in silico* analyses that can be used to develop new control methods, reducing the time, cost, and necessity of pathogen cultivation. In this study, we used bioinformatics' approach to predict common drug and vaccine candidates against the diphtheria toxin producing *Corynebacterium ulcerans* and the recently characterized *C. silvaticum* species, considered as a zoonotic potential. We found four common, non-host homologous, virulent, essential cytoplasmic druggable proteins that belong to metabolic pathways and are involved in the regulation in other essential genes. In addition, the docking analysis showed drug like natural compounds as possible drug candidate against the target proteins. We also identified nine vaccine candidates involved in transport and regulation of permeability of substances important to the cell. We hypothesize that these identified vaccine candidates, drug targets and drug molecules could be considered for prophylaxis and hence, should be subjected to further experimental validations.

Keywords: Drug target, homology modeling, reverse vaccinology, subtractive genomics, zoonotic.

Candidate common targets (vaccine and drug) and potential drug like molecules against *Corynebacterium ulcerans* and *C. silvaticum*

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1. Introduction

The genus *Corynebacterium* comprise Gram positive, non-motile and sporeless rods, with strains capable of producing diphtheria toxin (DT) or carrying the gene. *C. ulcerans* is a pathogenic bacterium belonging to the same clade as *C. diphtheria* and *C. tuberculosis*. It is a species of economic and social importance, as it is the main current etiological agent of diphtheria in addition to infecting several mammals. Some strains produce DT, others do not produce the toxin yet cause other human diseases (endocarditis, septic arthritis, osteomyelitis and sepsis) (Hacker et al. 2016). *C. silvaticum* was recently described and is closely related to *C. ulcerans*. This bacterium has pig and deer as reservoirs, but its potential host range is still unknown. It is known so far that it is cytotoxic in humans (Möller et al. 2021). Therefore, the bacterium has a zoonotic potential that leads to a potential economic loss in animal production

(Dangel et al. 2020; Viana et al. 2020).

C. ulcerans was selected based on its emergence as a causative agent of diphtheria and the fact that it has a wide range of reservoir species (Tiwari et al. 2008; Hacker et al. 2016), which increases its threat potential. *C. silvaticum* was selected based on its ability to infect production animals and has a, as yet, unknown potential to cause zoonotic infections (Dangel et al. 2020; Viana et al. 2020). The current vaccine for diphtheria uses the toxoid from *C. diphtheriae* (Rappuoli and Malito 2014) and the lack of sequence diversity could lower the efficiency of the

current vaccine and antitoxin against *C. ulcerans tox*⁺ strains (Otsuji et al. 2019) and could play a role in diphtheria cases in vaccinated individuals (Vandentorren et al. 2014). Other factors that impede the control of this disease is that immunization does not last for a lifetime and decreases with time, and has the possibility of immunized people being colonized and transmitting the disease (Truelove et al. 2020). Additionally, non-toxigenic strains of *C. ulcerans* can cause severe infections such as septic arthritis, endocarditis, osteomyelitis and sepsis (Zasada 2013; Grosse-Kock et al. 2017).

Genomic data can directly assist in the development of control methods. Reverse vaccinology is the prediction of target proteins for vaccine development based on genomic sequencing data, decreasing the time and cost required and the need for cultivation of the pathogen (Rappuoli 2000; Rappuoli et al. 2016). Examples of products developed from this approach are the vaccines *Bexsero* for *Neisseria meningitidis* (Christodoulides and Heckels 2017) and *Engerix-B* for Hepatitis B (Keating and Noble 2003). Recently, vaccine targets have been predicted *in silico* and tested *in vitro* or *in vivo* for *N. meningitidis* (Masignani et al. 2019), *Shigella* (Hajjalibeigi et al. 2021; de Alwis et al. 2021), *Leishmania mexicana* (Burgos-Reyes et al. 2021), and *Rhipicephalus bursa* (Couto et al. 2021). Drug targets and their respective drugs can also be predicted by a *in silico* approach (Das et al. 2021).

In this work, we applied a standard computational approach to identify common vaccine and drug targets and targeting drug like molecules against *Corynebacterium ulcerans* and *C. silvaticum*.

2. Materials and method

2.1. Samples

Genome data of 108 strains were retrieved from GenBank, 72 of *C. ulcerans* and 36 of *C. silvaticum* (Supplementary Table S1). Of the 108 genomes, 76 were available as sequencing

reads from the Sequence Read Archive (Supplementary Table S1, SRR accession only). They were assembled in PATRIC (Davis et al. 2020) using Unicycler (Wick et al. 2017). All the genomes were annotated in PATRIC using RASTtk (Brettin et al. 2015). The taxonomy of the samples had previously been verified as *C. ulcerans* or *C. silvaticum* (Viana et al. 2020) using FastANI (Jain et al. 2018), and were additionally submitted to the Type Strain Genome Server (Meier-Kolthoff and Göker 2019) to confirm their taxonomy.

2.2. Identification of new targets

2.2.1. Identification of core, non-homologous to human host, essential genes, and cell localization

The core proteome was obtained using OrthoFinder v2.3.11 (Emms and Kelly 2015). Of these, proteins not homologous to the host were identified by BLASTp (Camacho et al. 2009) against the human proteome (GenBank accession GRCh38.p13). Furthermore, the dataset of core and non-host homologous proteins were used for essential protein identification by the PipelineBuilder for Identification of Targets (PBIT) (Shende et al. 2017). Finally, the cellular location of the proteins was identified using the SurfG+ v1.2.1 (Barinov et al. 2009). Putative Surface Exposed (PSE), secreted and membrane proteins were screened for probable vaccine targets, while cytoplasmatic proteins were screened for probable drug targets.

2.2.2. Characterization of vaccine targets

For secreted PSE, and membrane proteins, the Vaxign2 web-based server (He et al. 2010) was used to identify likely vaccines targets using the following criteria: i) epitope binding to the Major Histocompatibility Complex I and II (MHC I and MHC II) and; ii) probability of being an adhesin greater than 0.51. For virulence prediction, the server Virulence Factor Database (VFDB) was used (CHEN *et al.*, 2005). These genes were reannotated for function using eggNOG-mapper (Huerta-Cepas et al. 2017). Genes in genomic islands (GI), specifically in pathogenic islands (PAIs), were predicted using GIPSy (Soares et al. 2016) with *C. glutamicum* 13032 (CP025533.1) as non-pathogenic reference.

2.2.3. Identification of drug targets and docking analysis

For the 84 cytoplasmatic proteins, 3D structures were predicted through the MHOLline pipeline (<http://www.mholline2.lncc.br/>). Genes with predicted structures classified as G2 (e-

value $\leq 10e^{-5}$ and Identity ≥ 0.25 and Length Variation Index ≤ 0.7) were selected for the following steps. The Ramachandran plot of the structures was generated PROCHECK (Laskowski et al. 1993) implemented in SAVES server v6.0 (<https://saves.mbi.ucla.edu/>). Furthermore, the target proteins were prioritized according to function and metabolic pathway using eggNOG-mapper, molecular weight using Uniprot (Wasmuth et al. 2017), virulence using (VFDB) (CHEN *et al.*, 2005), and genomic islands using GIPSy.

The residues with high drug scores were identified using DoGSiteScorer (Volkamer et al. 2012). InterProScan online software was used to identify the domains of each of these proteins (JONES *et al.*, 2014). Moreover, the DoGSiteScorer pocket with the highest drug score and the residues of the protein domain were used for the preparation of targets for docking using AutoDockTools (Morris et al. 2009). To identify drugs for these target proteins, a library of 5,008 drug-like natural compounds were obtained from the ZINC database (<http://zinc.docking.org/>). The docking analysis was made using AutoDock Vina (Trott and Olson 2010). The top 10 compounds were analysed for interactions between their residues interaction and the target proteins using Chimera (Pettersen et al. 2004).

2.3. Workflow

The methods used here and the total number of proteins in each step are briefed in Figure 1.

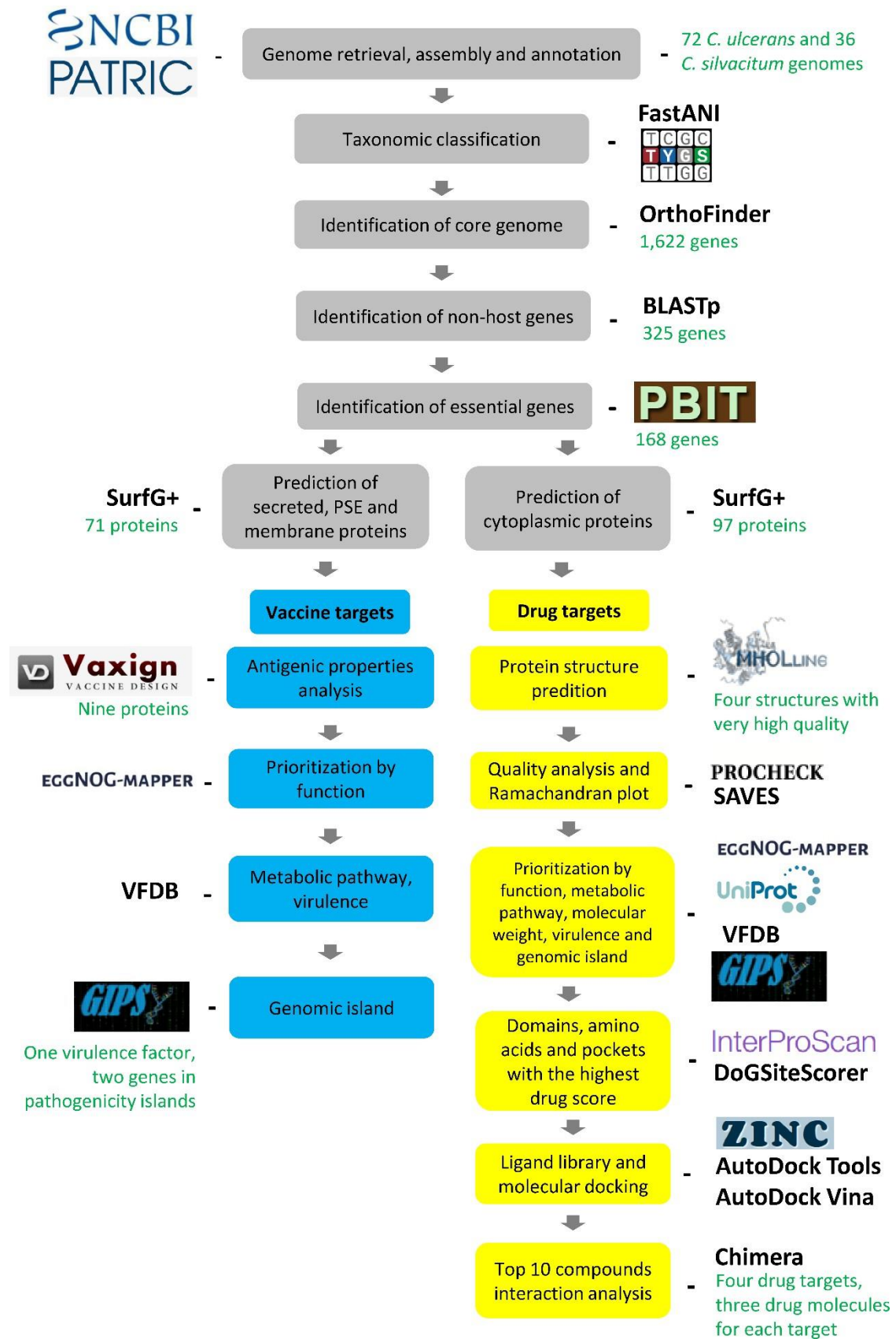


Figure 1. Workflow of the methodologies and the number of genes identified and selected at each step.

3. Results

3.1. Nine candidate vaccine targets are common to *C. silvaticum* and *C. ulcerans*

1,622 proteins were identified as the core proteome, and of these, 325 were characterized as non-host proteins, 168 as essential, 71 as PSE or secreted and 97 as cytoplasmatic proteins. Within the PSE, Vaxign2 predicted nine vaccine targets with adhesin probability > 0.51 and binding epitopes for MHC I and II (Figure 1, Table 1). Two of these were in pathogenicity islands. An ABC transporter, substrate-binding protein (*mntA*) was found in PAI 3, and a phosphate ABC transporter, substrate-binding protein PstS (TC 3.A.1.7.1) (*pstS*) was found in PAI 6 (Figure 2, Supplementary Table S2).

Table 1. Vaccine targets candidates for *Corynebacterium ulcerans* and *C. silvaticum*.

Product (Gene)	Adhesin probability	MHC class	Virulence factor /PATRIC ID (GI) ^a
ABC transporter, substrate-binding protein (cluster 14, Mn/Zn)	0.630	I and II	No 996634.5.peg.167
ABC transporter, substrate-binding protein (cluster 14, Mn/Zn) (<i>mntA</i>)	0.602	I and II	Yes (PAI 3) ^b 996634.5.peg.623
Glutamate ABC transporter, permease protein 1 GluC (<i>gluC</i>)	0.555	I and II	No 996634.5.peg.835
Integral membrane protein	0.566	I and II	No 996634.5.peg.316
Hypothetical protein	0.512	I and II	No 996634.5.peg.652
Hypothetical protein	0.586	I and II	No 996634.5.peg.823
Hypothetical protein	0.661	I and II	No 996634.5.peg.1276
Phosphate ABC transporter, substrate-binding protein PstS (TC 3.A.1.7.1) (<i>pstS</i>)	0.696	I and II	No (PAI 6) 996634.5.peg.1997
Phospholipase / thioesterase	0.571	I and II	No 996634.5.peg.1106

^a genomic island

^b pathogenicity island

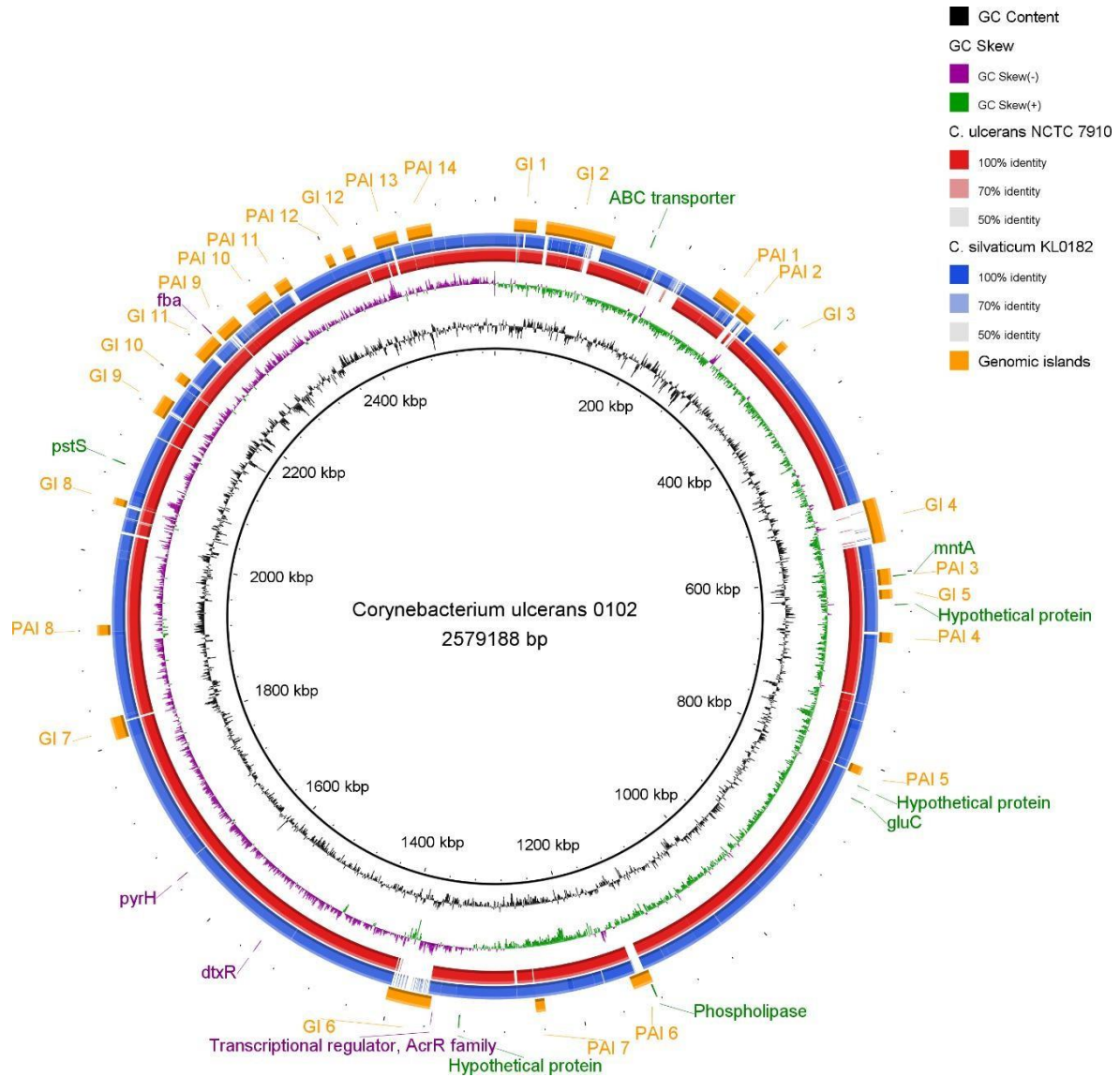


Figure 2. Circular map of *Corynebacterium ulcerans* 0102 showing genomic islands and the position of vaccine (green) and drug targets (purple). GI – genomic island, PAI – pathogenicity island.

3. Four common putative drug targets in *C. silvaticum* and *C. ulcerans*

For 84 proteins, the sequence alignment with a reference protein had the structure quality classified as G2 (e-value $\leq 10^{-5}$, identity $\geq 0.25\%$ and Length Variation Index ≤ 0.1). From these, only the four proteins with alignment quality classified as Very High (identity $\geq 0.75\%$ and Length Variation Index ≤ 0.1) were selected for further analysis (Table 2). The Ramachandran plot for 3D structure validation of these final 4 targets is shown in Supplementary Figure S1, and their respective active residues in Table S3. These four potential targets are essential and only DtxR is a virulence factor according to the PBIT database prediction (Supplementary Table S2). None of the drug targets were in a PAI or GI (Figure 2).

