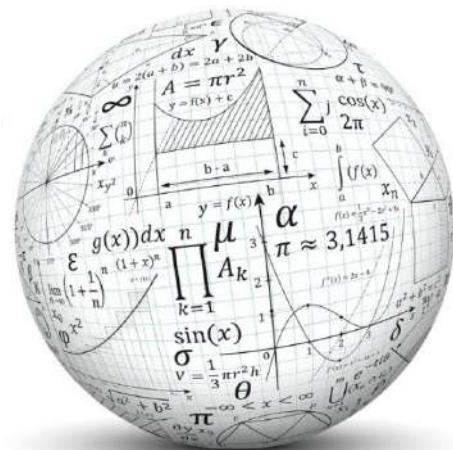


In silico prediction of Tumor Associated Antigens' immunogenicity

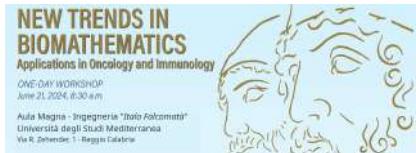
Filippo Castiglione



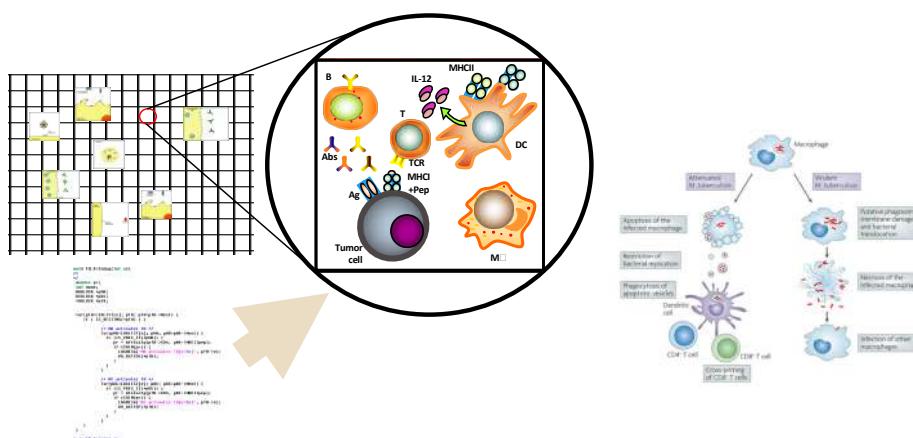
National Research Council of Italy, Italy



1

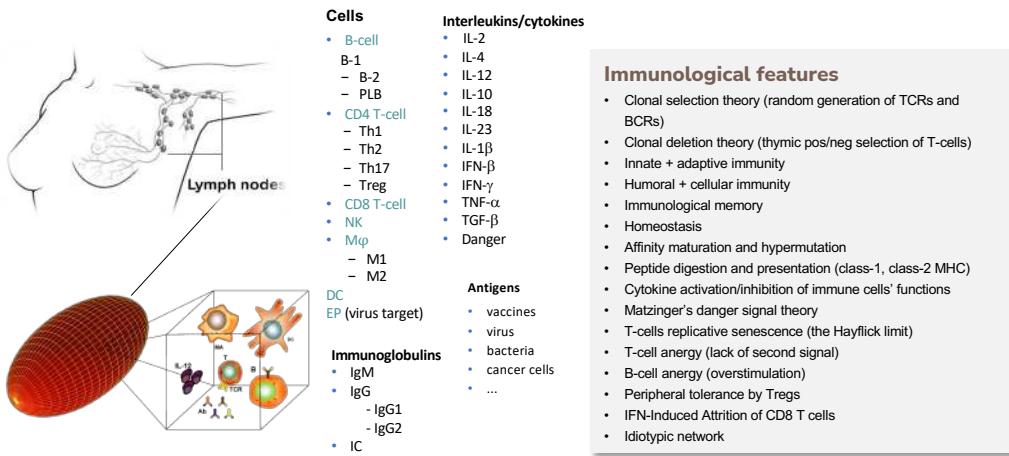


modeling biological processes with ABMs



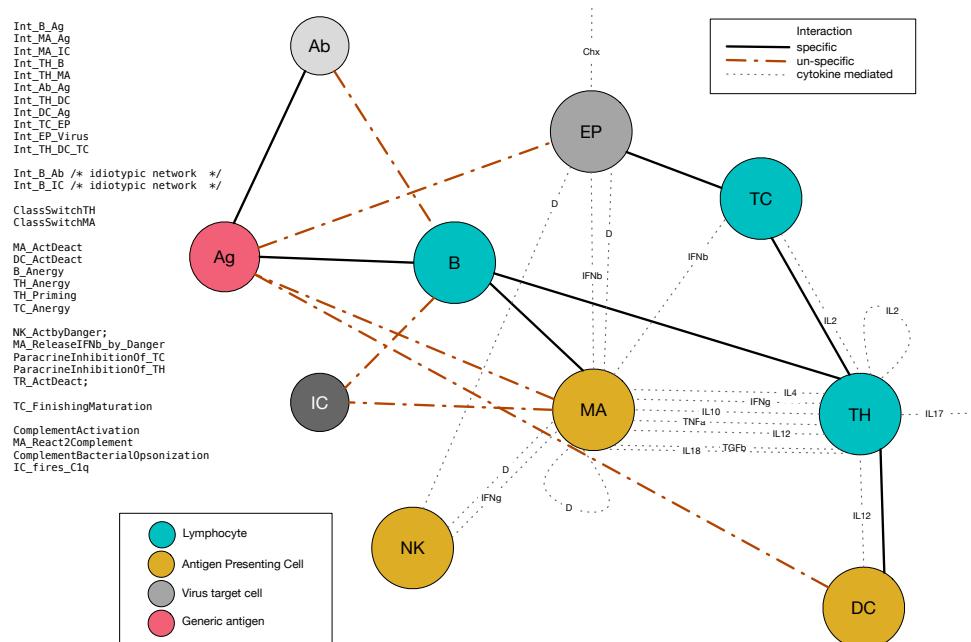
Agent-based Models (ABM) are mainly mechanistic models but may contain subprocesses described phenomenologically

stochastic ABM of the immune response

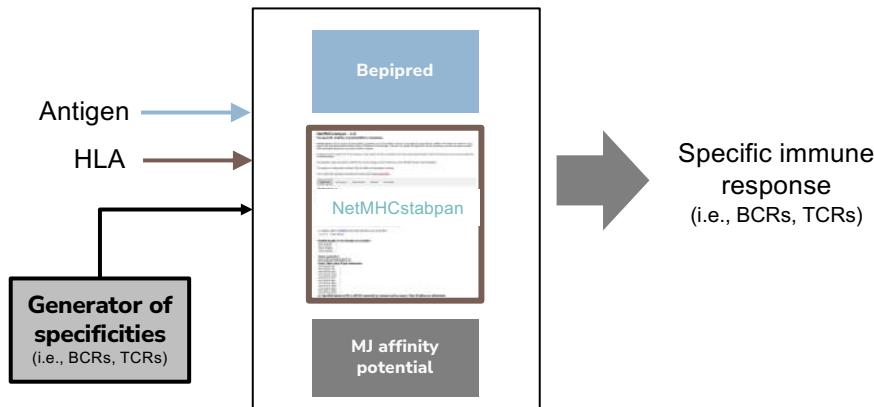


Celada F, Seiden PE. A computer model of cellular interactions in the immune system. Immunology Today 1992; 13:56-62

