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# Supplemental information

## A multi-omics systems vaccinology resource

to develop and test

## computational models of immunity

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#### **Supplementary Figures**



**Figure S1: CMI-PB central database schema.** The database schema is divided into information, experimental data, and metadata tables. The information tables capture subject and sample information, while the experimental data tables capture experimental data for each omics. In the *rnaseq* table, each row represents an Ensembl gene identifier (versioned\_ensembl\_gene\_id) along with specimen id (specimen\_id) and values for measured analytes (raw\_count and tpm\_count). The *ab\_titer* table contains a row for each specimen (specimen\_id), with the names and values for measured analytes (isotype, is\_antigen\_specific, antigen, ab\_titer, unit, and lower\_limit\_of\_detection) for the antibody titer experiment. The *olink* table contains a row for each specimen (specimen\_id), with the names and values for measured analytes (olink\_id and protein\_expression). Each row in the *live\_cell\_percentages* table represents a specimen (specimen\_id) along with the names and values for measured analytes (cell\_type\_name and percent\_live\_cell). Information on the mapping between olink\_id and uniprot\_id can be extracted using the *olink\_prot\_info* table. The metadata tables capture information about ID mapping between CMI-PB data and external databases. The gene, transcript, and protein tables map Ensembl gene, transcript, and protein IDs. The *live\_cell\_percentages\_info* and *cell\_type* tables provide information about gating information and the experimental technique used to run cell frequency experiments.



Figure S2: Plot of antibody titer data prior to normalization. Individual plots show the pre- and post-immune response of IgG and its subtypes against antigens. The average titer values differed between the train and test datasets.



**Figure S3: Plot of antibody titer data after normalization.** We performed separate standardization of the antibody data in the train and test datasets, using the baseline median as the normalization factor. This approach allowed more direct comparisons of the normalized datasets between the train and test datasets.



**Figure S4: Identification of published methods to predict vaccine responses.** The literature search identified previously presented methods for predicting vaccine responses. First, 40 papers were selected from the literature search using keyword searches and relevant references in other papers. From the 40 papers, 13 studies were confirmed to present relevant prediction methods based on a literature search for published vaccine response prediction methods. For several of the 13 relevant studies, multiple prediction methods were presented. In total, 24 prediction methods were implemented from 10 relevant 13 studies (mentioned in **Table S1**). The prediction methods that obtained significant results for both performance metrics in the CMI-PB train datasets were also evaluated in the CMI-PB test dataset used for the first challenge.

## Supplementary Tables

	Task Title	Task statement	Biological and clinical significance	Reference
		Antibody le	vel tasks	•
1	IgG-PT_D14	IgG levels against Pertussis Toxin (PT) antigen in plasma on 14 days post booster vaccination	Vaccine-induced immunity for <i>B. pertussis</i>	1
2	IgG-FHA_D 14	IgG levels against Filamentous hemagglutinin (FHA) antigen in plasma on 14 days post booster vaccination	Vaccine-induced immunity for <i>B. pertussis</i>	1
3	IgG-Pertacti n_D14	IgG levels against Pertactin antigen in plasma on 14 days post booster vaccination	Vaccine-induced immunity for <i>B. pertussis</i>	Figure 5 in <sup>1</sup>
4	IgG1-PT_D1 4	IgG1 levels against PT antigen in plasma on 14 days post booster vaccination	Vaccine-induced immunity for <i>B. pertussis</i>	Figure 5 in <sup>1</sup>
5	IgG1-FHA_ D14	IgG1 levels against FHA antigen in plasma on 14 days post booster vaccination	Vaccine-induced immunity for <i>B. pertussis</i>	Figure 5 in <sup>1</sup>
6	IgG4-PT_D1 4	IgG4 levels against PT antigen in plasma on 14 days post booster vaccination	Signature of Acellular Pertussis (aP) vs. whole-cell Pertussis (wP) response	Figure 5 in <sup>1</sup>
7	IgG4-FHA_ D14	IgG4 levels against FHA antigen in plasma on 14 days post booster vaccination	Signature of aP vs. wP response	Figure 5 in <sup>1</sup> 1
		Cell frequen	icy tasks	
8	Plasmablast_ D7	Plasmablast cells on day 7 post-booster vaccination	Vaccine-induced immunity for <i>B. pertussis</i>	Figure 3 in <sup>1</sup>
9	CD4TCM_D 3	CD4 TCM cells on 3 days post booster vaccination	Vaccine-induced immunity for <i>B. pertussis</i>	Figure 3 in <sup>1</sup>
10	Monocytes_ D3	Monocytes on day 1 post-booster vaccination	Vaccine-induced immunity for <i>B</i> . <i>pertussis</i>	Figure 3 in <sup>1</sup>
		Gene express	sion tasks	
11	CCL3_D3	CCL3 on day 3 post booster vaccination	Signature of aP vs. wP response	Figure 7 in <sup>1</sup>
12	IL6_D3	IL6 on day 3 post-booster vaccination	Signature of aP vs. wP response	Figure 7 in <sup>1</sup>