Table 2. Drug target candidates for *Corynebacterium ulcerans* and *C. silvaticum*.

Product (Gene)	Molecular Function	Biological Process	Mass (Da)	Metabolic Pathways	Virulence Factor	Ramachandran residues in most favorable region	PATRIC ID
Iron-dependent repressor IdeR/DtxR (<i>dtxR</i>)	DNA-binding transcription factor activity, protein dimerization activity, transition metal ion binding	Transcription, Transcription regulation	25,203	-	Yes	92.2%	996634.5.peg.1487
Uridine monophosphate kinase UMPK (EC 2.7.4.22) (<i>pyrH</i>)	ATP binding, UMP kinase activity	'de novo' CTP biosynthetic process	26,168	CTP biosynthesis via de novo pathway and in Pyrimidine metabolism	No	94.2%	996634.5.peg.1586
Fructose-Biphosphate aldolase classII (EC 4.1.2.13) (<i>fba</i>)	Fructose-bisphosphate aldolase activity, zinc ion binding	Glycolytic process	37,144	Synthesizes D-glyceraldehyde 3-phosphate and glyceraldehyde 3-phosphate from D-glucose	No	94.9%	996634.5.peg.2154
Transcriptional regulator, AcrR family	DNA binding, magnesium ion binding	Regulation of transcription	21,831	-	No	95.5%	996634.5.peg.1300

3.3 Top 12 candidate drug like molecules to target four common targets

We have used DoGSiteScorer to select the druggable pockets with score greater than

0.79 (Supplementary Table S3). For the docking analysis, we selected three natural compounds (for each target) based on lower binding energy values and the largest number of hydrogen bonds. The output was the 10 top ranked ligands, based on the best interactions with the proteins active residues (domain), according to the requisites mentioned. We selected the final 3 out of 10 best compounds for each target protein. The hydrogen bonds, binding energy values, ligands IDs and proteins information are shown in Table 3. For the graphic representation of the docking analysis, we selected the best one compound for each protein (Figures 3-6).

AcrR family protein is a transcriptional regulator that normally acts as a repressor and had better interaction with the compound ZINC04258896 (Figure 3). Fructose-biphosphate aldolase is an enzyme that participates in the glycolytic pathway and showed the best docking with the compound ZINC04235626 (Figure 4). The IdeR/DtxR target is an iron-dependent transcriptional regulator, classified as a virulence factor that showed the best interaction with the ligand ZINC03840461 (Figure 5). Finally, Uridine monophosphate kinase is a converting enzyme in the metabolism of pyrimidines, It had the best docking with the ligand ZINC08300249 (Figure 6).

Table 3. Docking results of candidate proteins targets and its ligands showing number of hydrogen bonds with its respective residues and binding energy values.

Protein (PATRICID)	Ligand ZINC ID	IUPAC name	Binding energy value (Kcal/mol)	H bond / Residues
Transcriptional regulator, AcrR family (996634.5.peg.1300)	ZINC04258896	1-(1,3-benzodioxol-5-yl)-N-methyl-2-[(E)-3-phenylprop-2-enoyl]-1,3,4,9-tetrahydropyrido[3,4-b]indole-3-carboxamide	-7.9	3/Arg21, Arg92
	ZINC04277685	N-[(1R,9S)-11-(naphthalene-2-carbonyl)-6-oxo-7,11-diazatricyclo[7.3.1.0 ^{2,7}]trideca-2,4-dien-5-yl]cyclohexanecarboxamide	-8	2/Tyr29, Asp54
	ZINC08300353	N-[(1R,9S)-11-(4-fluorobenzoyl)-6-oxo-7,11-diazatricyclo[7.3.1.0 ^{2,7}]trideca-2,4-dien-3-yl]-2H-1,3-benzodioxole-5-carboxamide	-8.2	2/Arg22, Asp94

Fructose- biphosphate aldolaseclass II (996634.5.peg.2154)	ZINC04235626	(3R,3aS,6S,6aR)-6-(naphthalene-2-sulfonamido)-hexahydrofuro[3,2-b]furan-3-ylN-(4-methoxyphenyl)carbamate	-9.1	3/Thr29, Ser53, Asp276
	ZINC04235829	(3R,3aS,6S,6aR)-6-(naphthalene-2-sulfonamido)-hexahydrofuro[3,2-b]furan-3-ylN-(4-acetylphenyl)carbamate	-9.3	3/Thr29, Ser53, His212
	ZINC08300421	N-[(1R,9S)-11-(1-acetylpiperidin-4-yl)-6-oxo-7,11-diazatricyclo[7.3.1.0 ^{2,7}]trideca-2,4-dien-3-yl]-2H-1,3-benzodioxole-5-carboxamide	-9.6	2/Gly213, Thr277
Iron-dependent repressor IdeR/DtxR (996634.5.peg.1487)	ZINC03840461	N-[(11S,13S)-5-(4-chlorophenyl)-2,10-dioxo-1,9-diazatricyclo[9.4.0.0 ^{3,8}]pentadeca-3(8),4,6-trien-13-yl]pyridine-3-carboxamide	-9.8	2/Ser70, Thr159
	ZINC08300249	(1R,9S)-11-(2H-1,3-benzodioxole-5-carbonyl)-5-[3-(trifluoromethyl)phenyl]-7,11-diazatricyclo[7.3.1.0 ^{2,7}]trideca-2,4-dien-6-one	-10.2	2/Thr124, Leu135
	ZINC04259070	(1R,9R)-5-[[3-(3-fluorophenyl)carbamoyl]amino]-N-(4-methoxyphenyl)-6-oxo-7,11-diazatricyclo[7.3.1.0 ^{2,7}]trideca-2,4-diene-11-carboxamide	-10.4	2/Thr24, Thr191
Uridine monophosphate kinase (996634.5.peg.1586)	ZINC08300249	(1R,9S)-11-(2H-1,3-benzodioxole-5-carbonyl)-5-[3-(trifluoromethyl)phenyl]-7,11-diazatricyclo[7.3.1.0 ^{2,7}]trideca-2,4-dien-6-one	-10.3	2/Gly59, Arg64
	ZINC08300441	3-[(1R,9S)-11-(1-acetylpiperidin-4-yl)-6-oxo-7,11-diazatricyclo[7.3.1.0 ^{2,7}]trideca-2,4-dien-3-yl]-1-(3-cyanophenyl)urea	-8.6	3/Glu22, Phe145, Thr147
	ZINC08300421	N-[(1R,9S)-11-(1-acetylpiperidin-4-yl)-6-oxo-7,11-diazatricyclo[7.3.1.0 ^{2,7}]trideca-2,4-dien-3-yl]-2H-1,3-benzodioxole-5-carboxamide	-9.9	2/Gly59, Arg64

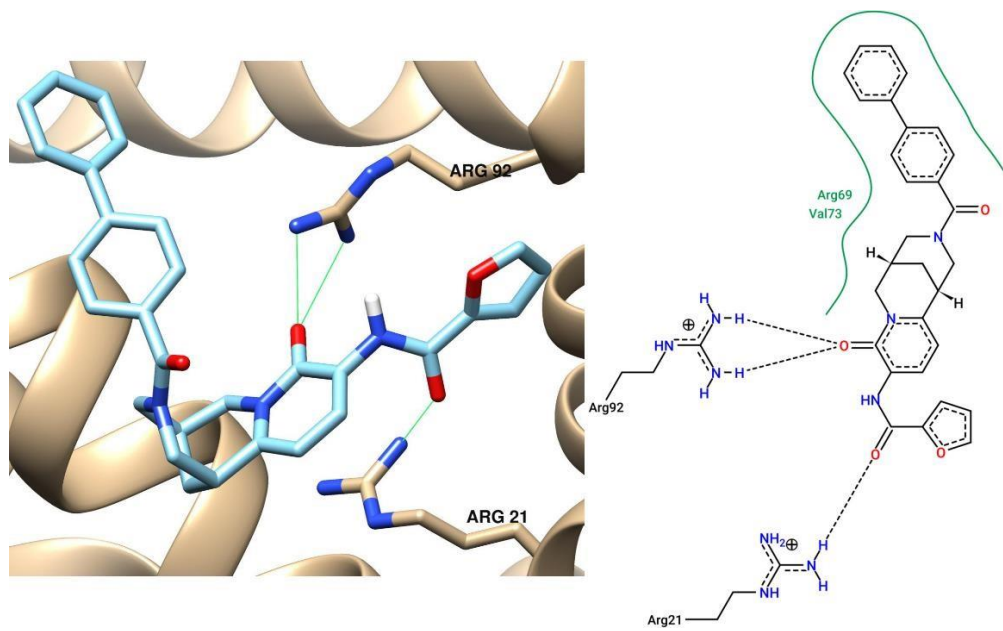


Figure 3. 3D and 2D graphic representation of the docking analysis for the most druggable protein cavity of drug target AcrR family protein with ZINC04258896.

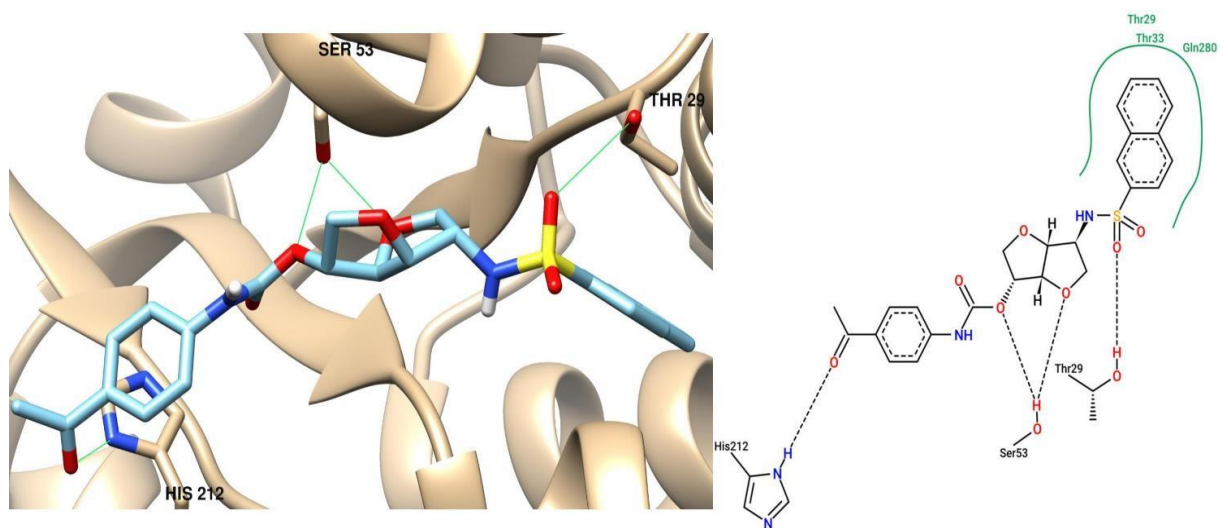


Figure 4. 3D and 2D graphic representation of the docking analysis for the most druggable protein cavity of drug target Fructose-bisphosphate aldolase class II with ZINC04235829.

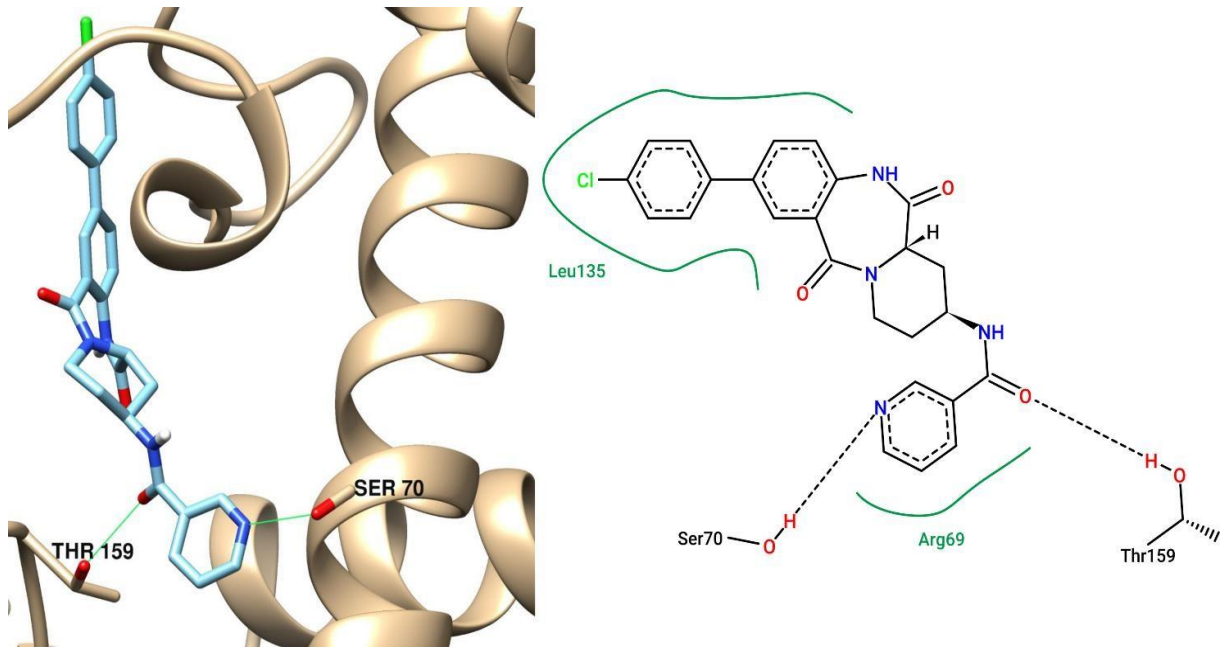


Figure 5. 3D and 2D graphic representation of the docking analysis for the most druggable protein cavity of drug target Iron-dependent repressor IdeR/DtxR with ZINC03840461.

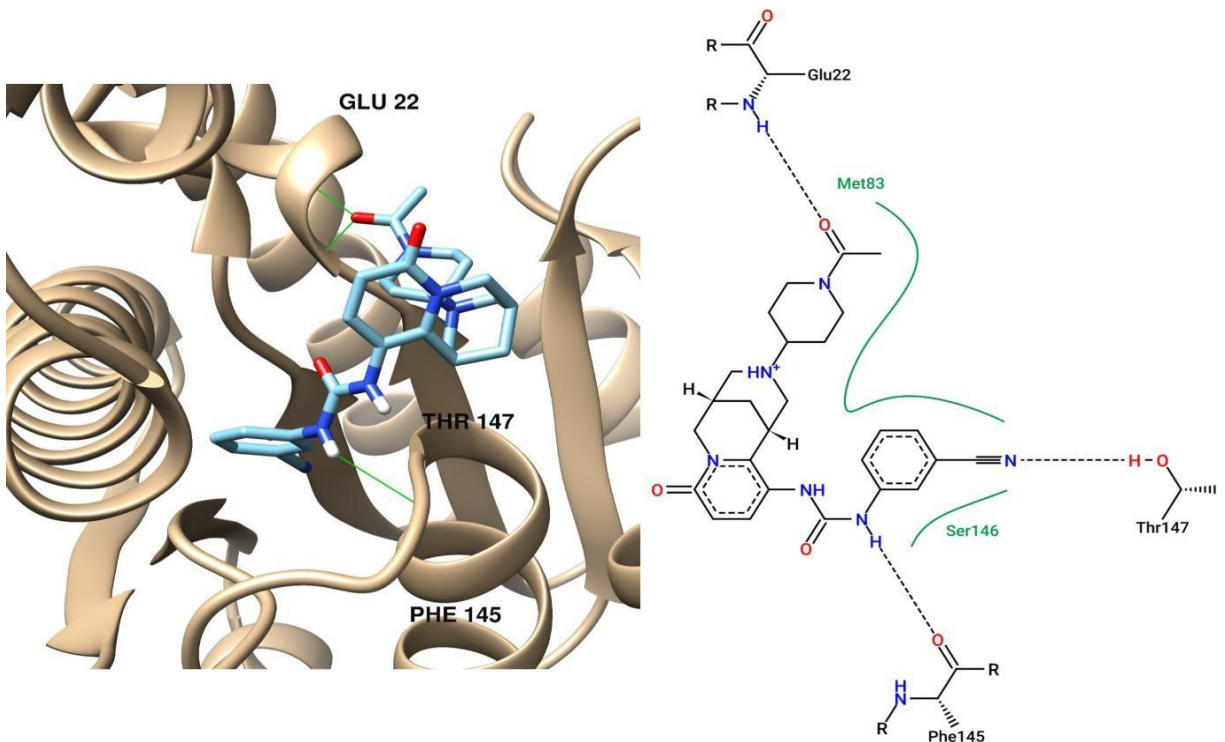


Figure 6. 3D and 2D graphic representation of the docking analysis for the most druggable protein cavity of drug target Uridine monophosphate kinase UMPK with ZINC08300441.

4. Discussion

C. ulcerans can infect humans and a variety of wild and domestic animals that act as a reservoir. Manifestations can be asymptomatic, or cause damage to the respiratory tract, mastitis and gangrenous dermatitis (Tiwari et al. 2008; Hacker et al. 2016). Non-toxigenic strains can be classified as tox-negative or non-toxigenic but *tox* gene-bearing (NTTB) (Fuursted et al. 2015; Dangel et al. 2019). *C. silvaticum* is a recently described NTTB species close to *C. ulcerans* that causes caseous lymphadenitis in wild animals such as wild boar and roe deer, which are reservoirs, and domestic pigs (Dangel et al. 2020). Its potential host range is unknown, and It could potentially cause zoonotic infections (Viana et al. 2020).

Subtractive genomics, as the name suggests, is an approach based on several steps of filtering proteomic sequences, in order to select and identify final proteins that are indispensable to microorganisms and that are not homologous to the host. This methodology is widely used in reverse vaccinology, a method which includes subtractive genomics, modeling, docking and other *in silico* techniques in order to predict vaccine and drug targets, without financial expense and in a short time (Hassan et al. 2018; Masignani et al. 2019). The Bexsero vaccine for *Neisseria meningitidis* (Christodoulides and Heckels 2017) and Engerix-B vaccine for Hepatitis B (Keating and Noble 2003) were developed using these approaches.