Castiglione F, Celada F. Immune System Modeling and Simulation. CRC Press, Boca Raton, London, New York, 2015



exploiting ML predictions



5

Choice of the HLAs

HLA database
www.allelefrequencies.net

class I

HLA-A02:01	44%
HLA-A03:01	17.4%
HLA-B35:01	20.9%
HLA-B51:01	28.7%

class II

DRB1*03:01	33.1%
DRB1*16:01	33.1%

% of individuals carrying the allele

The screenshot shows a search interface for HLA-A alleles in the "United Arab Emirates pop. 2 (n=573)". The results table includes columns for Line, Allele, Population, % of Individuals that have the allele, Allele Frequency (in decimal), Sample Size, IMGT/HLA Database, Distribution, Haplotype Association, and Notes. The results show the following top alleles:

Line	Allele	Population	% of Individuals that have the allele	Allele Frequency (in decimal)	Sample Size	IMGT/HLA Database	Distribution	Haplotype Association	Notes
1	A*02	United Arab Emirates pop. 2	44.0	0.2350	373	See	IP		
2	A*11	United Arab Emirates pop. 2	16.2	0.0860	373	See	IP		
3	A*03	United Arab Emirates pop. 2	17.4	0.0910	373	See	IP		
4	A*08	United Arab Emirates pop. 2	16.1	0.0840	373	See	IP		
5	A*26	United Arab Emirates pop. 2	14.2	0.0740	373	See	IP		
6	A*01	United Arab Emirates pop. 2	12.1	0.0600	373	See	IP		
7	A*23	United Arab Emirates pop. 2	11.8	0.0610	373	See	IP		
8	A*24	United Arab Emirates pop. 2	10.2	0.0500	373	See	IP		
9	A*30	United Arab Emirates pop. 2	9.7	0.0460	373	See	IP		
10	A*32	United Arab Emirates pop. 2	7.5	0.0390	373	See	IP		
11	A*25	United Arab Emirates pop. 2	6.2	0.0310	373	See	IP		
12	A*31	United Arab Emirates pop. 2	5.6	0.0280	373	See	IP		
13	A*29	United Arab Emirates pop. 2	3.2	0.0110	373	See	IP		

Prediction of potential linear B-cell epitopes

The screenshot shows the BepiPred-2.0 interface. An antigen sequence (NP_001639.1 prostate-specific antigen isoform 1 preproprotein) is submitted for analysis. The predicted epitopes and their scores are listed below.

Antigen → epitopes list and scores

Sequence: >NP_001639.1 prostate-specific antigen isoform 1 preproprotein [Homo sapiens]
 MWVPVVFTLSVTWIG**AAPLILSRVGGWECEKHSQPWQLVASRGRAVCGGVLVHPQWLTAHCIRNK**
SVILLGRHSLFHPEDTGQVFQVSHSFPHPLYDMSLLKNRFLRPGDDSSHDLMLLRLSEPAELTDAVKVMD
LPTQEPAALGTTCYASGWGSIEPEEFLTPKKLQCVDLHVISNDVCAQVHPQKVTKFMLCAGRW**GGKSTCS**
GDSGGPLVCNGVLQGITSWGESPCALPERPSLYTKVHYRKWIKDTIVANP

Jespersen MC, Peters B, Nielsen M, Marcatili P. BepiPred-2.0: improving sequence-based B-cell epitope prediction using conformational epitopes. Nucleic Acids Res 2017 (Web Server issue). doi: 10.1093/nar/gkx352

TCR-peptides affinity prediction

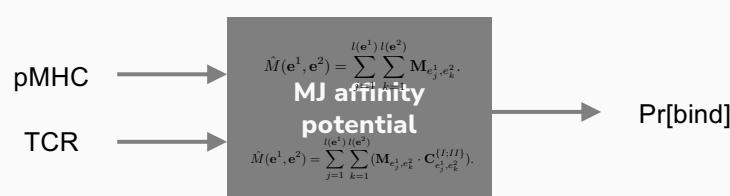


M. Rasmussen, E. Fenoy, M. Harndahl, A. Bregnbae Kristensen, I. Kallehave Nielsen, M. Nielsen, S. Buus. Pan-specific prediction of peptide-MHC-I complex stability; a correlate of T cell immunogenicity. *J Immunol* August 15, 2016, 197 (4) 1517-1524; doi:10.4049/jimmunol.1600582

Reynisson B, Barra C, Kaabinejad S, Hildebrand WH, Peters B, Nielsen M. Improved prediction of MHC II antigen presentation through integration and motif deconvolution of mass spectrometry MHC eluted ligand data. *J Proteome Res* 2020 Apr 30. doi: 10.1021/acs.jproteome.9b00874.

pMHC-TCR affinity calculation (and BCR-epitope)

Miyazawa-Jernigan TCR-MHCpep contact potential [4]



The work of Miyazawa and Jernigan on protein energy potentials provides a method for assessing the chance of direct interactions among proteins. The protein-protein potential concept was derived from the analysis of 3-dimensional structures in which the relative positions of amino acids were determined.

The contact potential matrix estimated by Miyazawa and Jernigan reflects the entropy between two residues. Low entropy means that the two residues have low energy and, therefore, that interaction between them is possible.

Miyazawa S, Jernigan RL (2000) Identifying sequence-structure pairs undetected by sequence alignments. *Protein Eng* 13: 459–475

Application 1

Computational modeling of prostate cancer (active)
immunotherapy

KLK3

- The introduction of testing for prostate-specific antigen (PSA), a member of the fifteen-gene family of kallikrein-related peptidases and also known as kallikrein-related peptidase 3 (KLK3), in blood has revolutionized both the detection and management of prostate cancer.
- Initially identified in 1966, PSA (KLK3), a 33-kDa glycoprotein secreted by prostatic epithelial cells, was first characterized in 1971 by Hara et al. in forensic studies as a marker for human semen^[*]

^[*] M. Hara, Y. Koyanagi, T. Inoue, and T. Fukuyama, "Some physico-chemical characteristics of "gamma-seminoprotein", an antigenic component specific for human seminal plasma. Forensic immunological study of body fluids and secretion. VII," *Japanese Journal of Legal Medicine*, vol. 25, no. 4, pp. 322– 324, 1971 (Japanese).

Prediction of potential linear B-cell epitopes

The screenshot shows the DTU Health Tech BiPePredict interface. At the top, there's a navigation bar with links for Home, Departments and Centers, Services, Contact, and Search. Below the navigation is the DTU logo. The main content area has tabs for News, Education, Research, Collaboration, Services and Products (which is selected), and About Us. A sub-navigation menu under Services and Products includes Home, Diseases and Pathogens, and Tools. Below this is a search bar with placeholder text 'Search' and a magnifying glass icon. The main content area is titled 'BiPePred - 2.0 Prediction of potential linear B-cell epitopes'. It contains a text input field for pasting protein sequences, a note about file formats, and a progress bar indicating the upload of 1000 amino acids. At the bottom, there's a large red arrow pointing from the 'Submit data' button to the right, where another red arrow points to the text 'epitopes list and scores'.