#### Table S1: List of prediction tasks and their biological significance.

13	NFKBIA_D 7	NFKBIA at day 7 post booster vaccination	Signature of aP vs. wP response	Figures 6 and 8 in <sup>1</sup>
14	XIST_D14	XIST on day 14 post-booster vaccination	Biological sex-specific marker	2

**Table S2: Implemented prediction methods for the first challenge.** Models 1-24: Supervised methods based on published vaccine response prediction studies. Models 25-43: unsupervised methods based on Multi-omics dimension reduction.

	Model title	Short Description	Method type	Output score	Experimental data	Refer ence
1	avey_2017_gene _sig	9-gene signature (RAB24, GRB2, DPP3,ACT B, MVP, DPP7, ARPC4, PLEKHB2, and ARRB1)	Classification (Diagonal Linear Discriminant Analysis: DLDA)	geneset Experime signature score PBMC an gene expr subset fre antibody to Pathogen virus	Experimental data: PBMC and whole blood gene expression, cell subset frequencies, antibody titers Pathogen: Influenza virus	Avey et al. 2017
2	avey_2017_M54	B-cell signaling Blood transcripto me module (BTM; Module M54)	Classification (DLDA)	geneset signature score		
3	avey_2017_M42	Platelet activation (III) (Module M42)	Classification (DLDA)	geneset signature score		
4	avey_2017_M33	Inflammator y response Module (M33)	Classification (DLDA)	geneset signature score		
5	kotliarov_2020_ TGSig	10 gene signature (C2orf63, CD101, ENPP1, RETN, SMC1A, ADAM12, EPHB1, PAPSS2,	gene signature score	geneset signature score	Experimental data: PBMC and whole blood gene expression, cell subset frequencies, cell surface proteins, antibody titers Pathogen: Influenza virus, Yellow fever virus	Kotli arov et al. 2020 2

		LONP2, C15orf57)				
6	kotliarov_2020_ SLE-Sig	Systemic lupus erythematos us (SLE-Sig) gene signature	gene signature score	geneset signature score		
7	kotliarov_2020_I FN-I-DCact	IFN-I-DCac t gene signature	gene signature score	geneset signature score		
8	tsang_2014_DL DA_top2	DLDA model (2 cell populations	Classification (DLDA)	Class probabilities	<b>Experimental data:</b> PBMC gene expression, cell subset frequencies, antibody titers	Tsan g et al. 2014
9	tsang_2014_DL DA_top5	DLDA model (5 cell populations)	Classification (DLDA)	Class probabilities	<b>Pathogen:</b> Influenza virus	
10	fourati_2015_Bio Age	BioAge	Classification (Naïve Bayes)	Class probabilities	Experimental data: Whole blood gene	Four ati et al. 2015 4
11	fourati_2015_M1	M1 of BioAge	Classification (Naïve Bayes)	Class probabilities	expression, cell subset frequencies, serum proteins, antibody titers	
12	fourati_2015_M1 +M16	M1+M16 of BioAge	Classification (Naïve Bayes)	Class probabilities	<b>Pathogen:</b> Hepatitis B virus	
13	fourati_2015_NB	Naïve Bayes classifier (15 DEGs)	Classification (Naïve Bayes)	The ratio between the post-probabilit ies and the log odds in the logistic model		
14	fourati_2015_LR	Logistic regression (4 cell populations)	Regression (logistic)	The ratio between the post-probabilit ies and the log odds in the logistic model		