We predicted nine vaccine targets, four drug targets and four probable drug molecules for *C. ulcerans* and *C. silvaticum* using reverse vaccinology and *in silico* drug targeting (Tables 1 and 2). Among nine predicted vaccine targets, four were annotated as subunits of ABC transporters. Two are substrate-binding protein for manganese (Mn) and zinc (Zn), one for phosphate, and one is a permease for glutamate. Both metals are vital for several organisms, required by transcriptional transporters and regulators and oxidative stress response (Mn), and several ribosomal proteins and tRNA synthetases (Zn). The acquisition of both metals is disputed with the host and those ABC transporters are required for colonization (Nies and Grass 2009; Juttukonda and Skaar 2015). One of the transporter's subunits (*mntA*) is in PAI 3 (Figure 2). Glutamate is an essential metabolite that plays a relevant role in the metabolism of nitrogen and carbon. The ABC glutamate transporter has already been characterized in other bacterial genera and is a known vaccine target, considering its indispensable role in the survival of the bacterium (Bhatia et al. 2014). Phosphate is essential for energy metabolism and is a component of nucleic acids, phospholipids, and other cell molecules. The Phosphate ABC transporter, substrate-binding protein PstS is part of the complex PstSACB involved in phosphate import (Santos-Beneit 2015). Inactivation of the *pstS* gene increased the susceptibility to penicillin in

Streptococcus pneumoniae (Soualhiné et al. 2005). The transport of these essential nutrients makes those proteins promising vaccine candidates.

Among the non-transporter vaccine target candidates is a bifunctional protein from the Phospholipase/Thioesterase family, with the gene located in pathogenicity island PAI 6 (Figure 2). This enzyme is involved in the non-ribosomal synthesis of peptides, including antimicrobial peptides (Schneider and Marahiel 1998; Santucci et al. 2018). The Integral membrane protein belongs to the “Putative Actinobacterial Holin-X, holin superfamily III”. Holins are produced by double-stranded DNA bacteriophages that utilize an endolysin-holin tactics to lysogenize their hosts (InterPro entry IPR009937). The other three candidates are hypothetical proteins classified by InterProScan as “Papain-like cysteine peptidase superfamily” (InterPro entry IPR038765)(PATRIC ID 996634.5.peg.652), “Calcium-dependent phosphotriesterase” (SSF63829) (996634.5.peg.1276) and a transmembrane protein (996634.5.peg.823).

Within the drug targets, DtxR was the only predicted virulence factor. It is an iron-dependent repressor in prokaryotes that controls the expression of genes involved in iron homeostasis and virulence factors, such as diphtheria toxin (*tox*) (Wagner et al. 2014). DtxR regulates genes according to the concentration of Fe_2^+ ions in the environment, allowing to adjust iron uptake. It is considered a virulence factor due to its importance for host colonization (Merchant and Spatafora 2014). Furthermore, DxtR has been considered an attractive drug target due to its role in regulating genes involved in iron homeostasis in bacteria, and such genes are generally virulence factors (Cheng et al. 2018; Parise et al. 2021). Residue SER 70 is hydrogen bonded to the nitrogen of the aromatic ring, THR 159 is hydrogen linked to the third oxygen of the compound and residues ARG 69 and LEU 135 have hydrophobic interactions with the three aromatic rings.

Uridine Monophosphate Kinase is an important enzyme that acts on the metabolism of pyrimidines in bacteria, specifically on uridine monophosphate (UMP). It converts UMP to UDP that is later used in the nucleoside biosynthesis. UMPK plays an essential role in the production of nucleotides that are constituents of nucleic acids. This enzyme is of interest, as it has been recognized as a potential drug target for tuberculosis (ARVIND et al. 2013) as it also plays a role in regulating the balance of pyrimidine and purine nucleosides. In addition, it is structurally different from the eukaryotic UMP kinases (Rostirolla et al. 2011), which makes it a more attractive candidate. ASP 79 makes two hydrogen bonds with NH (NH-O), GLY 21 makes two hydrogen bonds with the first oxygen of the molecule. The 2D image shows hydrogen bonding of MET 140 to the last nitrogen. MET 83 shows hydrophobic interaction with the last aromatic ring.

The Fructose-1,6-bisphosphate aldolase (FBA) is a glycolytic protein that participates in the glucose metabolic pathway and in the production of organic acid when oxygen is scarce (Altenhoff and Dessimoz 2009; Teramoto et al. 2010). In some genera, It has been shown that reducing FBA leads to cell death. This enzyme is present in the membrane and exposed on the bacterium's cell surface, and its association to virulence is already know as It binds to human plasminogen and mammalian cells (Shams et al. 2014). FBA and other proteins are expressed under conditions of alkaline stress, which indicates that this enzyme is also involved in the physiological breakdown of the bacterium during environmental changes. For these reasons, this protein has previously been recognized as a potential candidate for drug and vaccine targets in various pathogens such as bacteria, fungi and parasites of animals and humans (Pirovich et al. 2021)(Shams et al. 2014). SER 53 is hydrogen bonded to two oxygens (one from the pentose), HIS 212 is hydrogen bonded to an oxygen at the ligand end, and THR 29 is hydrogen bonded to one oxygen (O=S=O). THR 29, THR 33 and GLN 280 have a hydrophobic interaction with naphthalene.

The AcrR family of transcriptional regulators belongs to the one-component system and is associated with several essential cellular mechanisms of bacteria and archaea, such as cell signaling, carbon, nitrogen and lipid metabolism, amino acid metabolism and cofactor metabolism, production of antibiotics, among other physiological factors. They normally act as repressors, having different regulatory mechanisms in different species of bacteria, which can be positive or negative. In addition, these regulators play a key role in antibiotic resistance (Cuthbertson and Nodwell 2013). For these reasons, proteins from this family have been identified as a broad-spectrum drug target (Deng et al. 2013) and for novel treatments (González et al. 2018). ARG 21 has two hydrogen bonds with the oxygen that is attached to the aromatic ring, ARG 92 has a hydrogen bond with an oxygen of the ligand. The two aromatic rings show hydrophobic interactions with VAL 73 and ARG 69.

In this work we presented a first-time approach applied to these bacteria, in which we use common drugs and vaccines for both species. We predicted nine vaccines and four drug target candidates, as well as the respective drug molecules for *C. ulcerans* and *C. silvaticum*. This type of analysis could reduce the cost and time for the development of vaccine and identification of antimicrobial compounds and should be subjected to further experimental validation, as these species can be cultivated. This pipeline is not restricted to *Corynebacterium* but could be used on any prokaryotic or eukaryotic pathogen.

CONFLICT OF INTEREST

None declared.

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SUPPLEMENTARY MATERIAL

Supplementary Table S1. List of 108 genome samples, 72 of *Corynebacterium ulcerans* and 36 of *C. silvaticum*.

Supplementary Table S2. Functional annotation of the drug and vaccine targets shared by *Corynebacterium ulcerans* and *C. silvaticum*

Supplementary Table S3. Druggable pockets in drug targets with score greater than 0.79 predicted using DoGSiteScorer.

Supplementary Figure S1. Ramachandran plot for 3D structure validation of the final 4 drug targets.

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Supporting Information

Supplementary Table S1. List of 108 genome samples, 72 of *Corynebacterium ulcerans* and 36 of *C. silvaticum*.

Species	Strain	Host	Country	Accession number	SRR accession
<i>C. silvaticum</i>	KL0182	Wild boar	Germany	SDQO02	SRR7825394
<i>C. silvaticum</i>	KL0183	Wild boar	Germany	-	SRR7825401
<i>C. silvaticum</i>	KL0259	Wild boar	Germany	-	SRR7825414
<i>C. silvaticum</i>	KL0260	Wild boar	Germany	-	SRR7825417
<i>C. silvaticum</i>	KL0374	Wild boar	Germany	-	SRR7825405
<i>C. silvaticum</i>	KL0382	Wild boar	Germany	-	SRR7825406
<i>C. silvaticum</i>	KL0386	Wild boar	Germany	-	SRR7825407
<i>C. silvaticum</i>	KL0394	Wild boar	Germany	-	SRR7825389
<i>C. silvaticum</i>	KL0395	Wild boar	Germany	-	SRR7825388
<i>C. silvaticum</i>	KL0396	Wild boar	Germany	-	SRR7825387
<i>C. silvaticum</i>	KL0400	Wild boar	Germany	-	SRR7825386
<i>C. silvaticum</i>	KL0401	Wild boar	Germany	-	SRR7825393
<i>C. silvaticum</i>	KL0581	Wild boar	Germany	-	SRR7825370
<i>C. silvaticum</i>	KL0598	Wild boar	Germany	-	SRR7825373
<i>C. silvaticum</i>	KL0615	Wild boar	Germany	-	SRR7825366
<i>C. silvaticum</i>	KL0707	Wild boar	Germany	-	SRR7825360
<i>C. silvaticum</i>	KL0709	Wild boar	Germany	-	SRR7825361
<i>C. silvaticum</i>	KL0773	Wild boar	Germany	-	SRR7825362
<i>C. silvaticum</i>	KL0774	Wild boar	Germany	-	SRR7825363

<i>C. silvaticum</i>	KL0882	Wild boar	Germany	-	SRR7825436
<i>C. silvaticum</i>	KL0883	Wild boar	Germany	-	SRR7825437
<i>C. silvaticum</i>	KL0884	Wild boar	Germany	-	SRR7825442
<i>C. silvaticum</i>	KL0886	Wild boar	Germany	-	SRR7825443
<i>C. silvaticum</i>	KL0887	Wild boar	Germany	-	SRR7825440
<i>C. silvaticum</i>	KL0938	Wild boar	Germany	-	SRR7825434
<i>C. silvaticum</i>	KL0957	Wild boar	Germany	SDVC01	SRR7825429
<i>C. silvaticum</i>	KL0968	Wild boar	Germany	-	SRR7825428
<i>C. silvaticum</i>	KL1003	Wild boar	Germany	-	SRR7825431
<i>C. silvaticum</i>	KL1006	Wild boar	Germany	-	SRR7825430
<i>C. silvaticum</i>	KL1007	Wild boar	Germany	-	SRR7825433
<i>C. silvaticum</i>	KL1008	Wild boar	Germany	-	SRR7825432
<i>C. silvaticum</i>	KL1009	Wild boar	Germany	-	SRR7825425
<i>C. silvaticum</i>	KL1010	Wild boar	Germany	-	SRR7825424
<i>C. silvaticum</i>	KL1196	Deer	Germany	SDVD01	-
<i>C. silvaticum</i>	PO100/5	-	Portugal	CP021417.1	-
<i>C. silvaticum</i>	W25	Wild boar	Germany	VFEM01	-
<i>C. ulcerans</i>	0102	-	-	AP012284.1	-
<i>C. ulcerans</i>	809	-	Brazil	CP002790.1	-
<i>C. ulcerans</i>	2590	Human	Brazil	MPSS01	-
<i>C. ulcerans</i>	4940	Human	Belarus	LSWN01	-
<i>C. ulcerans</i>	05146	-	-	CP009716.1	-
<i>C. ulcerans</i>	131001	-	-	CP010818.1	-
<i>C. ulcerans</i>	131002	-	-	CP011095.1	-
				CP009583.1,	
<i>C. ulcerans</i>	210931	-	-	CP009584.1	-
<i>C. ulcerans</i>	210932	-	-	CP009500.1	-
<i>C. ulcerans</i>	0211	Human	Japan	AP019662.1	-
<i>C. ulcerans</i>	03-8664	Human	France	LGSY02	-
<i>C. ulcerans</i>	04-3911	Human	France	LG SX01	-
<i>C. ulcerans</i>	04-7514	Dog	France	LJVH01	-
<i>C. ulcerans</i>	BR-AD 2649	Dog	Brazil	MPST01	-
<i>C. ulcerans</i>	BR-AD22	-	Brazil	CP002791.1	-
<i>C. ulcerans</i>	FH20161	Human	Japan	AP019663.1	-
<i>C. ulcerans</i>	FRC11	-	France	CP009622.1	-
<i>C. ulcerans</i>	FRC58	-	France	CP011913.1	-
	KZN-2016-				
<i>C. ulcerans</i>	48390	-	South Africa	MIOS01	-
<i>C. ulcerans</i>	LSPQ-04227	Human	Canada	JZUS01	-
<i>C. ulcerans</i>	LSPQ-04228	Human	Quebec	JZUT01	-
<i>C. ulcerans</i>	MRi49	Horse	UK	CP046863.1	-
		Homo sapiens			
<i>C. ulcerans</i>	NCTC 12077	-	UK	AYUJ01	-
<i>C. ulcerans</i>	NCTC7908	-	-	LS483400.1	-
<i>C. ulcerans</i>	NCTC7910	-	UK	LT906443.1	-
<i>C. ulcerans</i>	NCTC8639	-	-	LS483416.1	-
<i>C. ulcerans</i>	NCTC8666	-	-	UFXR01	-
<i>C. ulcerans</i>					
(NTTB)	KL0195	Human	-	-	SRR7825399
<i>C. ulcerans</i>					
(NTTB)	KL0251-cb4	Cat	-	-	SRR7825402
<i>C. ulcerans</i>					
(NTTB)	KL0252-cb5	Cat	-	-	SRR7825415

<i>C. ulcerans</i> (NTTB)	KL0315-cb6	Human	-	-	SRR7825408
<i>C. ulcerans</i> (NTTB)	KL0318-cb7	Dog	-	-	SRR7825409
<i>C. ulcerans</i> (NTTB)	KL0345	Cat	-	-	SRR7825410
<i>C. ulcerans</i> (NTTB)	KL0387-cb8	Human	-	-	SRR7825412
<i>C. ulcerans</i> (NTTB)	KL0392-cb9	Cat	-	-	SRR7825413
<i>C. ulcerans</i> (NTTB)	KL0442	Human	-	-	SRR7825391
<i>C. ulcerans</i> (NTTB)	KL0468	Human	-	-	SRR7825384
<i>C. ulcerans</i> (NTTB)	KL0472	Human	-	-	SRR7825382
<i>C. ulcerans</i> (NTTB)	KL0501	Human	-	-	SRR7825378
<i>C. ulcerans</i> (NTTB)	KL0515	Human	-	-	SRR7825379
<i>C. ulcerans</i> (NTTB)	KL0541	Human	-	-	SRR7825377
<i>C. ulcerans</i> (NTTB)	KL0818	Human	-	-	SRR7825344
<i>C. ulcerans</i> (NTTB)	KL0846	Human	-	-	SRR7825340
<i>C. ulcerans</i> (NTTB)	KL0867	Human	-	-	SRR7825353
<i>C. ulcerans</i> (NTTB)	KL0870	Human	-	-	SRR7825352
<i>C. ulcerans</i> (NTTB)	KL0880	Human	-	-	SRR7825439
<i>C. ulcerans</i> (NTTB)	KL1015	Human	-	-	SRR7825347
<i>C. ulcerans</i> (NTTB)	KL1025	Cat	-	-	SRR7825349
<i>C. ulcerans</i> (tox-)	KL0199	Human	-	-	SRR7825398
<i>C. ulcerans</i> (tox-)	KL0349	Human	-	-	SRR7825411
<i>C. ulcerans</i> (tox-)	KL0451	Human	-	-	SRR7825390
<i>C. ulcerans</i> (tox-)	KL0853	Human	-	-	SRR7825339
<i>C. ulcerans</i> (tox-)	KL0876	Human	-	-	SRR7825438
<i>C. ulcerans</i> (tox-)	KL0941	Human	-	-	SRR7825435
<i>C. ulcerans</i> (tox-)	KL1017	Cattle	-	-	SRR7825348
<i>C. ulcerans</i> (tox+)	08-1143-CB1	Pig	-	-	SRR7825397
<i>C. ulcerans</i> (tox+)	KL0126-cb2	Human	-	-	SRR7825396
<i>C. ulcerans</i> (tox+)	KL0194	Human	-	-	SRR7825400
<i>C. ulcerans</i> (tox+)	KL0246-cb3	Human	-	-	SRR7825403

<i>C. ulcerans</i> (tox+)	KL0350	Human	-	-	SRR7825404
<i>C. ulcerans</i> (tox+)	KL0433	Human	-	-	SRR7825392
<i>C. ulcerans</i> (tox+)	KL0475	Human	-	-	SRR7825383
<i>C. ulcerans</i> (tox+)	KL0483	Human	-	-	SRR7825380
<i>C. ulcerans</i> (tox+)	KL0497	Human	-	-	SRR7825381
<i>C. ulcerans</i> (tox+)	KL0540	Human	-	-	SRR7825376
<i>C. ulcerans</i> (tox+)	KL0547	Human	-	-	SRR7825374
<i>C. ulcerans</i> (tox+)	KL0556	Human	-	-	SRR7825375
<i>C. ulcerans</i> (tox+)	KL0785	Human	-	-	SRR7825346
<i>C. ulcerans</i> (tox+)	KL0796	Human	-	-	SRR7825345
<i>C. ulcerans</i> (tox+)	KL0825	Human	-	-	SRR7825343
<i>C. ulcerans</i> (tox+)	KL0832	Human	-	-	SRR7825342
<i>C. ulcerans</i> (tox+)	KL0840	Human	-	-	SRR7825341

Supplementary Table S2. Functional annotation of the drug and vaccine targets shared by *Corynebacterium ulcerans* and *C. silvaticum*

GenBank ID	GIPSy	VFDB	PBIT	eggNOG-mapper			
Target ID	Gene	Product	Genomic island	Virulence Factor	Virulence Factor	Preferred name	Description
Vaccine CULC010 - 2_0158		ABC transporter, substrate-binding protein(cluster 14, Mn/Zn)	-	-	Yes	<i>troA</i>	Belongs to the bacterial solute- binding protein 9family
Vaccine CULC010 - 2_0597	<i>mntA</i>	ABC transporter, substrate-binding protein(cluster 14, Mn/Zn)	PAI 3	-	Yes	<i>mntA</i>	Belongs to the bacterial solute- binding protein 9 family
Vaccine CULC010 - 2_0803	<i>gluC</i>	Glutamate ABC transporter, permease protein 1 GluC	-	-	Yes	<i>gluC</i>	ABC-type amino acid transport system permease component
Vaccine CULC010 - 2_0624		Hypothetical protein	-	-	Yes		Cell wall-associated hydrolases
Vaccine CULC010 - 2_0791		Hypothetical protein	-	-	No	<i>yccF</i>	membrane
Vaccine CULC010 - 2_1236		Hypothetical protein	-	-	No	<i>lppL</i>	amine dehydrogenase activity