>NP_001639.1 prostate-specific antigen isoform 1 preproprotein [Homo sapiens]
MWPVVFVTLSTWIG**AAPLILSRI**VGGWECEKHSQPWQLVASRGRAVCGGVLVHPQWLTAHCIRNK
SVILLGRHSL**FHPEDT**GQVFQVSH**SFPHPLYDMSLLKNRFLRPGDDSSH**DLMLRLSEPAELTDAVKMD
LPTQEPAALGTTCYASGWGS**EPEEFLTPKKL**QCVDLHVIS**NDVCAQVHPQKVTKF**MLCAGRWT**GGKSTCS**
GDSGGPI VCNGVI QGTTSWGS**FPCAI PFR**PSI YTKV**VHYRKWTKDT**TVANP

Strong binder peptides were selected with a predicted affinity value <100nM and stability >1hr

Prediction of T-cell peptides

T-cell peptides in the study were predicted with the servers

- NetMHCpan-4.1
 - <https://services.healthtech.dtu.dk/service.php?NetMHCpan-4.1>
- NetMHCIIPan-3.2
 - <https://services.healthtech.dtu.dk/services/NetMHCIIPan-3.2>

T-cell peptides affinity prediction



M. Rasmussen, E. Fenoy, M. Harndahl, A. Bregnbae Kristensen, I. Kallehaug Nielsen, M. Nielsen, S. Buus. Pan-specific prediction of peptide-MHC-I complex stability: a correlate of T cell immunogenicity. *J Immunol* August 15, 2016, 197 (4) 1517-1524; doi:10.4049/jimmunol.1600582

Reynisson B, Barra C, Kaabinejad S, Hildebrand WH, Peters B, Nielsen M. Improved prediction of MHC II antigen presentation through integration and motif deconvolution of mass spectrometry MHC eluted ligand data. *J Proteome Res* 2020 Apr 30. doi: 10.1021/acs.jproteome.9b00874.

Most prevalent HLAs in the UAE



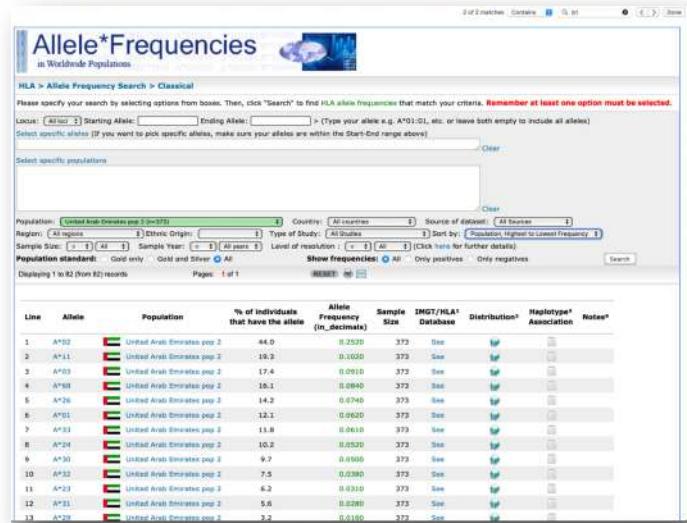
Source: HLA database
(www.allelefrequencies.net)

class I

HLA-A02:01	44%
HLA-A03:01	17.4%
HLA-B35:01	20.9%
HLA-B51:01	28.7%

class II

DRB1*03:01	33.1%
DRB1*16:01	33.1%



% of individuals that have the allele

HLA-I peptides (CTL peptides)

A0201

>NP_001639.1 prostate-specific antigen isoform 1 preproprotein [Homo sapiens]
MWVPVFLTLSVTWIGAAPLILSRIVGGWCEKHSQPWQVLVASRGRVLVHPQWQVNLTAHCIRNK
Number of strong binders 1
Number of weak binders 6
S VILLGRHSLFHPEDTGQVFQVSHSFPHPLYDMSL LKNRFLRPGDDSSHDLMLRLSEPAELTDAVKVMD
LPTQEPALGTTCYASGWGSIEPEEFLTPPKLKLOCVDLHVTSNDVCAQVHPQKVTKFMLCAGRWTGGKSTCS
GDSGGPLVCNGVLQGITSWGSEPCALPERPSLYTKVVHYRKWIKDITIVANP

A0301

>NP_001639.1 prostate-specific antigen isoform 1 preproprotein [Homo sapiens]
MWVPVFLTLSVTWIGAAPLILSRIVGGWCEKHSQPWQVLVAVCGGVLVLPQWVLTAAHCIRNK
Number of strong binders 3
Number of weak binders 10
S VILLGRHSLFHPEDTGQVFQVSHSFPHPLYDMSL LKNRFLRPGDDSSHDLMLRLSEPAELTDAVKVMD
LPTQEPALGTTCYASGWGSIEPEEFLTPKKLKLOCVDLHVTSNDVCAQVHPQKVTKFMLCAGRWTGGKSTCS
GDSGGPLVCNGVLQGITSWGSEPCALPERPSLYTKVVHYRKWIKDITIVANP

B3501

>NP_001639.1 prostate-specific antigen isoform 1 preproprotein [Homo sapiens]
MWVPVFLTLSVTWIGAAPLILSRIVGGWCEKHSQPWQVLVASHRAVCGGLVLPQWVLTAAHCIRNK
Number of strong binders 6
Number of weak binders 8
S VILLGRHSLFHPEDTGQVFQVSHSFPHPLYDMSL LKNRFLRPGDDSSHDLMLRLSEPAELTDAVKVMD
LPTQEPALGTTCYASGWGSIEPEEFLTPKKLKLOCVDLHVTSNDVCAQVHPQKVTKFMLCAGRWTGGKSTCS
GDSGGPLVCNGVLQGITSWGSEPCALPERPSLYTKVVHYRKWIKDITIVANP

B5101

>NP_001639.1 prostate-specific antigen isoform 1 preproprotein [Homo sapiens]
MWVPVFLTLSVTWIGAAPLILSRIVGGWCEKHSQPWQVLVASHRAVCGGLVLPQWVLTAAHCIRNK
Number of strong binders 1
Number of weak binders 13
S VILLGRHSLFHPEDTGQVFQVSHSFPHPLYDMSL LKNRFLRPGDDSSHDLMLRLSEPAELTDAVKVMD
LPTQEPALGTTCYASGWGSIEPEEFLTPKKLKLOCVDLHVTSNDVCAQVHPQKVTKFMLCAGRWTGGKSTCS
GDSGGPLVCNGVLQGITSWGSEPCALPERPSLYTKVVHYRKWIKDITIVANP

HLA-II peptides (HTL peptides)