15	furman_2013_ag e	gene signature scores Age	Regression (Elastic net)	age	Experimental data: Whole blood gene expression, cell subset frequencies serum cytokines, antibody titers, hemagglutinin peptides Pathogen: Influenza virus	Fur man et al. 2013 5	
16	iulio_2021_HBV _transfer_sig	HBV pre-vaccine transfer signature	Classification (Random Forest)	geneset signature score	Experimental data: PBMC and whole blood gene expression Pathogen: Influenza virus, Hepatitis B virus, Mycobacterium tuberculosis	Experimental data: PBMC and whole blood gene expression	Iulio et al. 2021 6
17	iulio_2021_Inf_ M_transfer_sig	Influenza M pre-vaccine transfer signature	Classification (Random Forest)	geneset signature score			
18	iulio_2021_Inf_F _transfer_sig	Influenza F pre-vaccine transfer signature	Classification (Random Forest)	geneset signature score			
19	iulio_2021_TB_t ransfer_sig	TB pre-vaccine transfer signature	Classification (Random Forest)	geneset signature score			
20	fourati_2021_RF	Random Forest model (top 500 varying genes)	Classification (Random Forest)	Class probabilities	Experimental data: PBMC and whole blood gene expression, cell surface proteins, cell subset frequencies, antibody titers Pathogen: Influenza, smallpox, Yellow fever virus. Pneumococcal	Four ati et al. 2021 7	
21	bartholomeus 20	23	Classification	Class	meningococcal Experimental data:	Bart	
	18_gene_sig	differentiall y expressed genes	(Random Forest)	probabilities	Whole blood gene expression, absolute numbers of white blood cells, red blood cells,	holo meu s et al.	

22	bartholomeus_20 18_NB	Naïve Bayes classifier (first 5 PCs)	Classification (Naïve Bayes)	Class probabilities	and platelets, antibody titers <b>Pathogen:</b> Hepatitis B virus	2018 8
23	qui_2018_gene_s ig	55 up- and 15 down-regul ated genes	gene signature score	geneset signature score	Experimental data: PBMC gene expression, antibody titers Pathogen: Hepatitis B virus	Qui et al. 2018 9
24	franco_2013_gen e_sig	49 genes correlated with antibody response	gene signature score	geneset signature score	Experimental data: Whole blood gene expression, whole-genome genotyping, antibody titers Pathogen: Influenza virus	Fran co et al. 2013 <sup>10</sup>
25	jive.elastic_net_c v	Elastic net regression with cross-valida tion	Regression (Elastic net)		Experimental data: PBMC gene expression, cell frequencies, antibody titers	Lo ck et al. 20
26	jive.elastic_net	Elastic net regression	Regression (Elastic net)			13
27	jive.lasso_cv	lasso regression with cross-valida tion	Regression (lasso)			
28	jive.lasso	lasso regression	Regression (lasso)			
29	jive.lr	Linear regression	Regression (linear)			
30	baseline		Regression (lasso)		Baseline tasks, clinical and demographic features	
31	MCIAbasic		Regression (lasso)		Experimental data: transcriptome, cell	

			frequencies, antibody levels, cytokines	
32	MCIAplus	Regression (lasso)	Factors from the MCIAbasic model and features from the baseline model	

**Table S3**. **Summary of questions asked by contestants on the Solutions center.** In total, we received 23 questions that we divided into three categories.

Category of Questions	Number of Questions	Summary
Data access and format	14	Most questions were raised to know details about how contestants can access more data on the analytes' names, how analytes were measured, and the description of the measured unit. Several questions raised issues, like how QC was performed by Olink assay and the role of limit detections on filtering analyte values. Contestants found multiple proteins with duplicate expressions (i.e., IL5). All these questions have been assessed and resolved by the CMI-PB team. Several questions are raised to inquire about difficulties in accessing data API. The CMI-PB team provided direct file download options to contestants. The major concern discussed was antibody titer data was not predictive. Ab titer data normalization was performed using the median at day 0 as a normalization factor for CMI-PB data. Normalization was separately performed on training and test datasets.
Data processing	6	Questions were raised on how to deal with missing values in the data. Missing values arose if analyte measurements were below the limit of the detection threshold and samples needed to be included for subjects to make a complete multi-omics dataset. Out of two that worked with machine learning approaches, one decided to work only with available analyte measurements, and another team worked on data imputation after the decision in a regular meeting.
Miscellaneous (Contest, website)	3	Other questions were specifically asked about how to access challenge tasks, and contestants mentioned that they could not access information about data files on the website.