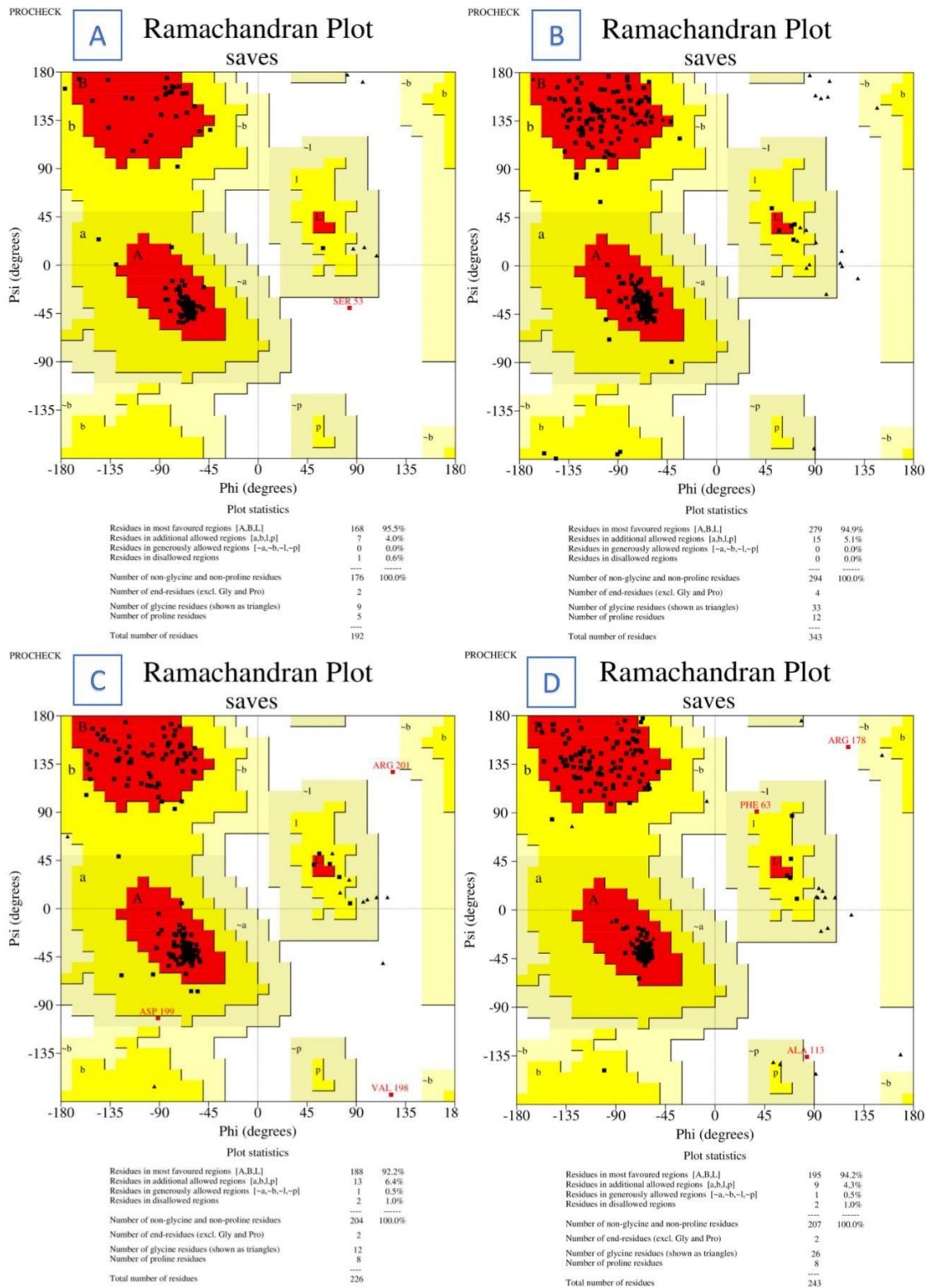
Vaccine	CULC010 - 2_0305	Integral membrane protein	-	-	No		Putative Actinobacterial Holin-X, holin superfamily III
Vaccine	CULC010 <i>pstS</i> 2_1930	Phosphate ABC transporter, substrate-binding protein PstS (TC3.A.1.7.1)	-	-	No	<i>pstS</i>	Part of the ABC transporter complex PstSACB involved in phosphate import
Vaccine	CULC0102- _1070	Phospholipase/thioesterase	PAI 6	-	No	<i>pla</i>	Cutinase
Drug	CULC0102 <i>dtxR</i> _1442	Iron-dependent repressor IdeR/DtxR	-	Yes	Yes	<i>ideR</i>	repressor
Drug	CULC0102 <i>fba</i> _2084	Fructose-bisphosphate aldolase class II	-	No	No	<i>fbaA</i>	aldolase
Drug	CULC0102 <i>pyrH</i> _1536	Uridine monophosphate kinase	-	No	No	<i>pyrH</i>	Catalyzes the reversible phosphorylation of UMP to UDP
Drug	CULC0102- _1260	Transcriptional regulator, AcrR family	-	No	No	<i>acnR</i>	transcriptional

Supplementary Table S3. Druggable pockets in drug targets with score greater than 0.79 predicted using DoGSiteScorer.

DoGSiteScorer results						
GenBank ID	Name	Volume (Angstrom3)	Surface Area (Angstrom2)	Drug Score	Active Residues	InterProScan Domain Active Residues
CULC0102_1442	Iron-dependent repressor IdeR/DtxR	979,26	1337,53	0,79	TYR 16, GLU 19, GLU 20, GLU 21, GLY 22, VAL 23, THP 24, PRO 25, MET 64, ARG 69, THR 70, THR 73, ALA 74, ARG 77, ASP 110, PRO 131, PRO 133, GLY 134, GLU 137, LEU 138, SER 158, MET 159, PRO 160, ARG 161, LYS 162, VAL 193, ILE 195, SER 205, HIS 206, ASN 207, GLY 208, ASP 210.	GLU83, HIS79, HIS98, MET10, CYS102, GLU105, HIS106, LEU4

CULC0102_1536	Uridine monophosphate kinase (EC 2.7.4.22)	823,24	1117,43	0,85	VAL 212A, LEU 213A, GLY 216A, LEU 217A, ARG 218A, VAL 219A, ALA 220A, ALA 222A, PHE 225A, SER 226A, MET 229A, ASP 166B, ALA 169B, ALA 170B, VAL 181B, MET 183B, LYS 185B, VAL 187B, GLY 189B, VAL 190B, PHE 191B, ALA 192B, GLU 193B, ASP 194B, PRO 195B, VAL 207B, VAL 212B, LEU 213B, GLY 216B, LEU 217B, ARG 218B, VAL 219B, ALA 220B, ASP 221B, ALA 222B, ALA 224B, PHE 225B, SER 226B, CYS 228B, MET 229B, ILE 235B, VAL 237B, THR 257B, VAL 259B.	GLU135,HIS145, GLN132, ALA113, GLY131, ARG123, ARG132, LYS130, ARG126, ARG123, PRO118, LEU120, LEU122,PRO121
CULC0102_2084	Fructose-bisphosphate aldolase classII (EC 4.1.2.13)	478,72	877,8	0,86	ALA 23, ASN 27, GLN 51,SER 53, ASP 95, HIS 96,ASN 210, VAL 211, HIS212, GLY 213, VAL 214, TYR 215, LYS 216, PRO217, GLY 218, ASN 219,VAL 220, LEU 222, PHE251, HIS 252, GLY 253,GLY 254, SER 255, GLY256, SER 257, LEU 258,GLU 261, ASN 274, VAL275, ASP 276, THR 277,ZN 352.	HIS96, HIS212, HIS252
CULC0102_1260	Transcriptional regulator, AcrR Family	721,15	705,54	0,84	SER 162A, ARG 163A, THR 166A, GLY 167A, ALA 168A, ALA 6B, ASP66B, ALA 67B, MET 70B,TRP 95B, MET 96B, ARG99B, LEU 100B, LEU106B, ARG 113B, TRP116B, ILE 117B, GLN120B, LEU 123B, ASP 124B, VAL 127B, ARG 128B, GLU 147B, HIS150B, THR 151B, GLU154B, LEU 157B, ASP158B, ILE 161B, SER162B.	LEU153, LEU131, LEU149, VAL184, MET140, ASN134, ARG130, GLU181, ARG185

Supplementary Figure S1. Ramachandran plot for 3D structure validation of the final 4 drugtargets.



IV. CAPÍTULO II

ARTIGO DE PESQUISA

Neste capítulo, foi feito um primeiro estudo de desenvolvimento in silico de vacinas de subunidades para a prevenção de doenças causadas por *C. ulcerans* e *C. silvaticum*. O design de 4 vacinas multi-epítipo foi feito por meio da imunoinformática, a partir das 9 proteínas encontradas na pesquisa do Capítulo I. Experimentos in silico de docking molecular, simulação imune, clonagem e expressão em vetor bacteriano foram realizados e mostraram bons resultados que levaram à proposição de 4 vacinas candidatas para a proteção de doenças causadas por ambas as bactérias citadas.

Pan-vaccinomics approach toward a universal multi-epitope based vaccines candidate against *Corynebacterium ulcerans* and *C. silvaticum*

Janaína Canário Cerqueira^a, Lucas Gabriel Rodrigues Gomes^a, Marcus Vinicius Canário Viana^a, Arun Kumar Jaiswal^a, Thaís Cristina Vilela Rodrigues^a, Siomar Castro Soares^b, Debmalya Barh^a, **Vasco Azevedo^{a*}** and Sandeep Tiwari^{a*}

O artigo de pesquisa foi submetido na revista **Immunoinformatics** em 25/05/2022.

Pan-vaccinomics approach toward a universal multi-epitope based vaccines candidate against *Corynebacterium ulcerans* and *C. silvaticum*

Janaína Canário Cerqueira^a, Lucas Gabriel Rodrigues Gomes^a, Marcus Vinicius Canário Viana^a, Arun Kumar Jaiswal^a, Thaís Cristina Vilela Rodrigues^a, Siomar Castro Soares^b, Debmalya Barh^a, Vasco Azevedo^{a*} and Sandeep Tiwari^{a*}

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Abstract: Zoonotic diseases have been rising in underdeveloped countries with low vaccination rates, posing risks to human and cattle populations. Diphtheria, one such disease, is mainly caused by *C. ulcerans*, which can also cause other diseases of the respiratory system, against which the current diphtheria vaccine (DTPa) does not protect. The recently described species *C. silvaticum*, of the same genus, is cytotoxic in the human epithelium and has zoonotic potential. The objective of this research was to use immunoinformatics to propose multi-epitope vaccines for both species, which are composed of selected epitopes from proteins essential to both bacteria. The designed multi-epitope vaccine contains 25 common epitopes from 8 proteins from both species with great predicted immunological properties. These constructs' stability, physicochemical properties, allergenicity, antigenicity, and toxicity were stipulated. In addition, docking and molecular dynamics simulations were performed to certify the link between TLR2 and each constructed vaccine. Immune stimulation and cloning in pET- 28a(+) vector in *E. coli* were performed. The results demonstrated that these four vaccine constructs might induce innate and adaptive immune responses to protect against *C. ulcerans* and *C. silvaticum*. However, experimental validation is still required to prove the *in silico* models predicted here *in vitro* and *in vivo*.

Keywords: *Corynebacterium*, immunoinformatics, multi-epitope vaccine, pan-vaccinomics, zoonotic pathogen.

1. INTRODUCTION

The genus *Corynebacterium* comprises Gram-positive and coryneform bacteria, with more than 90 described species. About 54 species are of medical importance worldwide, 14 of which are zoonotic bacteria infecting reservoirs, usually mammals, ranging from cattle to pets [1].

C. ulcerans is one of these zoonotic pathogens, the leading cause of diphtheria, even more than *C. diphtheriae*, the principal producer of diphtheria toxin (DT). Some *C. ulcerans* produce DT, which causes diphtheria in humans, while others produce other virulence factors such as the well-known PLD (Phospholipase D), which kills macrophages. *C. ulcerans* causes respiratory system diseases such as pneumonia, tonsillitis, pharyngitis, and other related diseases and is commonly transmitted from farm animals. Infection and disease cases have been increasing in underdeveloped countries despite vaccination with the DTPa immunizer to prevent diphtheria, tetanus, and pertussis [2].

C. silvaticum is a new species described in 2020 that infects roe deer, wild boar, and pigs [3]. This bacterium is genetically related to *C. ulcerans* and another pathogen, *C. pseudotuberculosis*. Recent genomic and *in vivo* studies have shown the ability of this organism to infect the human epithelium, demonstrating its zoonotic potential and relevance to human health [4,5].

The circumstances above point to the economic importance of these pathogens in the meat, milk, and animal reproduction sectors and consequently their importance in public health, demonstrating the need for their prevention and control. Through computational tools and the use of genomics-based approaches, it is possible to analyze the genomes of these species to answer questions and propose strategies to solve problems related to infectious agents.

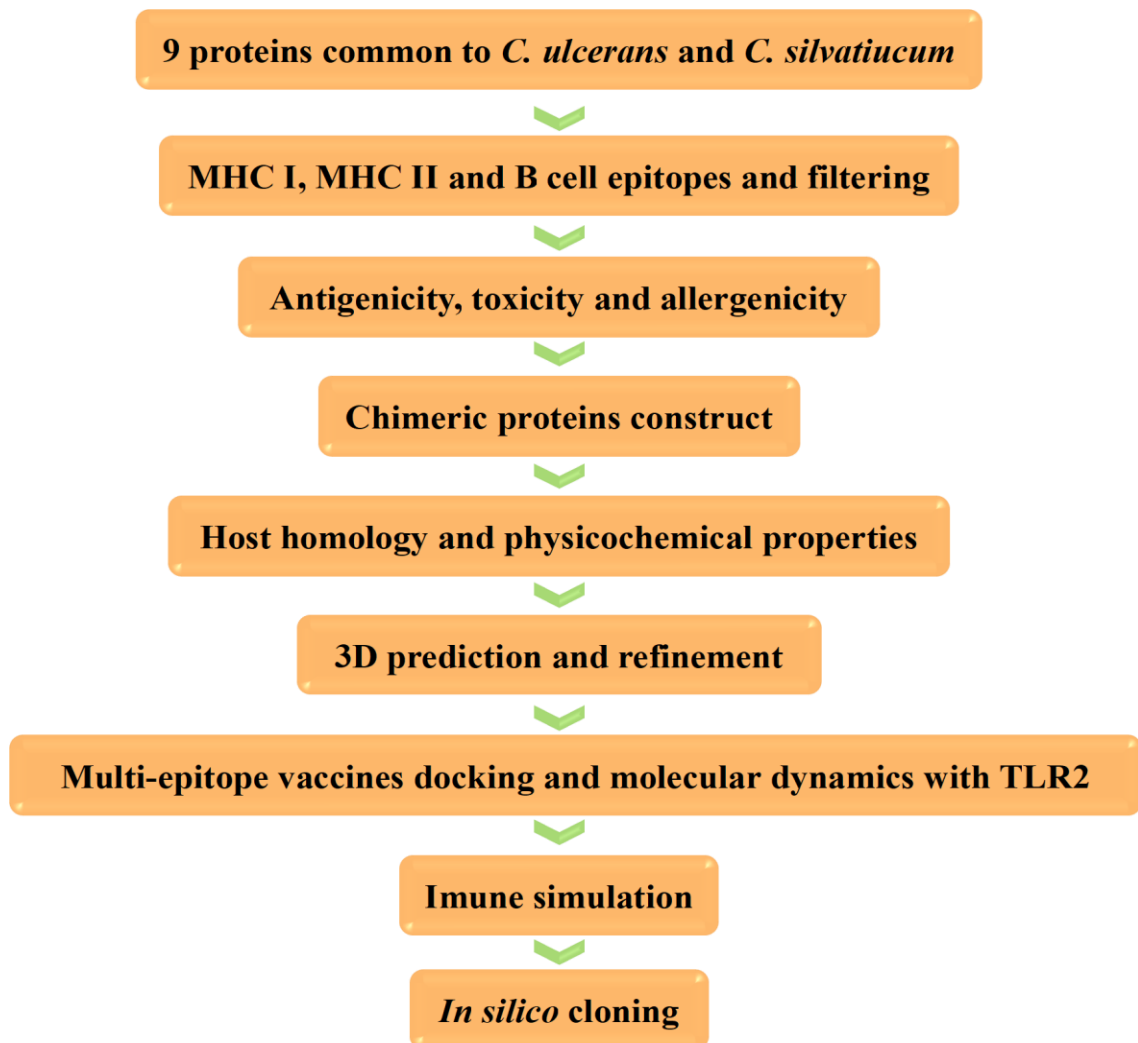
Subtractive genomics and immunoinformatics are *in silico* methods that contribute to vaccinological research and vaccine design. Such methods allow identifying target proteins of drugs and vaccine candidates and constructing chimeric proteins (multi-epitope vaccines) based on these candidates. Reverse vaccinology and immunoinformatics make it possible to propose new vaccines quickly, with the low cost necessary for their development [6,7]. This methodology has been previously used in the development of control methods for various microorganisms, such as the important causative agents of yellow fever [8], schistosomiasis [9], leprosy [10], pneumonia [11], and syphilis [12].

The objective of this work was to propose four multi-epitope vaccines for the control and prevention of zoonoses caused by *C. ulcerans* and *C. silvaticum*. These vaccines have been

designed with immunogenic epitopes derived through vaccine candidates common to both species and have displayed good immunogenic potential. However, the experimental validation of these constructs *in vivo* and *in vitro* is still necessary to ensure their function.