DRB1_0301

Number of strong binders: 3
Number of weak binders: 21

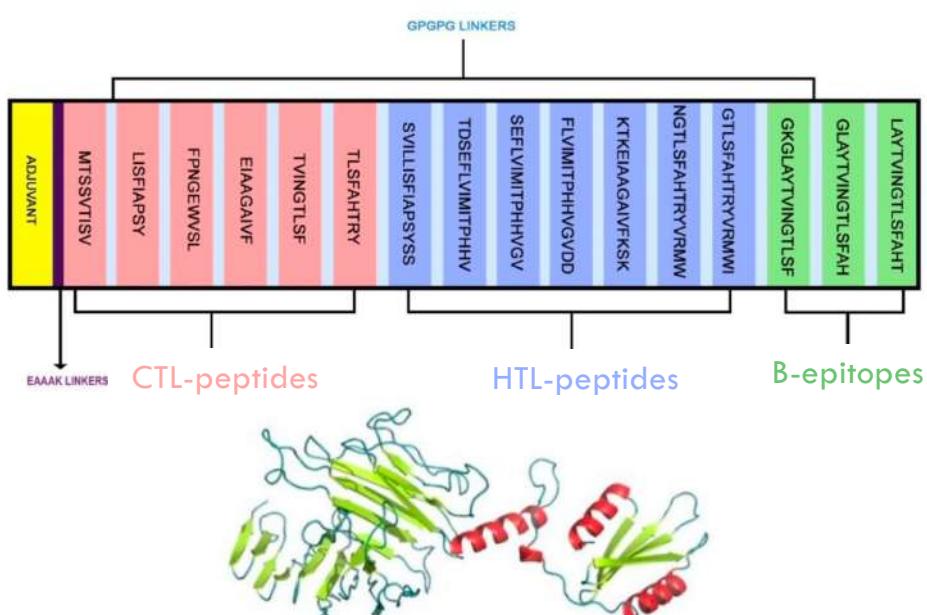
```
>NP_001639.1 prostate-specific antigen isoform 1 preproprotein [Homo sapiens]
MWVPVVFTLSVTWIGAAP[LILSRIVGWECEKHSOPWV]VASRGRAV[CGGVLVHPOWV]TAAH[CIR]NK
[S]V[I]LLGRHSLFHPEDTGQVFQVSHSFPHPLY[DMSLLKNRFLRP]GDDSSH[DLMLLRL]SEPAELTDAVKVMD
[LPTQ]EPALGTTCYASGWGSIEPEEFLTPKKLQCVDLHV[ISNDVCAQV]VHPQKVTKFMLCAGRWTGGKSTCS
GDGGPLVCNGVLQGITSWGSEPCALPERPS[LYTKVVHYRKWIKD]TIVANP
```

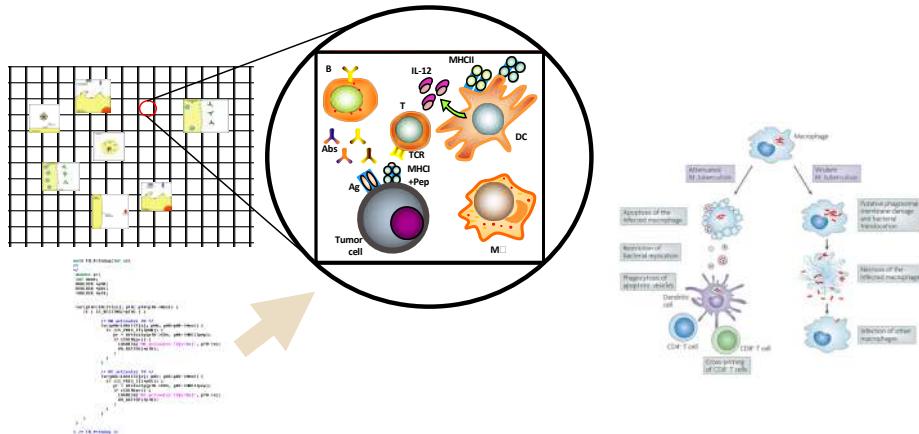
DRB1_1601

Number of strong binders: 8
Number of weak binders: 18

```
>NP_001639.1 prostate-specific antigen isoform 1 preproprotein [Homo sapiens]
MWVPVVFLTLSVTWIGAAP[TL]SRIVGGWECEKHSOP[WV]VASRGRAV[CGGVLVHPOWV]TAAH[CIR]NK
[S]V[I]LLGRHSLFHPEDTGQVFQVSHSFPHPLY[DMSLLKNRFLRP]GDDSSH[DLMLLRL]SEPAELTDAVKVMD
[LPTQ]EPALGTTCYASGWGSIEPEEFLTPKKLQCVDLHV[ISNDVCAQV]VHPQKVTKFMLCAGRWTGGKSTCS
GDGGPLVCNGVLQGITSWGSEPCALPERPS[LYTKVVHYRKWIKD]TIVANP
```

Multi-epitope subunit vaccine

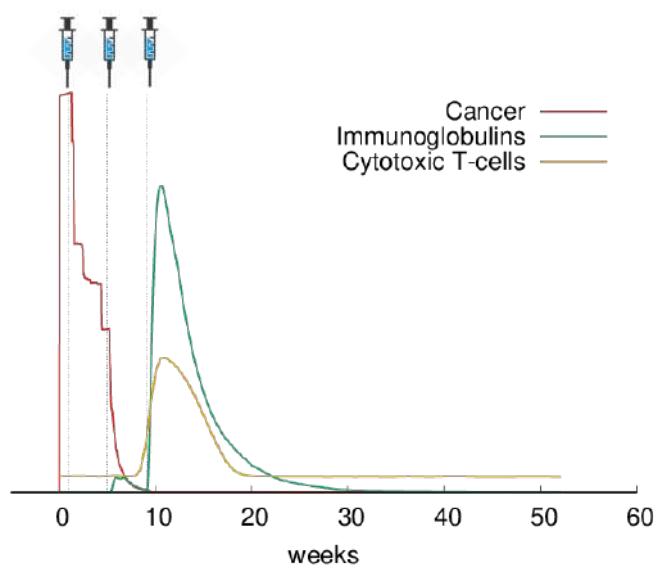
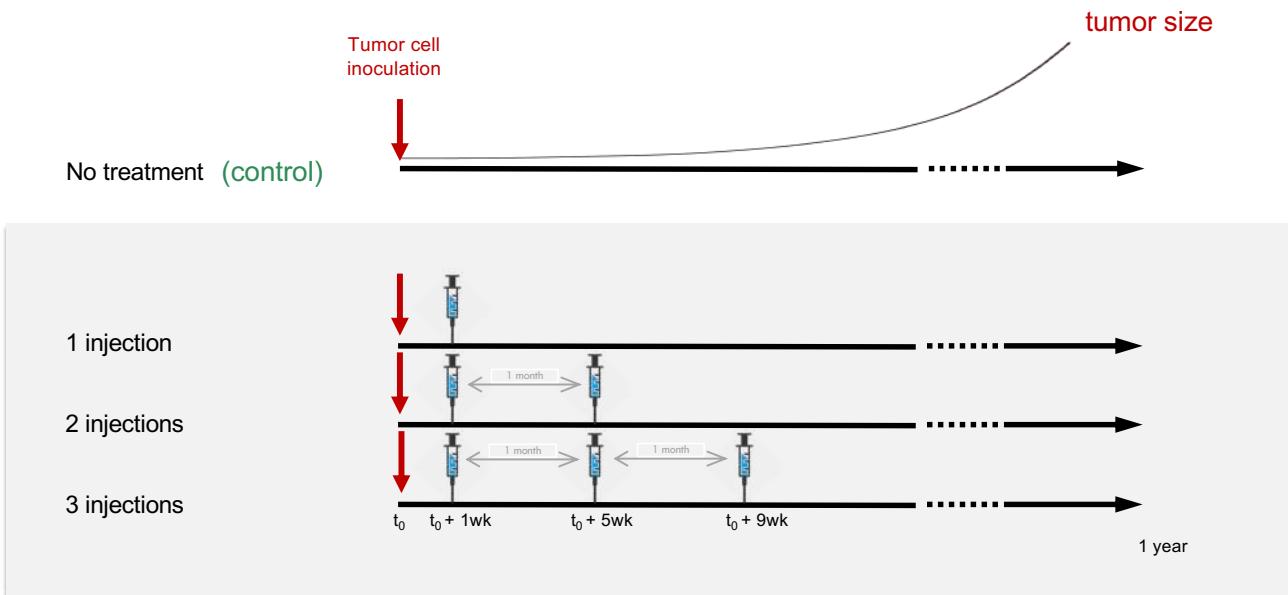