## Table S4. Cohort demographic information

Subject id	Infancy vaccination	Biological sex	Ethnicity	Race	Year of birth	Date of boost
61	wP	Female	Not Hispanic or Latino	Unknown or Not Reported	1/1/1987	4/8/2019
62	wP	Female	Not Hispanic or Latino	Asian	1/1/1993	11/26/2018
63	wP	Female	Not Hispanic or Latino	White	1/1/1995	11/26/2018
64	wP	Male	Not Hispanic or Latino	Asian	1/1/1993	11/26/2018
65	wP	Male	Not Hispanic or Latino	White	1/1/1990	12/3/2018
66	wP	Female	Not Hispanic or Latino	Black or African American	1/1/1976	12/3/2018
67	wP	Female	Hispanic or Latino	White	1/1/1972	1/28/2019
68	wP	Male	Hispanic or Latino	White	1/1/1972	1/28/2019
69	wP	Female	Hispanic or Latino	White	1/1/1990	1/28/2019
70	aP	Male	Not Hispanic or Latino	American Indian/Alaska Native	1/1/1998	1/28/2019
71	aP	Female	Not Hispanic or Latino	White	1/1/1998	1/28/2019
72	wP	Female	Not Hispanic or Latino	White	1/1/1991	2/25/2019
73	wP	Female	Not Hispanic or Latino	White	1/1/1995	2/25/2019
74	wP	Female	Not Hispanic or Latino	White	1/1/1995	2/25/2019
75	aP	Female	Not Hispanic or Latino	Native Hawaiian or Other Pacific Islander	1/1/1998	2/25/2019
76	aP	Female	Not Hispanic or Latino	Asian	1/1/1998	2/25/2019
77	wP	Male	Not Hispanic or Latino	White	1/1/1988	3/18/2019
78	wP	Female	Not Hispanic or Latino	White	1/1/1993	3/18/2019
79	wP	Male	Not Hispanic or Latino	White	1/1/1987	3/18/2019
80	wP	Female	Not Hispanic or	Asian	1/1/1992	3/18/2019

			Latino			
81	wP	Male	Not Hispanic or Latino	White	1/1/1993	3/18/2019
82	aP	Female	Not Hispanic or Latino	More Than One Race	1/1/1998	3/18/2019
83	aP	Female	Not Hispanic or Latino	White	1/1/1999	4/8/2019
84	aP	Female	Not Hispanic or Latino	More Than One Race	1/1/1997	4/8/2019
85	aP	Female	Hispanic or Latino	White	1/1/2000	4/29/2019
86	aP	Female	Not Hispanic or Latino	Asian	1/1/1998	4/29/2019
87	aP	Male	Not Hispanic or Latino	Asian	1/1/2000	4/29/2019
88	aP	Male	Not Hispanic or Latino	Asian	1/1/1900	4/29/2019
89	aP	Female	Not Hispanic or Latino	Asian	1/1/1997	6/3/2019
90	aP	Female	Not Hispanic or Latino	Asian	1/1/1999	6/3/2019
91	aP	Male	Unknown	Unknown or Not Reported	1/1/1998	6/3/2019
92	aP	Female	Hispanic or Latino	White	1/1/2000	6/24/2019
93	aP	Female	Not Hispanic or Latino	More Than One Race	1/1/1996	6/24/2019
94	aP	Male	Not Hispanic or Latino	Unknown or Not Reported	1/1/1999	6/24/2019
95	aP	Female	Hispanic or Latino	Unknown or Not Reported	1/1/1998	6/24/2019
96	aP	Male	Hispanic or Latino	Unknown or Not Reported	1/1/2000	6/24/2019

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