2. MATERIALS AND METHOD

The overall workflow used in this study is shown in **Figure 1**.



2.1 Selection and retrieval of core antigenic proteins

The nine core genomes, non-host homologous proteins, and antigenic properties derived from *C. ulcerans* and *C. silvaticum* strains were selected from our previous work (Cerqueira et al. 2022, unpublished), in which common drug targets and vaccine candidates from these

species were identified using 108 genomes of both species. The characteristics of these nine vaccine candidates' proteins are shown in Table 1. The FASTA sequences of these proteins were retrieved from the National Center for Biotechnology Information (NCBI) database.

Table 1. Proteins selected for the multi-epitope analysis

Name	NCBI ID
ABC transporter, substrate-binding protein (cluster 14, Mn/Zn)	WP_013910564
Integral membrane protein	WP_013910876
ABC transporter, substrate-binding protein (cluster 14, Mn/Zn)	WP_046693528
Hypothetical protein	WP_014836045
Phospholipase/thioesterase	WP_014836279
Hypothetical protein	WP_014836377
Hypothetical protein	WP_014525496
Phosphate ABC transporter PstS	WP_014836748
Glutamate ABC transporter, permease protein 1 GluC	WP_013911078

2.2 Prediction of B cell, MHC class I, and MHC class II epitopes

We used the server ABCpred [13] for the B cell epitope prediction and the Immune Epitope Database and Analysis Resource (IEDB) server [14] for the MHC class I and class II T cell epitope predictions based on mouse alleles for the nine proteins sequences previously selected (**Table 1**). The data were filtered based on Ic_{50} lower than 500nM and percentile rank lower than 1.0 for MHC class I and lower than 5.0 for MHC class II epitopes. Default parameters were used in both predictors.

2.3 Epitope screening

The selected epitopes from IEDB and APCpred were then filtered based on antigenicity, allergenicity, and toxicity by the software VaxiJen v2.0 [15], AllerTOP 2.0server [16], and ToxinPred [17], respectively. Afterward, the screened peptides were subjected to an overlap analysis performed by an *in-house* python script made by our research group. The overlapping was made for MHC-I epitopes against B cell epitopes and MHC-II against B cell epitopes, and redundant epitopes were excluded.

2.4 Construction of the chimeric protein (vaccine)

Four adjuvants (**Table 2**) were selected to increase the immunogenicity of multi-epitope vaccines and added to the chimeric protein sequence. The adjuvants were 50s ribosomal L7/L12 protein, Beta-defensin, HBHA protein (*M. tuberculosis*), and HBHA conserved sequence [18]. We constructed four chimeric proteins (V1, V2, V3, and V4), each with an adjuvant sequence. For the vaccine construct, linker sequences were used for joining the MHC-I/B and MHC-II/B epitopes: AAY and GPGPG, respectively. The EAAAK linker was used to attach the adjuvant to the first epitope of the chimeric protein[11]. We also placed a histidine tag (HHHHHH) at the end of the sequences.

2.5 Prediction of antigenicity, allergenicity, solubility, and host homology of the vaccine construct

The solubility of these four vaccines was calculated using Protein-Sol [19] and SOLpro [20] servers. The AllerTOP 2.0 server [16] was used to predict allergenicity and VaxiJen 2.0 [15] for antigenicity prediction with a threshold of 0.4 (for the bacteria model). Homology analysis with typical host proteomes (*Bos taurus* taxid:9913, *Sus scrofa* taxid:9823, *Ovis aries* taxid:9940, *Equus ferus caballus* taxid:9796, *Capra aegagrus hircus* taxid:9925 and *Homo sapiens* taxid: 9606) was performed by BlastP at the National Center for Biotechnology Information (NCBI) with standard databases and "BLASTp" algorithm.

2.6 Prediction of tertiary structure and refinement

We used AlphaFold v2.0 Colab [21] to predict the tertiary structure of the vaccine. After that, we used the GalaxyRefine available at the GalaxyWEB server [22] to make the refinement in the 3D structure. To analyze the quality of the structure of the chimeric protein, we performed a Ramachandran plot analysis using the PROCHECK program available on the SAVES web server [23].

2.7 Molecular docking between the chimeric protein and the TLR2

We performed molecular docking with TLR2 (Toll-Like Receptor) and the multi-epitope vaccines. The TLR2 is commonly used in *Corynebacterium* research. The TLR structure (**PDB ID: 2Z81**) was retrieved from the PDB database, and the Chimera software [24] was used to

remove water molecules, ligands, and other additional molecules. We performed molecular docking between the TLR2 and each chimeric protein using the ClusPro web server [25]. To determine and visualize the interactions in the complex, we used LigPlot+ [26].

2.8 Molecular dynamic simulation of the receptor-ligand complex

The iMODS server was used to evaluate the flexibility and stability of vaccine- TLR2 complexes. The program uses Normal Mode Analysis (NMA) to predict dynamic characteristics such as eigenvalue, deformability, and protein variance [27].

2.9 Vaccine Immuno simulation

We used the CImmSim server [28] to evaluate the immune response of the multi-epitope vaccines. To this end, three injections were given over four weeks (each one). Each inoculum carried 1000 vaccine proteins and the periods were 1, 84, and 168 (each period means 8h in real life, and time '1' is the initial inoculation time = 0). The simulation steps of 1050 were used, keeping the other parameters as default.

2.10 *In silico* cloning in plasmid

We carried out *in silico* cloning to investigate the cloning and expression potential of the vaccines in a plasmid vector. The peptide sequences of the vaccines were converted into their coding DNA sequences, optimizing codon usage for expression in *Escherichia coli*. The Java Codon Adaptation Tool (JCat) [29] program was used for this adaptation, using default settings and selecting the *E. coli* K12 strain. To ensure the quality of expression in the *E. coli* expression system, the Codon Adaptation Index (CAI) of the sequences should be above 80%, and the GC content rate should be between 30 to 70%. For the *in silico* cloning, the SnapGene software (from Insightful Science, available at snapgene.com) was used, and the *E. coli* pET28a(+) vector was selected as the sequence. Restriction sites for the *SacI* and *BamHI* restriction enzymes were inserted into the coding sequences of the vaccine constructs, and the optimized sequences were inserted into the vector through the SnapGene software to ensure the expression of the vaccine.

3. RESULTS

3.1 Predicted and filtered MHC class I, MHC class II and cell B epitopes

We predicted epitopes in the IEBD (MHC I: 20648 and MHC II: 7581) and ABCpred (B cell: 265) predictors for the 9 proposed proteins. The predictions were based on mouse alleles, with a size of 9 amino acids for MHC and a size of 16 for B cells. Epitopes with Ic_{50} values below 500nM were excluded. Afterward, the epitopes with percentile rank values below 1.0 and 5.0 were selected for MHC I and MHC II. 1198 epitopes (B/MHC-I/MHC-II) after these filters remained. These 1198 epitopes were further analyzed to filter based on criteria antigenicity, non-allergenicity, overlapping between the filtered epitopes of MHC II against B cells and MHC II against B cells, followed by removal of redundancy to get the best epitopes.

3.2 Antigenicity, allergenicity, and toxicity analysis

Antigenicity, allergenicity, and toxicity analyzes were performed using standard parameters. Epitopes that did not fit these criteria were excluded, the remaining 120. Overlapping between the filtered epitopes of MHC II against B cells and MHC II against B cells was performed to remove redundant peptides, decreasing the number of 76 epitopes. Similar peptides remained after the overlapping; they were excluded based on lower percentile rank. The integral membrane protein (WP_013910876) did not pass these selected filter criteria and was excluded from further analysis. As a result, the final immunogenic epitopes of 18 MHC-I and 7 MHC-II epitopes from eight proteins were used for the final candidate's multi-epitope vaccine construction (**Table 2**).

Table 2. Final epitopes sequences for the vaccine construct.

Proteins	MHC I/B epitopes	Percentile rank	MHC II/B epitopes	Percentile rank
ABC transporter, substrate-binding protein (cluster 14, Mn/Zn)	DTYTGALRY	0.57	VPVIFQDASANPQAV	1.62
ABC transporter, substrate-binding protein (cluster 14, Mn/Zn)	PEFTHDPHI	0.45		
Hypothetical protein	CVPLFHA AV	0.45	GTPYQWGGTSTAGFD	0.89
	AALLLPRVF	0.48	NPEPLEPNTPNVNES	1.65
	GTQVTHEQL	0.86		
Phospholipase / thioesterase	GSKTGDFDRF	0.52	VGIPAAGAQAQCPR	2.25

	NSATTTTGY	0.58		
	KVGAPIWWL	0.87		
Hypothetical protein	ARVNNEHTI	0.42	STSGVQRTAIAGLLA	1.81
	SATGDGLQI	0.98		
	LPCPDGVHV	0.60		
	ASDTIGGQL	0.92		
Hypothetical protein	GIPLAIANI	0.91	STSSSNPLAG	2.60
	VIWFLVAGL	0.39		
Phosphate ABC transporter PstS	YDQTTRDQV	0.81	ETGSSSGAGAST	2.75
	EKDAYPLVL	0.86		
	LPFVIGPVA	0.89		
Glutamate ABC transporter, permease protein 1 GluC	RSGINTVDF	0.29		

3.3 Multi-epitope vaccine construct and host homology

Four chimeric proteins were constructed using identical epitope sequences with different adjuvants (**Figure 2**). The four adjuvants used were: 50s ribosomal L7/L12 protein, Beta-defensin, HBHA protein (*M. tuberculosis*), and HBHA conserved sequence. Each adjuvant was followed by the EAAAK linker and then the epitopes. AAY peptides linked the 18 CTL/B epitopes, GPGPG linked the 7 HTL/B epitopes, and 6 tagged histidine residues were added at the end. The schematic of these four multi-epitope vaccine constructs is shown (V1: 518 amino acids; V2: 509 amino acids; V3: 404 amino acids, and V4: 489 amino acids) in Figure 2. In the host homology analysis, all four multi-epitope vaccine constructs did not show any homology with the host proteome.

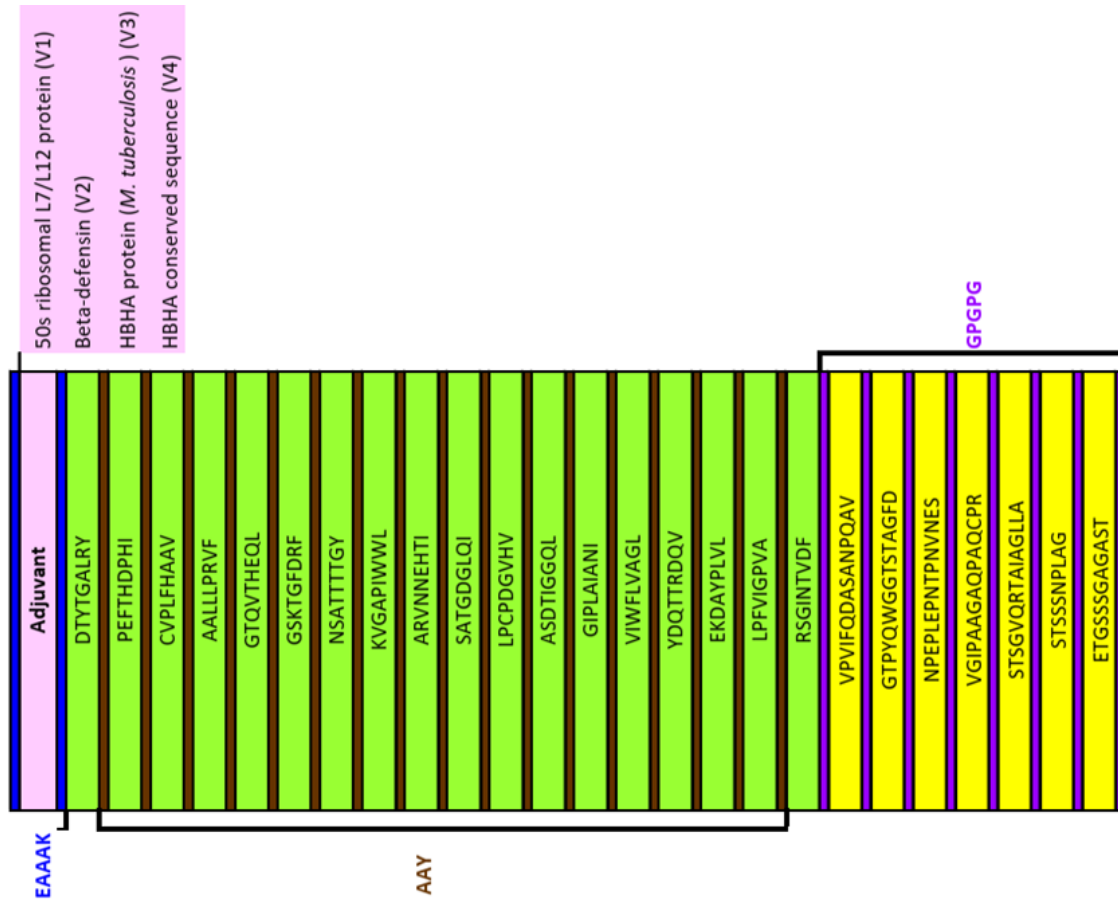


Figure 2. Construction of multi-epitope vaccines. The green boxes represent the MHC-I epitopes, and the yellow boxes represent the MHC-II epitopes. In the boxes are the epitopes interspersed with the linkers; in the first box, each adjuvant is shown laterally (light pink).

3.4 Physical-chemical and structural properties

The ProtParam server analyzed the physicochemical properties of constructed multi-epitope vaccines. The vaccines showed a molecular weight between 41 to 54 kDa (Table 4). All vaccines are non-allergenic, antigenic, non-toxic, and soluble. The V3 construct had a pI value of 8.08, which differs from the other 3 constructs with similar pI values, between 4.98-5.06. V1, V2, and V3 showed negative GRAVY results indicating they are hydrophilic and capable of interacting with water molecules. V4 showed a positive GRAVY value but very close to zero. The instability index was below 40.0 for all vaccines, which indicates stability in biological processes. The aliphatic index (AI) ranged from 72.75 to 81.78, indicating high-temperature stability.

We obtained the best models from AlphaFold and subjected them to refinement. As shown in figure 3, the four vaccines showed good structural quality with expected values - of residues above 90% in more favorable regions (Table 3). After refinement, V1 showed 90.8%

of the residues in the most favorable regions and 1.6% in the disallowed regions, V2 showed 92.8% of the residues in the most favorable regions and 0.5% in the disallowed regions, V3 showed 92.7% of the residues in the most favorable regions and 0.9% in disallowed regions, and V4 showed 96.5% of residues in more favorable regions and 0.5% in disallowed regions.

Table 3. Physicochemical properties of multi-epitope vaccines. **MW:** Molecular Weight; **pI:** isoelectric pH; **GRAVY:** Grand Average of Hydrophaticity. NA = non-allergen.

Name	Antigenicity	Solubility	MW	pI	Allergenicity	GRAVY	AI
V1	0.7995	0.350/0.979565	54012.04	5.06	NA	-0.152	79.02
V2	0.8046	0.390/0.989099	52893.81	4.98	NA	-0.127	80.96
V3	0.8747	0.287/0.829559	41544.59	8.08	NA	-0.078	72.75
V4	0.7635	0.511/0.984098	49823.87	5.05	NA	0.048	81.68

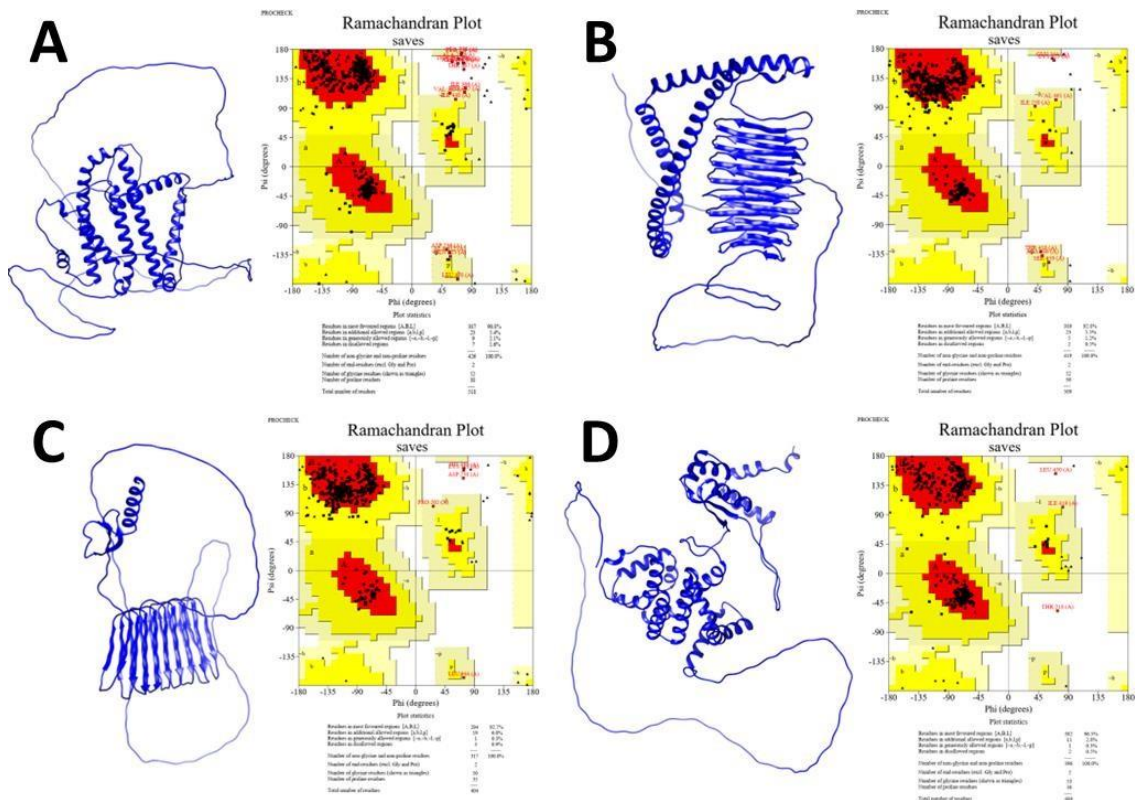


Figure 3. The refined 3D structure and Ramachandran plot of multi-epitope vaccine construct. **A:** The Ramachandran plot for V1 construct after refinement shows 90.8% residues in the most favored regions and 1.6% in disallowed regions.; **B:** The Ramachandran plot for V2 construct after refinement, showing 92.8% residues in most favored regions and 0.5% in disallowed regions.; **C:** The Ramachandran plot for V3 construct after refinement shows 92.7% residues in most favored regions and 0.9% in disallowed regions.; **D:** The Ramachandran plot for V4 construct after refinement shows 96.5% residues in most favored regions and 0.5% in disallowed regions.