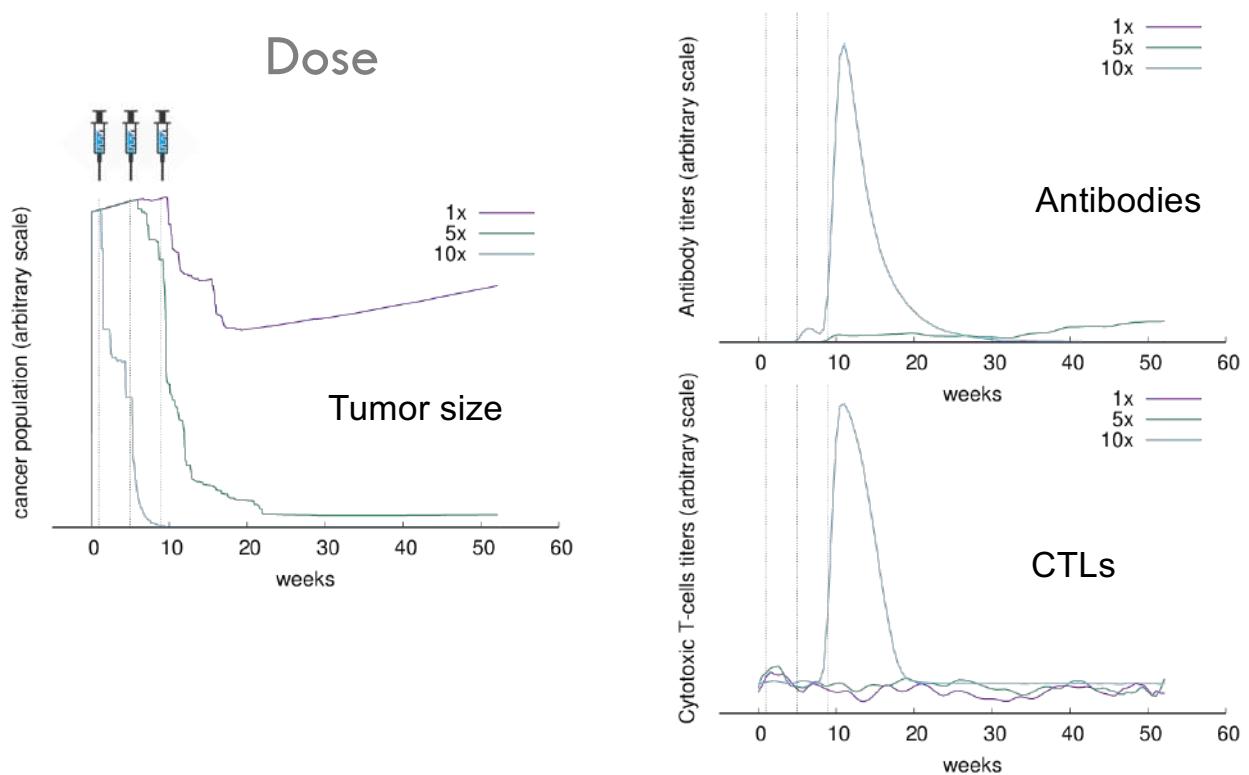
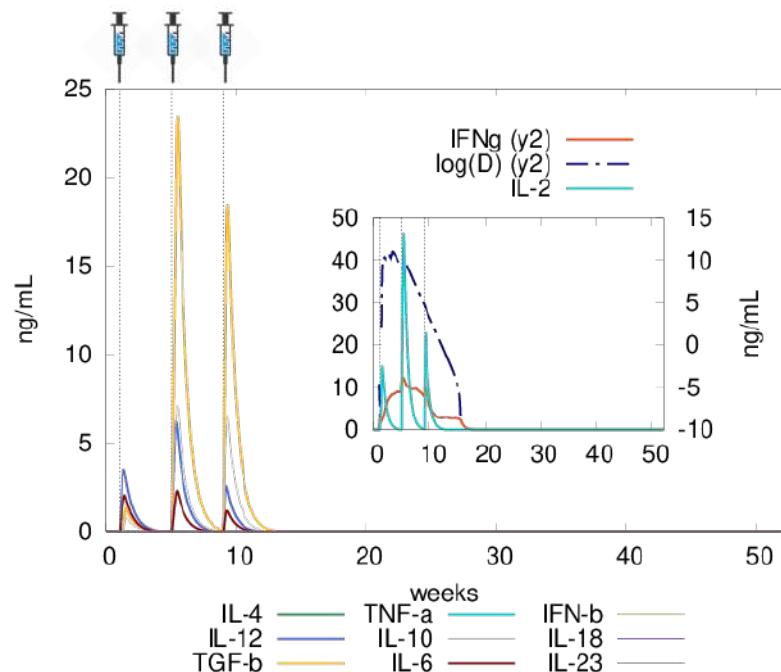


events / rules

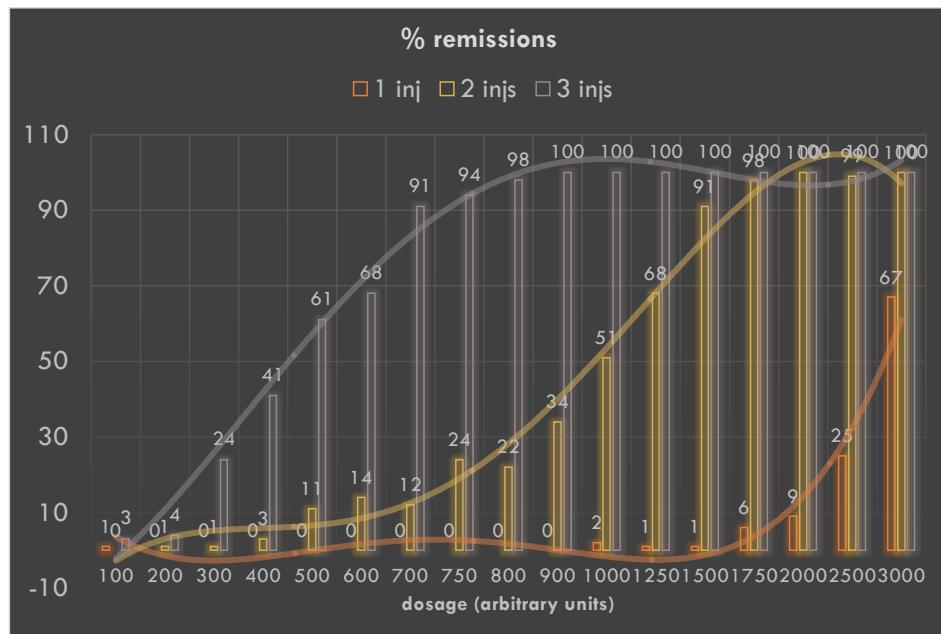
1. **Injection:** A dose $V(0) = V$ is injected into the simulated volume representing 10 microliters discretised in $L \times L \times L = 10 \times 5 \times 5 = 250$ lattice points.
2. **B phagocytosis:** B cells phagocytose, internalise, process and present viral peptides on class II HLA
3. **Response to Danger:**
 - NK response: Natural killer cells (NKs) release IFNg upon bystander stimulation by danger
 - M response: Macrophages (M) respond to danger (e.g., DAMPs) via TLR4 releasing TNFa and IL-6
4. **M activation:** macrophages become activated by IFNg (activated M have a greater phagocytic activity). This is modeled as a Bernoulli event with parameter $p = c \times e^{I/E}$, where $c = 0.9$, I is local concentration of IFNg (i.e., in lattice site x) and E is a parameter representing the efficiency of interferon in activating M
5. **Active M**
 - M phagocytosis: M internalise, process and present viral peptides on class II HLA; in presence of IFNg they release IL-12; they also release TNFa
 - DC activation: M release TNFa which activate dendritic cells (DC)
6. **DC phagocytosis & endocytosis:** DC phagocytose, internalise, process and present viral peptides on class II HLA (exocytic pathway) but also on class I HLA (endocytic pathway)
7. **Th activation:** in presence of danger signal, resting T helper lymphocytes are activated by interaction with peptide-bound HLAs on professional antigen presenting cells (M and DC, mainly DC) surface by means of specific interaction with their T-cell receptors (TCR); if no danger is present, the Th cells becomes anergic upon interaction of its TCR with the HLA-peptide complex
8. **Th stimulation by APCs:** activated Th interacting with antigen presenting cells (M, DC)
 - Th duplication: start clone expansion; 50% of the daughter cells become memory
 - Th cells release IL-2
 - M release IL-6
 - Th1 release IFNg
 - Th2 release IL-4
 - release IL-12 in presence of high local concentration of IFNg
 - Treg release TGFB and IL-10
9. **Th stimulation by B:** activated Th interacting with B cells
 - B duplication: stimulate B cells to clone expansion; 50% of the daughter cells become memory
 - Th duplication: start clone expansion; 50% of the daughter cells become memory
 - M release IL-6
 - Th1 release IFNg
 - Th2 release IL-4
 - Th release IL-12 in presence of high local concentration of IFNg
 - Treg release TGFB and IL-10
10. **Th stimulation by B:** activated Th interacting with B cells
 - B duplication: stimulate B cells to clone expansion; 50% of the daughter cells become memory
 - Th duplication: start clone expansion; 50% of the daughter cells become memory
 - Th release IL-2, IL-12
 - Th1 release IFNg
 - Th2 release IL-4
 - Treg release TGFB and IL-10
11. **Th differentiation:** depending on the local concentration of IFNg, IL-10, IL-4, IL-6, IFNb, IL-12, IL-18, IL-2, TGFB and IL23, active T helper cells undergo class switch into Th1 and Th2
12. **B differentiation:** B cells differentiate to antibody-secreting plasma B cells (PLB). 50% of duplicating B cells become PLBs. If the B lymphocyte is a memory cells then it generates 80% of PLBs
13. **Isotype switch:** B cells perform immunoglobulin class switching, that is, change production of immunoglobulin from the isotype IgM to the isotype IgG. This is modeled as a Bernoulli event with parameter p depending on the local concentration of IL-2
14. **Antibodies production:** Plasma cells secrete antibodies at a rate of about 2 ng/day
15. **Humoral response:** antibodies inhibit viral particles by opsonization; the result are the immuno-complexes that are eventually cleared by macrophages
16. **Tc activation:** in presence of IL-2, resting cytotoxic T cells (Tc) are activated by the interaction of their TCR with DC presenting on class I HLA the viral peptides but only in presence of IL-2
17. **Tc duplication:** activated Tc interact with cancer cells presenting viral peptides on class I HLA molecule
 - Tc start duplication. 50% of the daughter cells become memory cells
18. **Cytotoxic response:** activated Tc kill infected CC (this will further release danger signal)

Tested protocols





Dose escalation experiment



Application 2

identification of viral antigens sharing sequence and structural homology with tumor-associated antigens

in-silico prediction of TAA immunogenicity (active immunotherapy)

Identification and validation of viral antigens sharing sequence and structural homology with tumor-associated antigens (TAAs)

Methods We looked for homology between published TAAs and non-self-viral-derived epitopes.

Results Several homologies (structural similarities) have been found between paired TAAs and viral peptides. These show eliciting cross-reacting CD8⁺ T cell responses which possibly drive the fate of cancer development and progression.

Conclusions An established antiviral T cell memory may turn out to be an anticancer T cell memory, able to control the growth of a cancer developed during the lifetime if the expressed TAA is similar to the viral epitope.

C. Ragone, et al. Identification and validation of viral antigens sharing sequence and structural homology with tumor associated antigens (TAAs). *J ImmunoTherapy of Cancer*. 9:e002694 (2021)

The screenshot shows the article details on the BMJ Journals website. The title is 'Identification and validation of viral antigens sharing sequence and structural homology with tumor-associated antigens (TAAs)'. It includes authors' names (C. Ragone, G. Manzini, B. Gagliozzi, A. Mazzoni, M. Liss, Tornatore, F. Di Giuseppe, R. Capriglione, L. Viggiani, E. Iaccarino), a figure, and a table. The abstract discusses the immune system's equilibrium with viral antigens and how it can recognize tumor-associated antigens. The background section notes the presence of viral antigens in the body and their role in driving cancer development. The results section highlights structural similarities between TAAs and viral peptides. The conclusions section suggests that antiviral T cell memory can be harnessed for cancer therapy.

Cancer antigenic peptide database <https://caped.icp.ucl.ac.be/Peptide/list>

The screenshot shows a table from the Caped database. The columns include: Mutation, Tumor-specific, Differentiation, Overexpressed, Potential, GeneProtein ID, Name(Tumor), HLA ID, HLA Name, Peptide Sequence, Peptide ID, Peptide Length, Lymphocyte Stimulation, Reference ID, and Note. The table lists various mutations and their associated peptides across different HLA types. At the bottom, there are links for 'Print Table' and 'Did you find an article that is not present in the database? Please let us know.'

Mutation	Tumor-specific	Differentiation	Overexpressed	Potential	GeneProtein ID	Name(Tumor)	HLA ID	HLA Name	Peptide Sequence	Peptide ID	Peptide Length	Lymphocyte Stimulation	Reference ID	Note
alpha-actinin-4	lung carcinoma				A2	44	IAAGHGVLY	11B-127	autologous tumor cells	50Schultz_2001	11			
AP2S1	Colorectal cancer				A24	20	AYLDEAKRIF	70-87	peptide	50Kwak_2001	11			
ATOT1	melanoma				D85	18	YSDVWHLNGDITVTH		autologous tumor cells	50Wang_2000	11			
B-RAF	melanoma				D84	24	EDLVYNSQGLATDKIANSIGSHQEQQLS	698-614	peptide	50Shayegi_2004	11			
BCR-ABL fusion protein (BcrAb)	chronic myeloid leukemia				A8	14	GPKQGKVAL	922-933	peptide	50YNTs_1998	11			
					D84	24	ATGPGQGKVALQFQPKS	600-620	peptide	50Becht_1996	11			
					A2	48	SSKAKGRP	836-838	peptide	50YNTs_1998	11			
					D98	3	ATQPKQGKVALQFQPKS	600-630	peptide	50Mata_1992	11			
beta-catenin	melanoma				A24	39	SYLDSGGHIF	29-37	autologous tumor cells	50Rekhi_1998	11			
CD3BP-N	ovarian, gastric, and endometrial carcinoma				A2	45	FLWLRGHTM	67-74	peptide	50Schultz_2004	11			
CD3BP-N	head and neck squamous cell carcinoma				B95	30	FPRDEEDHYY	476-484	autologous tumor cells	50Mata_1992_1993	11			
CD47	melanoma				D98	38	PRWWAWDPHD	765-771	autologous tumor cells	50Wang_1999	11			
CSK12	melanoma				A11	13	CDLDLDTK	926-933	autologous tumor cells	50Rekhi_1998	11			
CSKA	melanoma				A2	44	ACDPSHQEV	39-53	autologous tumor cells	50Wang_1999	11			
CSK12A	melanoma				A11	13	AVDPMVTRN	105-112	autologous tumor cells	50Wang_2000	11			

The screenshot shows the main interface of the Caped database. It features a navigation bar with 'Database', 'Search', 'Help', and 'About'. Below the navigation is a cartoon illustration of a proteasome breaking down a protein into peptides, with a CTL cell and HLA molecules nearby. The text 'Welcome to the Cancer Antigenic Peptide Database' is displayed. There are buttons for 'Browse the database' and 'Search/filter the database'. A note at the bottom encourages users to report missing articles and share the site on social media.