3.5 Molecular docking with TLR2 and MD simulation

After the docking analysis, the UCSF Chimera software was used to analyze the top 5 democratic docking models for each docked multi-epitope with a TLR2 complex. The best-docked complex was selected based on the lowest binding energy (stable), the highest number of hydrogen bonds, hydrophobic interaction, and other intermolecular interactions. The docking result showed binding energy ranging from -1230.7 to -1410.7 with the highest number of intermolecular interactions, which suggests a good interaction between TLR2 multi-epitope vaccine constructs (**Table 4, Figures 4 and 5**). Furthermore, these best complexes were subjected to the iMOD server for the molecular dynamic simulation analysis. The Normal Mode Analysis (NMA) was used, a computer method for analyzing the flexibility of the protein. The B-factor graph depicts the complex's link between the Normal Mode Analysis and PDB regions. The eigenvalues of all docked complexes (**Table 4 and Supplementary Figure 2**).

Table 4. The docked properties of the best-selected vaccine -TLR2 complexes and iMOD eigenvalues.

Vaccine	Energy	Hydrogen Bonds	Hydrophobic Bonds TLR2	Hydrophobic Bonds Vaccine	Note	Eigenvalues
V1	-1410.7	18	25	18	2 salt bridge	3.895739 e-09
V2	-1230.7	36	22	20	2 salt bridge	1.892432 e-06
V3	-1284.8	6	25	8	NA	4.953906 e-08
V4	-1372.6	16	32	13	2 salt bridge	3.509202 e-07

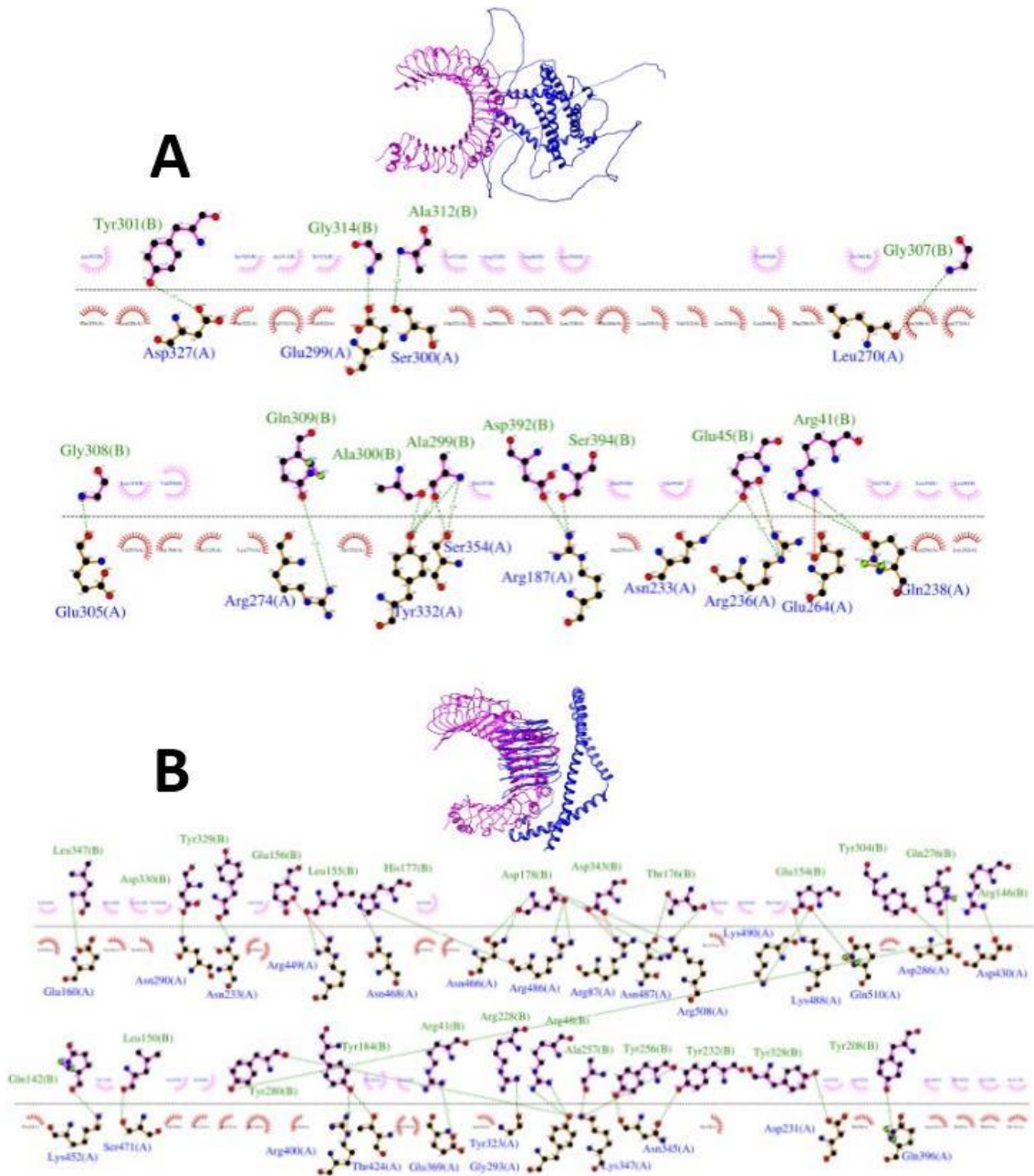


Figure 4. 2D and 3D representation of TL2-vaccine interactions. **A:** V1-TLR2 complex showed 18 hydrogen bonds. Also, 18 residues of V1 and 25 residues of TLR2 are involved in the hydrophobic interactions, and 2 residues for salt bridge formation. **B:** V2-TLRD complex showed 36 hydrogen bonds. Also, 20 residues of V2 and 22 residues of TLR2 are involved in the hydrophobic interactions, and 2 residues for salt bridge formation.

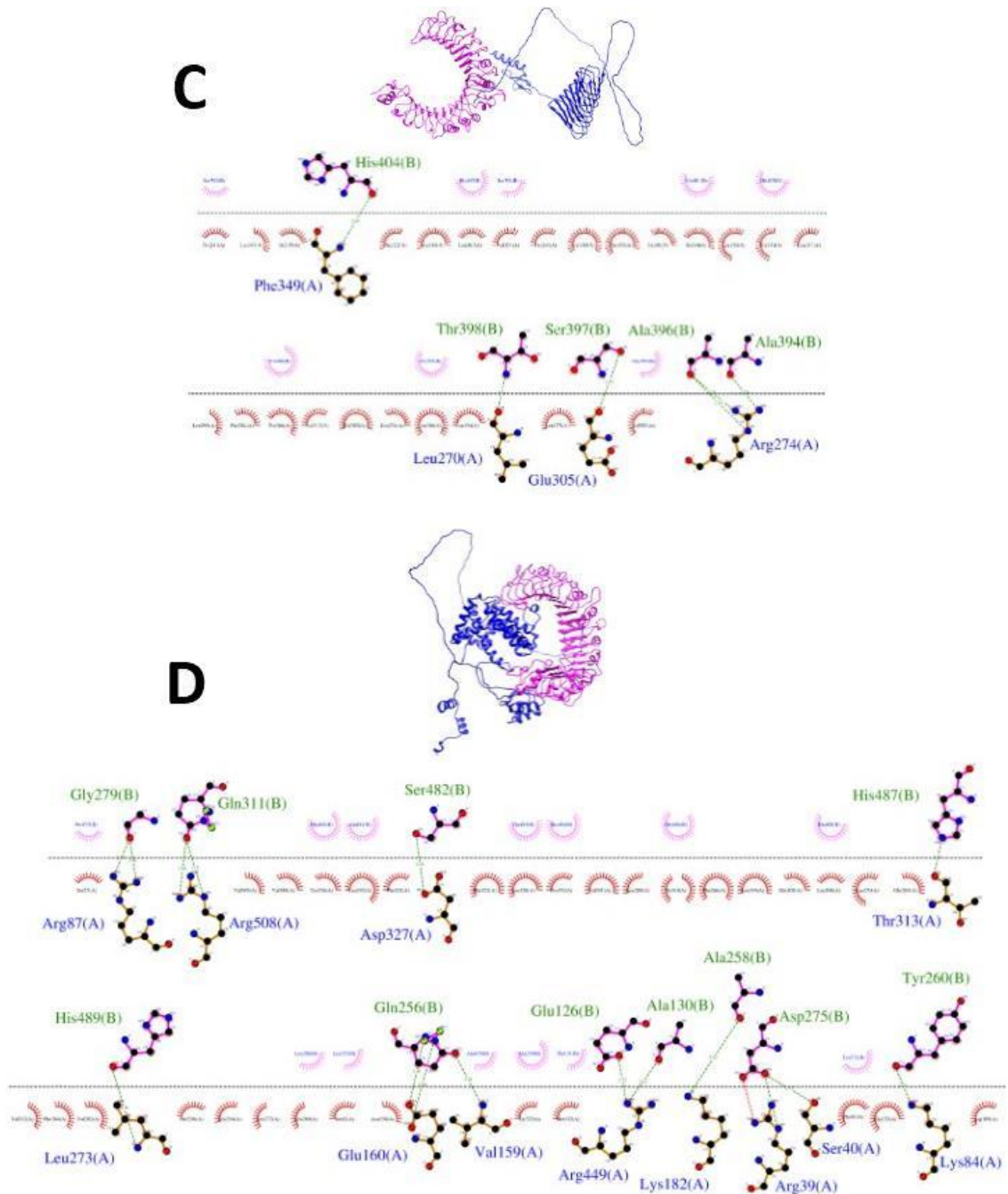


Figure 5. 2D and 3D representation of TL2-vaccine interactions. **C:** V3-TLR2 complex showed 6 hydrogen bonds. Also, 8 residues of V3 and 25 residues of TLR2 are involved in the hydrophobic interactions. **D:** V4-TLR2 complex showed 16 hydrogen bonds. Also, 13 residues of V4 and 32 residues of TLR2 are involved in the hydrophobic interactions, and 2 residues for salt bridge formation.

3.6 Vaccines Immune simulation

The Immune simulation analysis showed the four vaccines to induce immune responses with similar results. The simulation result for the multi-epitope vaccine (V1) is shown in figure 6. The result for other multi-epitope vaccines is shown in **supplementary material figures 3-5**.

The complexes were able to activate and increase B cell and T cell populations after the first injection, showed in graphs A-C. Peak activation and population growth of B and T cells are observed between 50 and 100 days after injections.

The B cells differentiated into isotypes for immunoglobulin expression and memory cells (total in black line), which supposedly confirms the recognition of immunogens by the defense cells (Figure 6A). In the same figure in B, a significant increase in active T cells cytotoxic (TC) (pink) is observed after the first injection until about day 100, in contrast to the low number of deactivated cells (blue). After this period of infection, activated cells decrease over time. Helper T cells (TH) are also in high concentrations (total black). Memory TH cells decrease a little over time but maintain a certain blood concentration even after 350 days. TH cells help activate other defense cells such as B cells, TC cells, and macrophages. In D, activation and increase of active macrophages are also observed between days 50 and 100 (blue), and their concentration is maintained for an extended period (after 350 days). The total number of macrophages (black) is increased from the first inoculum.

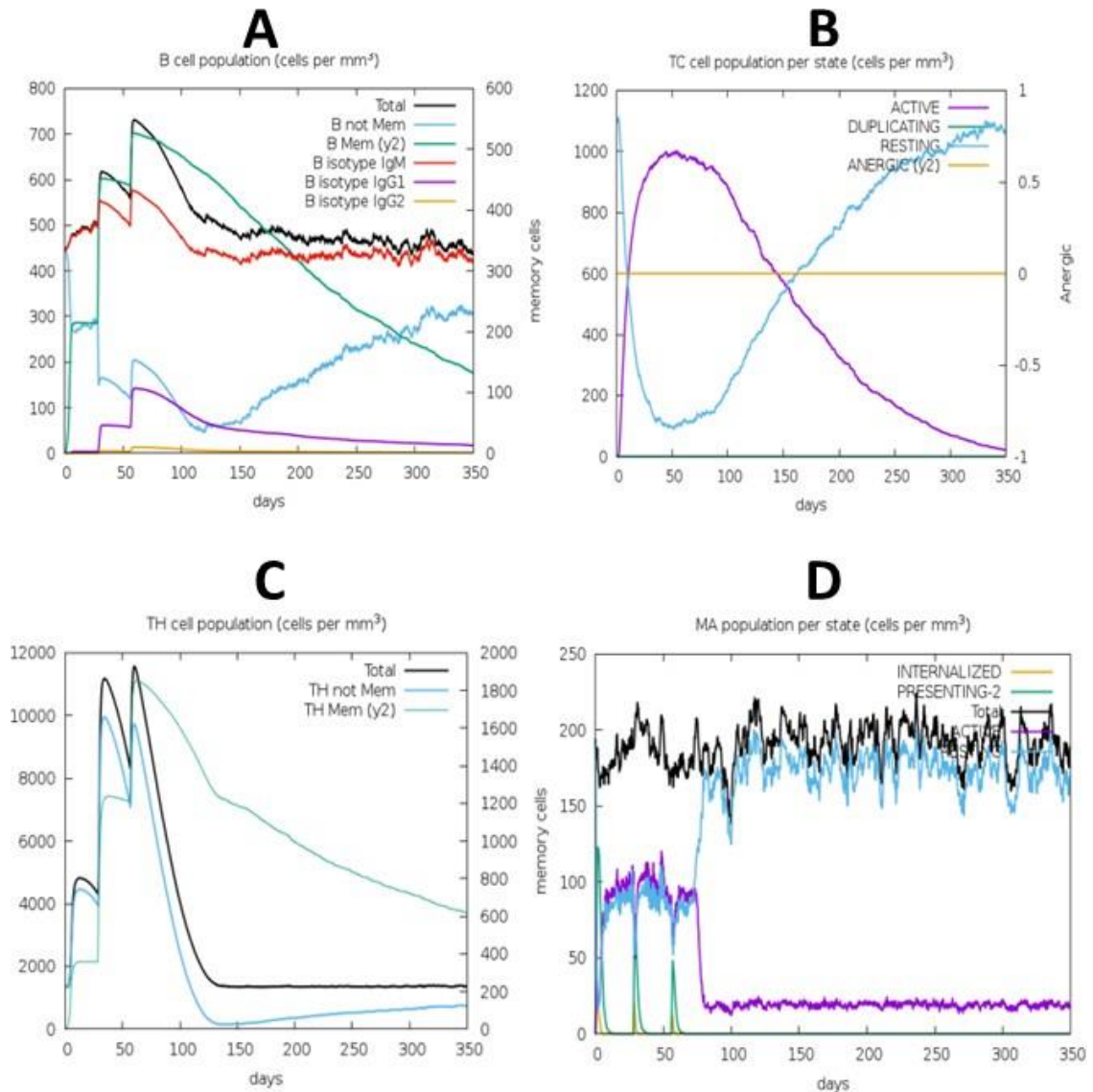


Figure 6. Immune simulation results show (Vaccine construct V1) the amount (cells permm³) and activation of the leading defense cells (vaccine 1 representation). **A:** B cell population; **B:** TC cell population per state; **C:** TH cell population; macrophages population per state.

3.10 *In silico* cloning in plasmid

Codon adaptation for expression in *E. coli* (strain K12) was performed through the JCat program. The CAI score of all vaccine constructs was 1.0. The GC content for Vaccine 1 was 57.66, for Vaccine 2 was 56.97, for Vaccine 3 was 57.84, and for Vaccine4 was 56.03. The optimized sequences' GC content and CAI scores were within the optimal range. Restriction sites for *SacI* and *BamHI* were added to the 5' and 3' ends of the inserts, respectively, and the sequence was cloned into the pET28a(+) vector.

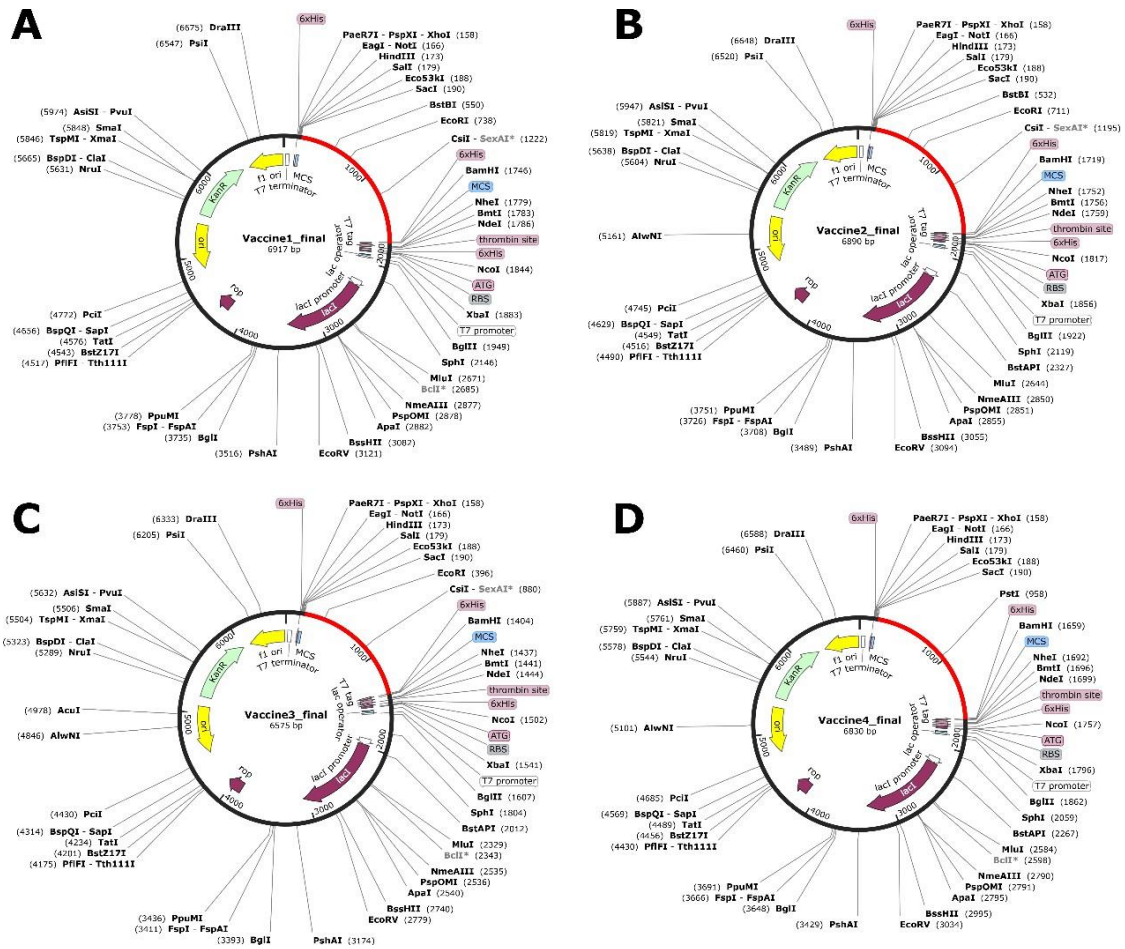


Figure 7. *In silico* cloning circular map. DNA sequences of vaccines (red) in pET28a(+)insertion in a vector of *E. coli* K12. A; B, C, D

4. DISCUSSION

The immunoinformatics approach produces a hetero-valent vaccine capable of targeting multiple pathogens/strains, reducing the risk of the pathogen escaping the immune system [30]. Zoonotic diseases have been rising in underdeveloped countries with low vaccination rates, posing risks to human and cattle populations [31]. *C. silvaticum* has just been described, and a study has already shown that this bacterium can infect human epithelial cells, in addition to having been previously isolated from infectious lesions of pigs and roe deer. Genomic studies already showed the zoonotic potential of this species, which was inserted in the clade called *C. diphtheriae* complex [4,5]. Some strains of *C. ulcerans* are *tox* positive (producing DT), but many strains are pathogens causing other diseases common in the respiratory system, and it is a zoonotic pathogen of economic, veterinary, and public health importance [2]. These traits

indicate the necessity of developing preventative and diagnostic methods for *C. ulcerans* and *C. silvaticum*. A previous comparative genomics study performed with *C. ulcerans* (72 genomes) and *C. silvaticum* (36 genomes) strains [32] has revealed that the transmembrane and secreted proteins represent suitable vaccine candidates. These proteins are localized in the outer membrane, accessible to the host immune system, are known to be associated with virulence, and are crucial for the survival of bacteria. Here we applied an immunoinformatics strategy to predict the epitopes based on proteins common to all strains of both species. The designed multi-epitope vaccine contains 25 common epitopes from 8 proteins from both species with great predicted immunological properties. Appropriate linkers connected the selected epitopes. The used linkers were experimentally confirmed in antigen presentation assays and have been reported to function as cleavage sites for different sets of immune cells and comprise stereochemical properties required for protein stability [33].

The final four multi-epitope vaccines proved to be potent immunogenic properties, non-host homologous, non-toxic, and non-allergen with a high aliphatic index, which shows their stability in a wide range of temperatures. In the protein-protein docking and molecular dynamic simulation analyses, we observed that all four multi-epitope vaccines demonstrated stable intermolecular interactions with TLR 2. TLR2 and TLR4 are widely used in immunological studies of gram-positive and gram-negative bacteria, as they recognize many components of the bacterial cell wall. TLR2 has a major function in recognition of gram-positive bacteria, is known to sense mycolic acids (glycolipids) from the cell wall of corynebacteria, and is involved in the activation of macrophages (Takeuchi et al. 1999; Schick et al. 2017). In addition, the physicochemical features for these multi-epitope candidates suggest stable, soluble and heterologous expression.

The immune simulation suggests the vaccine constructs may have the potential for the induction of humoral and cellular immune responses. These immune responses are already known in *C. ulcerans* and its closely related species, such as *C. pseudotuberculosis* and *C. jeikeium* [36,37]. Corynebacteria induce innate and adaptive immune responses in humans and animals by recruiting and activating B and T cells and macrophages. An *in vivo* study in a murine model demonstrated that *C. ulcerans* causes macrophage immune responses. Both toxigenic and non-toxigenic strains of *C. ulcerans* cause infections in the respiratory system, kidneys, spleen, and peripheral blood. As a result of this spread, it causes systemic infection [38]. The *in silico* immune simulation using the C-ImmSim server showed increased immunoglobulins observed, indicating the production of memory B-cells.

These immunogens can be used as multi-epitope subunit vaccines to control different diseases caused by *C. ulcerans* and prevent *C. silvaticum* infections. They can be used in diagnostic analysis at any stage of disease or in possible asymptomatic cases of both bacteria [39]. Interestingly the level of T-helper lymphocytes and the activation of cytotoxic T lymphocytes significantly support that the vaccine candidates have great potential to elicit humoral and adaptive immune responses.

CONCLUSION

In this study, we built four multi-epitope vaccines based on eight common proteins of *C. ulcerans* and *C. silvaticum*, with immunogenic properties and the ability to induce innate and adaptive immune responses. *In silico* approaches such as genomics and immunoinformatics allow these experiments to be carried out without financial costs and in a short time. The limitation of computational tools is that they will never imitate nature, which is why experimental validations *in vitro* and *in vivo* are always necessary, but when reaching these final stages, the financial investment and time will be much lower.

CONFLICT OF INTEREST

None declared.

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SUPPLEMENTARY MATERIAL

Supplementary Figure 1. Three-dimensional structure modelling by Alphafold and Ramachandran plots for the initial models (Before refinement).

Supplementary Figure 2. Molecular dynamics simulation, Normal Mode Analysis. The figure represents Eigenvalue of vaccine-TLR2 complex (**A**: V1; **B**: V2; **C**: V3 and **D**: V4).

Supplementary Figure 3. Immune simulation results show (Vaccine construct V2) the amount (cells per mm³) and activation of the main defense cells (vaccine 1 representation). **A**: B cell population; **B**: TC cell population per state; **C**: TH cell population; **D**: macrophages population per state.

Supplementary Figure 4. Immune simulation results show (Vaccine construct V3) the amount (cells per mm³) and activation of the main defense cells (vaccine 1 representation). **A**: B cell population; **B**: TC cell population per state; **C**: TH cell population; **D**: macrophages population per state.

Supplementary Figure 5. Immune simulation results show (Vaccine construct V4) the amount (cells per mm³) and activation of the main defense cells (vaccine 1 representation). **A**: B cell population; **B**: TC cell population per state; **C**: TH cell population; **D**: macrophages population per state.

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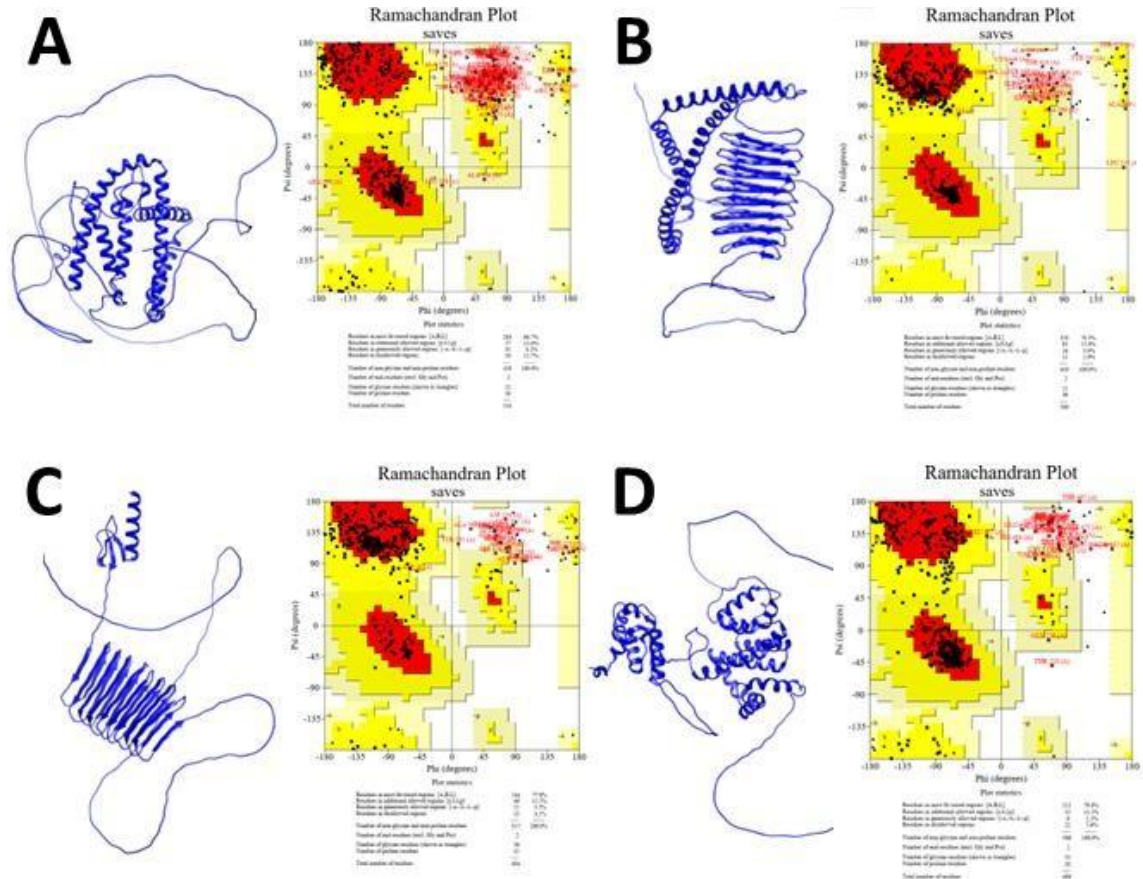
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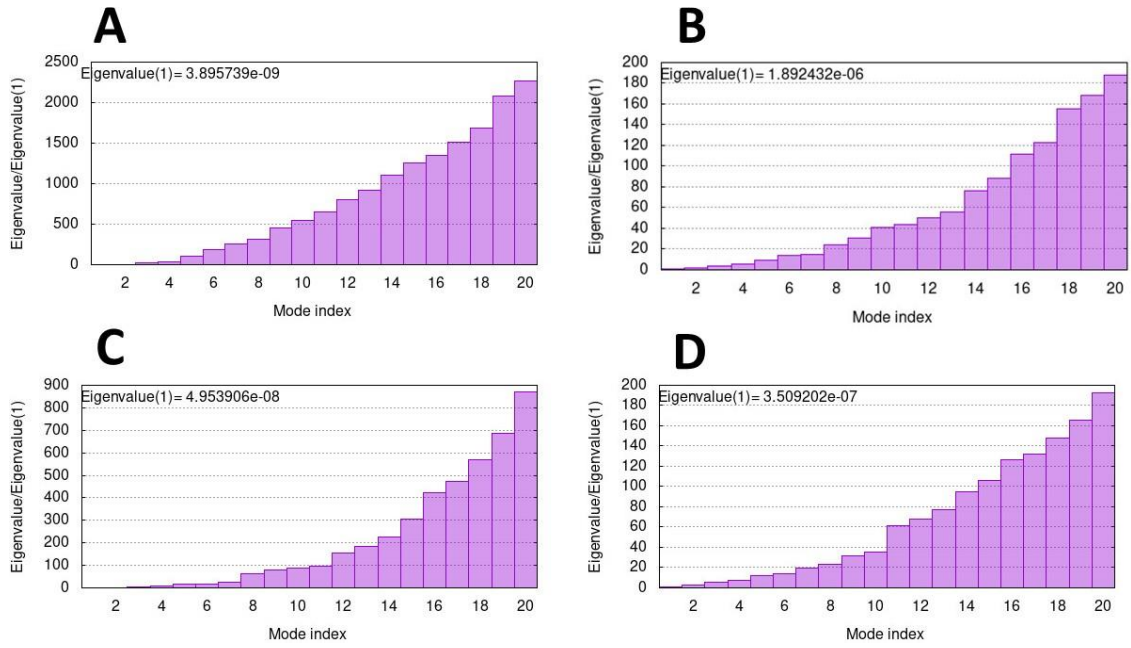
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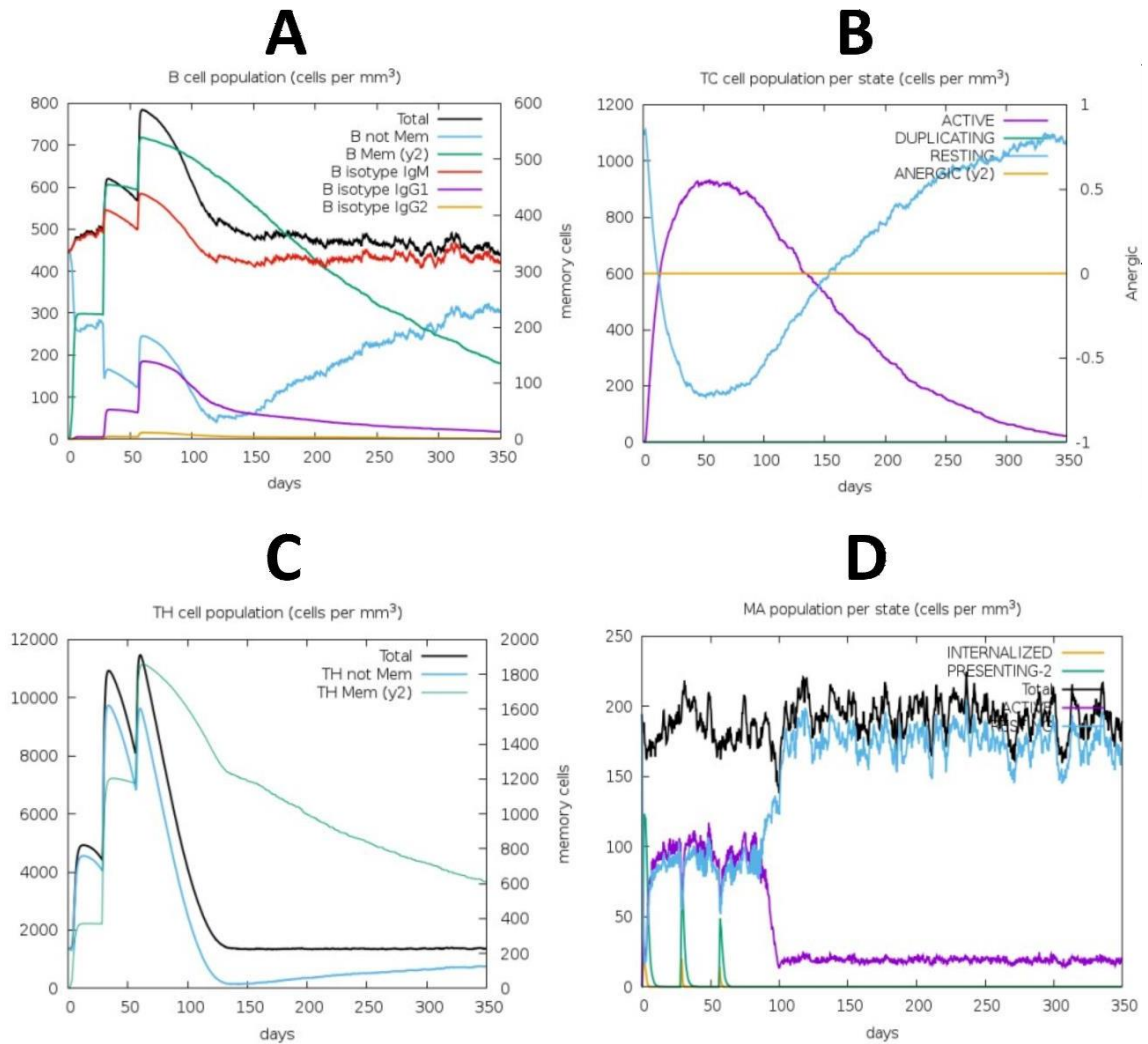
Supporting information