Tumor Associated Antigen

- *CLPP*
- *Gp100*
- *HSPH1*
- *HEPACAM*
- *CD274*
- *MUC1*
- *KIF20A*
- *Tyrosinase*
- *CEA*
- *Telomerase*
- *Secernin1*

versus

Viral protein

- *E1 HPV*
- *UL20 HCMV*
- *large tegument protein HSV-2*
- *polyprotein encephalomyelitis virus*
- *ENV HIV*
- *ORF46 HHV8*
- *Env HIV*
- *Gag HIV and Env HERV*
- *Env HIV*
- *Env HIV*
- *PolB1 influenza*

Viral nanomer peptides for the four most prevalent MHC class I alleles where chosen by homology search (BLAST) and selected by NetMHCstapan v.1.0

31

Class I alleles (<http://www.allelefrequencies.net>)

- HLA-A*01:01
- HLA-A* 02:01
- HLA-A* 03:01
- HLA-A* 24:02

altogether cover about 50% of the world population

~ 60% of the European as well as the north American Caucasian populations

~ 50% of the Japanese population

~ 30% of the Chinese population

~ 20% of the Indian population

have been selected with a predicted affinity value <100 nM (strong binders, SBs: netMHCpan)

Polymorphic Region	Population Studies	Gene Allele	Haplotype	Genotype	Data
HLA	1319	1304	680	-	292
HLA	267	292	-	-	102
Cytokeratin	123	125	-	-	123
PEC	94	64	23	-	94
Total	1,801	1,795	683	102	1,795

The current number of frequencies stored in our database is: 130,400 (HLA), 6,731 (KIR), 4,276 (Cytokeratin) and 8,77 (PEC) from 24,223,879 individuals.

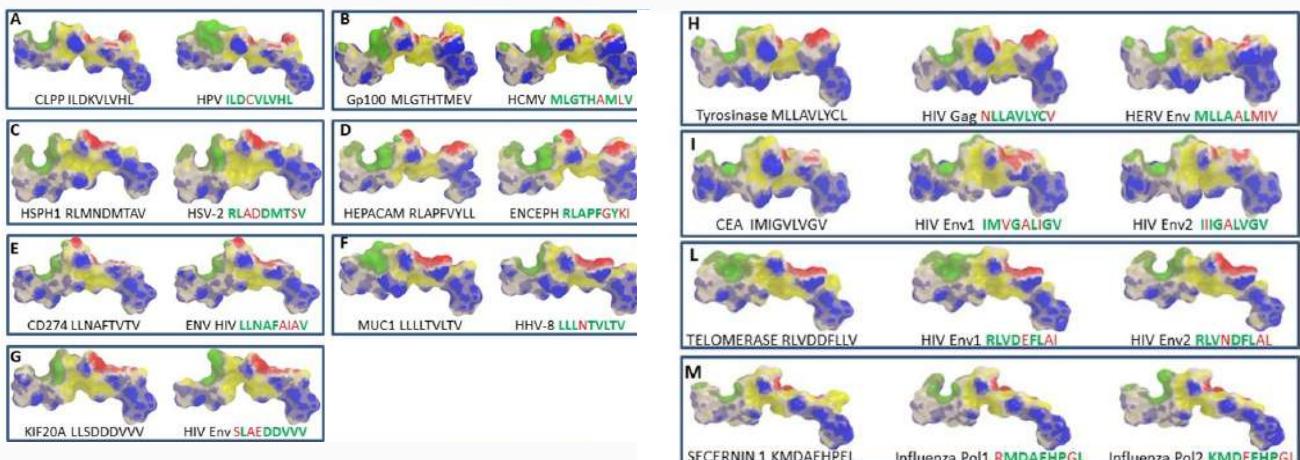
We have updated the website with the new IPD/HLA Nomenclature ([details](#)).
IPD-HLA last update: 2.0.0, 08 Jun 2018 | IPD-KIR last update: 2.0.0, 17 February 2018.

©2003-2014 The Allele Frequency Net Database

Homologies between TAAs and viral peptides

Structural predicted conformation of the paired viral and tumor-associated antigen peptides bound to the HLA-A*02:01.

Blue areas = contact points with HLA molecule;
Red areas=contact points with the TCR α chain;
Green areas=contact points with the TCR β chain.

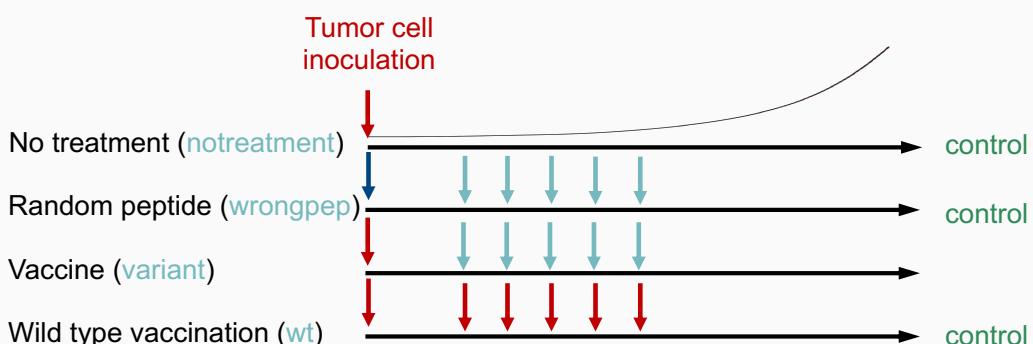


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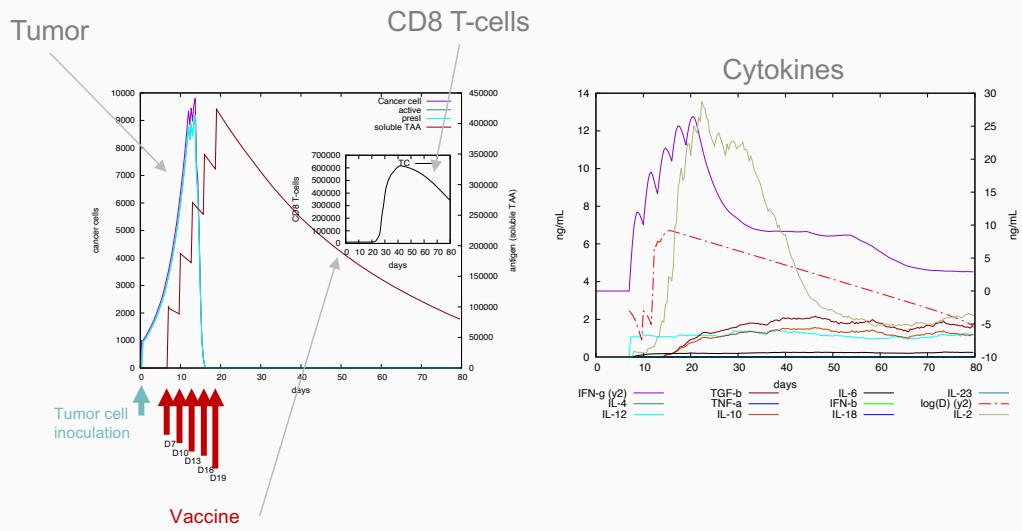
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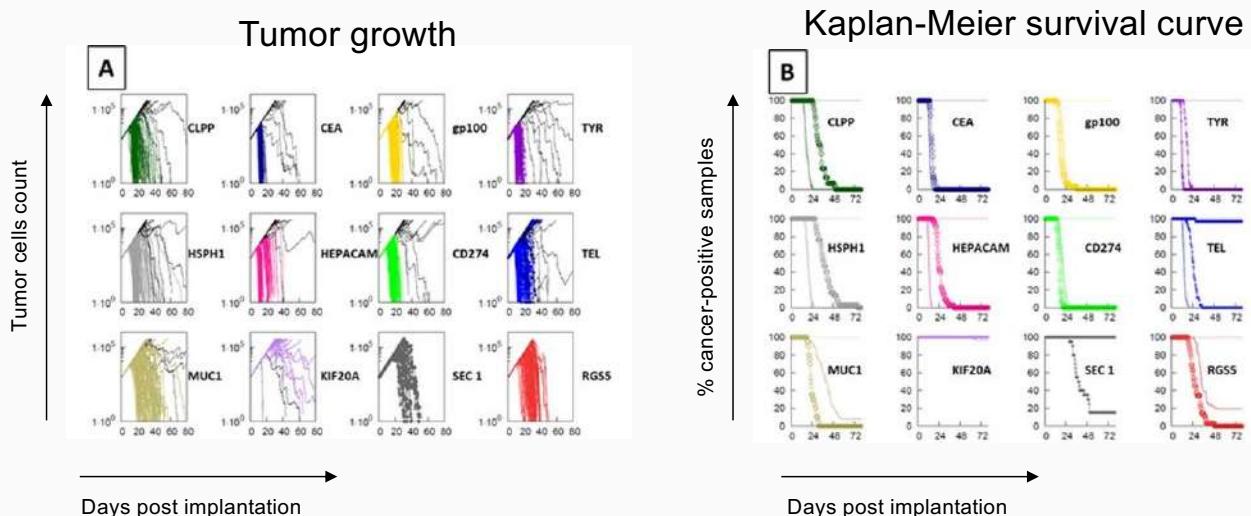
PROTOCOL	CLPP	Threshold for Strong binding peptides 0.500		Thal(h)	%Rank_S 1-log50k	Aff(nM)	%Rank_eff	Combined	combined %rank	Bin	dLevel
		HLA	peptide	Identity Pred							
HLA-A*02:01											
8	CLPP	HLA-A*02:01	ILDKVVLVHL	CLPP 0.899	6.54	0.40	0.703	24.98	1.00	0.742	0.77
	CLPPV	HLA-A*02:01	ILDCVVLVHL	CLPPV 0.879	5.36	0.60	0.757	13.90	0.80	0.781	0.75
9											
	CEA	HLA-A*02:01	IMIGVVLGVG	CEA 0.977	29.43	0.01	0.851	5.02	0.15	0.876	0.05
	CEAV	HLA-A*02:01	IIIGALVGV	CEAV 0.927	9.10	0.20	0.795	14.23	0.80	0.789	0.5
10											
	gp100	HLA-A*02:01	MLGTHHTMEV	gp100 0.897	6.40	0.40	0.812	7.67	0.30	0.829	0.32
	gp100V	HLA-A*02:01	MLGTHAMLV	gp100V 0.794	3.00	1.20	0.747	15.38	0.80	0.756	0.86
	tyrosinase	HLA-A*02:01	MLLAVLYCL	tyrosinase 0.945	12.30	0.12	0.840	5.62	0.17	0.861	0.16



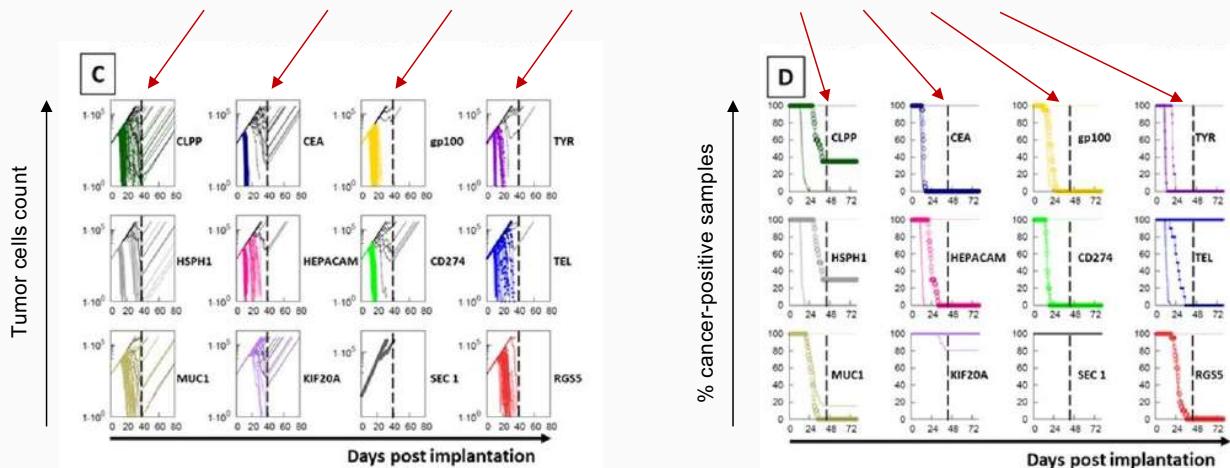
PROTOCOL	Secernin 1	HLA-A*02:01										combined %rank	Bin	dLevel	
		HIA-A*02:01	KIF20A	0.686	1.84	2.50	0.672	34.69	1.50	0.675	1.63				
18	SECV1	HLA-A*02:01	KMDAEHPGL	Secerni 0.512	1.03	4.50	0.643	47.73	1.50	0.617	1.73	<= WB			
	SECV2	HLA-A*02:01	KMDDEHPGL	SECV1 0.492	0.98	4.50	0.532	158.63	3.00	0.524	3.21				
HLA-A*03:01															
19	RG55	HIA-A*03:01		LASFKSFLK	RG55	0.177	0.40	3.50	0.542	142.57	0.80	0.469	0.95	<= WB	
		HIA-A*03:01		LAAFKSFLK	RG55V	0.220	0.46	3.00	0.563	113.17	0.80	0.494	0.94	<= WB	



Prediction of cross-reactive antitumor T cell response



CD8 (virtual) ablation experiment



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Conclusions

- Computer simulations and immuno-informatics prediction tools enable in-silico trials
- They provide comprehensive control over the configuration of virtual experiments and in-depth analysis of the outcomes. For example, this includes
 - selecting the haplotype
 - determining the vaccine's subunits
 - assessing the impacts of modifying the injection schedule
 - conducting dose-escalation studies
- While they represent an approximation of reality, they can serve as a valuable tool in the clinical optimization of immunotherapies

Thank you for listening

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