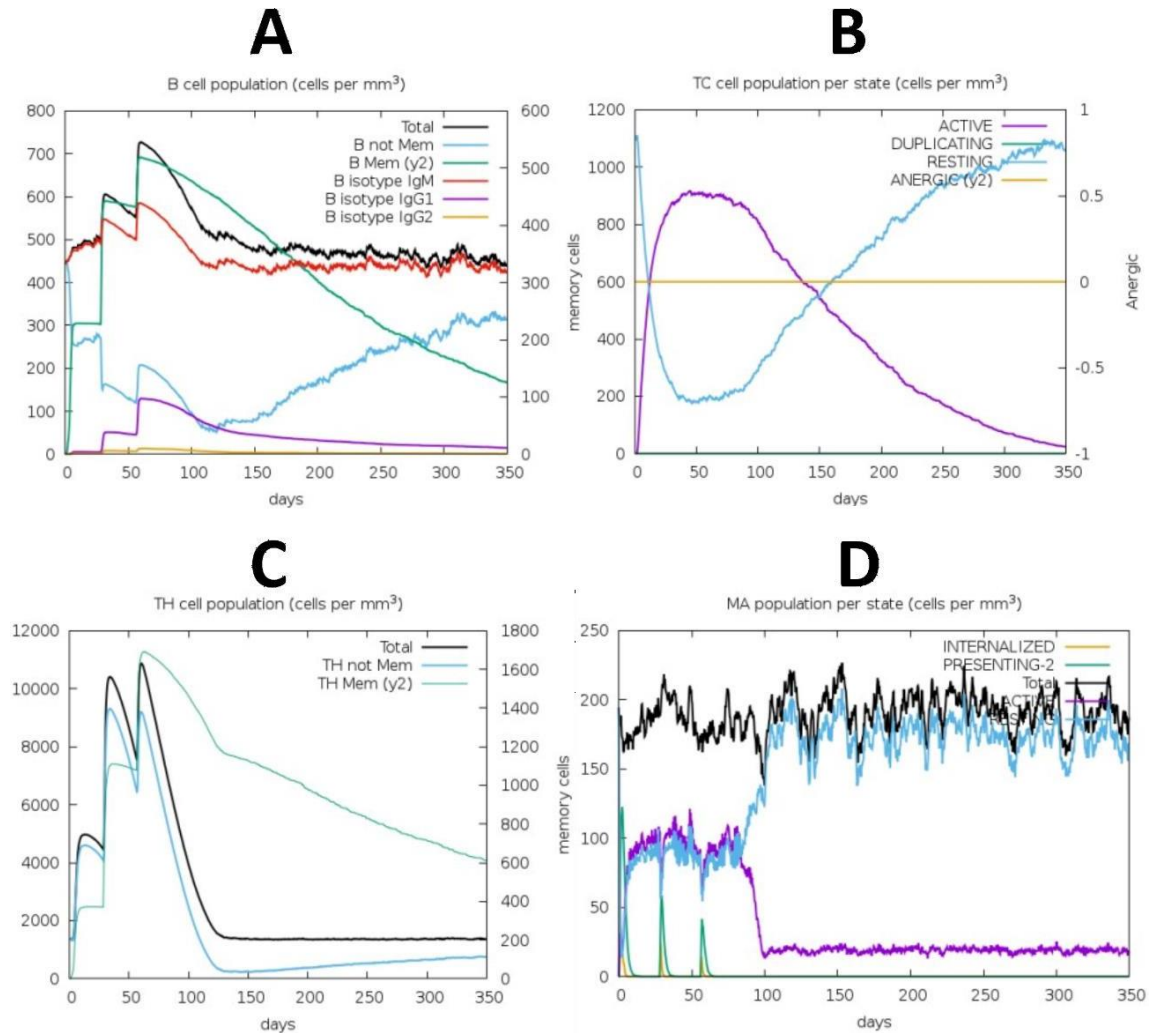
Supplementary Figure 1. Three-dimensional structure modelling by AlphaFold and Ramachandran plots for the initial models (Before refinement). **A:** V1 showing 66.7% residues in most favored regions and 11.7% in disallowed regions; **B:** V2 showing 78.3% residues in most favored regions and 2.9% in disallowed regions; **C:** V3 showing 77.0% residues in most favored regions and 4.1% in disallowed regions and **D:** V4 showing 79.0% residues in most favored regions and 5.6% in disallowed regions.



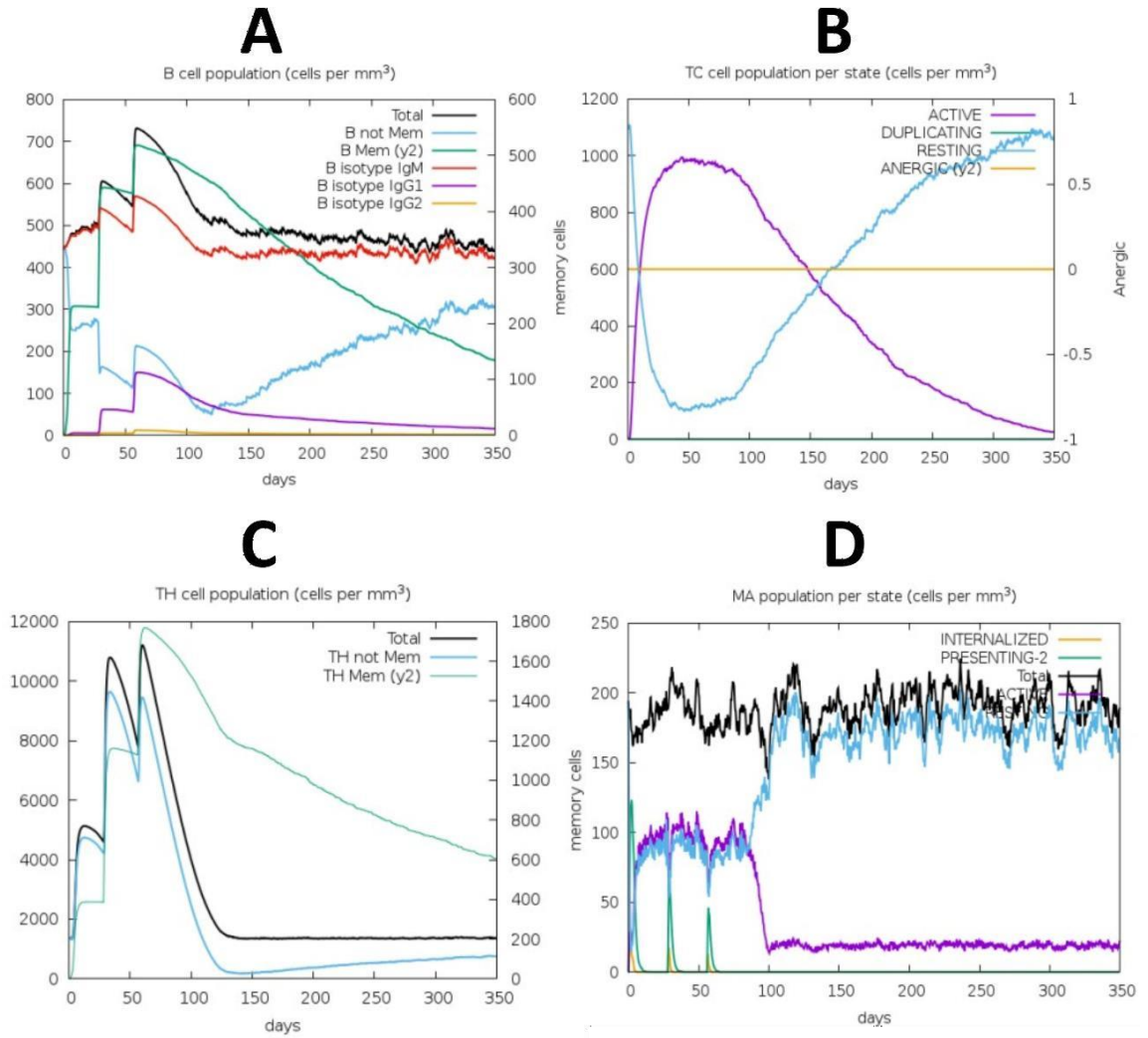
Supplementary Figure 2. Molecular dynamics simulation, Normal Mode Analysis. The figure represents Eigenvalue of vaccine-TLR2 complex (**A**: V1; **B**: V2; **C**: V3 and **D**: V4).



Supplementary Figure 3. Immune simulation results show (Vaccine construct V2) the amount (cells per mm^3) and activation of the main defense cells (vaccine 1 representation). **A:** B cell population; **B:** TC cell population per state; **C:** TH cellpopulation; **D:** macrophages population per state.



Supplementary Figure 4. Immune simulation results show (Vaccine construct V3) the amount (cells per mm^3) and activation of the main defense cells (vaccine 1 representation). **A:** B cell population; **B:** TC cell population per state; **C:** TH cellpopulation; **D:** macrophages population per state.



Supplementary Figure 5. Immune simulation results show (Vaccine construct V4) the amount (cells per mm^3) and activation of the main defense cells (vaccine 1 representation). **A:** B cell population; **B:** TC cell population per state; **C:** TH cell population; **D:** macrophages population per state.

V. CONCLUSÕES GERAIS

1. A aplicação de métodos *in silico* da genômica como a genômica subtrativa, vacinologia reversa e imunoinformática permitiu a predição de 9 alvos de vacinas e 4 alvos de quimioterápicos para *C. ulcerans* e *C. silvaticum*, respectivamente. Esta informação associada à validação experimental contribuirá para a redução de custo e tempo necessário para o desenvolvimento destes métodos de controle.
2. Foram selecionados os respectivos compostos naturais com melhor capacidade de inibir a função das proteínas citoplasmáticas (alvos de drogas) e, assim, interditar a colonização e crescimento desses organismos.
3. Esse trabalho é o primeiro a identificar alvos de drogas e vacinas, e propor imunógenos em comum para duas bactérias patogênicas do gênero *Corynebacterium*.
4. Através da imunoinformática, pôde-se propor 4 vacinas multi-epítipo para a prevenção de doenças causadas por *C. ulcerans* e *C. silvaticum*. As quatro vacinas de subunidades, foram construídas por epítipos preditos de 8 proteínas em comum das duas bactérias, tais vacinas apresentaram características adequadas de imunogenicidade, indução de respostas imunes esperadas por corinebactérias. Além de poderem ser expressas em vetor plasmídico de *Escherichia coli*.

VI. PERSPECTIVAS

1. *C. ulcerans* e *C. silvaticum* podem ser cultivadas e, portanto, é necessário realizar validação experimental por meio de análises *in vitro* e *in vivo*, a fim de testar a capacidade de indução das respostas imunes além da estabilidade estrutural da vacina candidata.
2. Realizar o monitoramento *in silico* da evolução dos genomas através da pangenômica de *C. ulcerans* e *C. silvaticum*, em consequência de futuros depósitos de sequências genômicas.
3. Proceder análises futuras de novos alvos de vacinas, à medida em que os bancos de dados genômicos e de compostos quimioterápicos vão aumentando e se atualizando.
4. Fazer análises de suscetibilidade e resistência antimicrobiana de ambas as espécies.

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VIII. ANEXOS

Nesta seção estão apresentadas as colaborações em artigos publicados em periódicos revisados por pares, além de certificados de participação em cursos e eventos, organização de eventos e apresentação de seminários, bem como certificados de cursos de formação complementar.

A: COLABORAÇÃO EM ARTIGOS PUBLICADOS EM PERIÓDICOS



Evidence of episodic positive selection in *Corynebacterium diphtheriae* complex of species and its implementations in identification of drug and vaccine targets

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ABSTRACT

Background: Within the pathogenic bacterial species *Corynebacterium* genus, six species that can produce diphtheria toxin (*C. belfantii*, *C. diphtheriae*, *C. pseudotuberculosis*, *C. rouxii*, *C. silvaticum* and *C. ulcerans*) form a clade referred to as the *C. diphtheria* complex. These species have been found in humans and other animals, causing diphtheria or other diseases. Here we show the results of a genome scale analysis to identify positive selection in protein-coding genes that may have resulted in the adaptations of these species to their ecological niches and suggest drug and vaccine targets.

Methods: Forty genomes were sampled to represent species, subspecies or biovars of *Corynebacterium*. Ten phylogenetic groups were tested for positive selection using the PosiGene pipeline, including species and biovars from the *C. diphtheria* complex. The detected genes were tested for recombination and had their sequences alignments and homology manually examined. The final genes were investigated for their function and a probable role as vaccine or drug targets.

Results: Nineteen genes were detected in the species *C. diphtheriae* (two), *C. pseudotuberculosis* (10), *C. rouxii* (one), and *C. ulcerans* (six). Those were found to be involved in defense, translation, energy production, and transport and in the metabolism of carbohydrates, amino acids, nucleotides, and coenzymes. Fourteen were identified as essential genes, and six as virulence factors. Thirteen from the 19 genes were identified as potential drug targets and four as potential vaccine candidates. These genes could be important in the prevention and treatment of the diseases caused by these bacteria.

Subjects Bioinformatics, Genomics, Microbiology, Molecular Biology

Keywords *Corynebacterium*, Positive selection, Drug target, Vaccine target

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Characterization of the first vaginal *Lactobacillus crispatus* genomes isolated in Brazil

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ABSTRACT

Background: *Lactobacillus crispatus* is the dominant species in the vaginal microbiota associated with health and considered a homeostasis biomarker. Interestingly, some strains are even used as probiotics. However, the genetic mechanisms of *L. crispatus* involved in the control of the vaginal microbiome and protection against bacterial vaginosis (BV) are not entirely known. To further investigate these mechanisms, we sequenced and characterized the first four *L. crispatus* genomes from vaginal samples from Brazilian women and used genome-wide association study (GWAS) and comparative analyses to identify genetic mechanisms involved in healthy or BV conditions and selective pressures acting in the vaginal microbiome.

Methods: The four genomes were sequenced, assembled using ten different strategies and automatically annotated. The functional characterization was performed by bioinformatics tools comparing with known probiotic strains. Moreover, it was selected one representative strain (*L. crispatus* CRI4) for in vitro detection of phages by electron microscopy. Evolutionary analysis, including phylogeny, GWAS and positive selection were performed using 46 public genomes strains representing health and BV conditions.

Results: Genes involved in probiotic effects such as lactic acid production, hydrogen peroxide, bacteriocins, and adhesin were identified. Three hemolysins and putrescine production were predicted, although these features are also present in other probiotic strains. The four genomes presented no plasmids, but 14 known families insertion sequences and several prophages were detected. However, none of the mobile genetic elements contained antimicrobial resistance genes. The genomes harbor a CRISPR-Cas subtype II-A system that is probably inactivated due to fragmentation of the genes *csn2* and *cas9*. No genomic feature was associated with a

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Probiotic potential of novel Brazilian *Lactobacillus crispatus* strains

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ABSTRACT. *Lactobacilli* are predominant bacterial species colonizing the vaginal surfaces of healthy women, where they play a protective role against opportunistic and polymicrobial infections such as Bacterial Vaginosis (BV). Several *Lactobacillus* species, especially *L. crispatus*, have been prospected for probiotic applications due to their potential antimicrobial and anti-inflammatory capacity. In the last decade, several genomic studies have investigated the genetic factors of *L. crispatus* strains for identifying novel probiotic strains and evaluating their application for improving human or animal health. This mini-review highlights the primary genes associated with *L. crispatus* protective mechanisms identified, in our previous work, on four novel strains isolated from healthy Brazilian women of reproductive age. Among the probiotic features discussed, the role of a pyruvate oxidase-encoding gene, lactate synthesis-related enzymes, bacteriocins genes, and genomic islands is reviewed, and the following steps to confirm their activity are pointed out.



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MICROBIOLOGY

Novel insights in bacterial vaginosis etiology through genomic approaches

MARCELLE O. ALMEIDA, MARCUS VINICIUS C. VIANA, JANAÍNA C. CERQUEIRA,
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RODRIGO D.O. CARVALHO

Abstract: Bacterial vaginosis (BV) has been considered as dysbiosis state whose etiology is not fully understood. This condition affects a large number of women of reproductive age and its study has been highly relevant due to the growing association of BV with and gynecological and obstetric complications and diseases, in addition to a greater susceptibility to sexually transmitted diseases, including HIV. The vaginal microbiota composition presents high variability among different ethnic groups of women, although, generally, the prevalence of lactobacilli species has been reported. Several studies suggest they may play a protective role, especially *Lactobacillus crispatus* whose population is typically present in low proportions in women with BV. This review article describes the contributions and limitations of genomic approaches in elucidating protective characteristics and mechanisms associated with colonization and persistence of lactobacilli strains. Although some genetic features were associated with resilience of *L. crispatus* during BV, further studies are required to uncover their functions.

Key words: bacterial vaginosis, genomics, health, *Lactobacillus*.

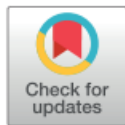
RESEARCH ARTICLE

Taxonomic classification of strain PO100/5 shows a broader geographic distribution and genetic markers of the recently described *Corynebacterium silvaticum*

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Abstract

The bacterial strain PO100/5 was isolated from a skin abscess taken from a pig (*Sus scrofa domestica*) in the Alentejo region of southern Portugal. It was identified as *Corynebacterium pseudotuberculosis* using biochemical tests, multiplex PCR and Pulsed Field Gel Electrophoresis. After genome sequencing and *rpoB* phylogeny, the strain was classified as *C. ulcerans*. To better understand the taxonomy of this strain and improve identification methods, we compared strain PO100/5 to other publicly available genomes from *C. diphtheriae* group. Taxonomic analysis reclassified it and three other strains as the recently described *C. silvaticum*, which have been isolated from wild boar and roe deer in Germany and Austria. The results showed that PO100/5 is the first sequenced genome of a *C. silvaticum* strain from livestock and a different geographical region, has the unique sequence type ST709, and could produce the *diphtheriae* toxin, along with strain 05–13. Genomic analysis of PO100/5 showed four prophages, and eight conserved genomic islands in comparison to *C. ulcerans*. Pangenome analysis of 38 *C. silvaticum* and 76 *C. ulcerans* genomes suggested that *C. silvaticum* is a genetically homogeneous species, with 73.6% of its genes conserved and a pangenome near to be closed ($\alpha > 0.952$). There are 172 genes that are unique to *C. silvaticum* in comparison to *C. ulcerans*. Most of these conserved genes are related to nutrient uptake and metabolism, prophages or immunity against them, and could be genetic markers for species identification. Strains PO100/5 (livestock) and KL0182^T (wild

B: PARTICIPAÇÃO EM EVENTOS, SEMINÁRIO E CURSOS DE CURTA DURAÇÃO





CERTIFICADO

Certificamos que

JANAÍNA CANÁRIO CERQUEIRA

participou, na qualidade de OUVINTE, do **Host-Pathogen Interaction Meeting 2021**, realizado no período de 08 a 12 de novembro de 2021, na modalidade remota, contabilizando carga horária total de 20 horas.

Curitiba, 28 de novembro de 2021

Dr. Alexander Welker Biondo
Coordenador do Programa de Pós-graduação
em Biologia Celular e Molecular da UFPR

Dr. Marcel Ivan Ramirez
Coordenador Geral do Host-Pathogen
Interaction Meeting 2021

Organização



Apoio



1st Latin American Environmental DNA Metagenomics Symposium

CERTIFICATE OF ATTENDANCE

This certificate is presented to

Janaína Canário Cerqueira

for attending the
1st Latin American Environmental DNA Metagenomics Symposium
held on the days 20 to 22 October, 2021, with the duration of 13 hours.

Prof. Daniel Cardoso de Carvalho
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