

Supporting information

Machine learning assisted prediction and generation of antimicrobial peptides

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Table S1: The physiochemical peptide features selected for model training.

Category	Property
Physiochemical	Overall charge (at pH 7)
	Charge density (at pH 7)
	Molecular weight
	Isoelectric point
	Aromaticity
	Gravy
	Boman index
	Aliphatic index
	Hydrophobicity profile – slope
	Hydrophobicity profile – intercept
	Hydrophobicity profile – moment
Secondary Structures	α -helix (%)
	β -sheets (%)
	Turns (%)

Table S2: The Table below shows the evaluation metrics (accuracy, precision, recall, F1, ROC AUC and confusion matrix's true and false negatives and positives) of four ML models across 5 Cross validation datasets and their respective mean and standard deviation.

<i>XGBOOST</i>	<i>1</i>	<i>2</i>	<i>3</i>	<i>4</i>	<i>5</i>	<i>mean</i>	<i>Std dev</i>
<i>Accuracy</i>	0.87564	0.87564	0.87564	0.84987	0.89059	0.87347	0.01314
<i>Precision</i>	0.87	0.86634	0.88144	0.87097	0.88833	0.87542	0.00818
<i>Recall</i>	0.88325	0.88833	0.86802	0.82234	0.89286	0.87096	0.02571
<i>F1 score</i>	0.87657	0.87719	0.87468	0.84595	0.89059	0.873	0.01466
<i>ROC AUC</i>	0.94906	0.93257	0.94553	0.92308	0.95944	0.94194	0.01276
<i>True Negatives</i>	171	170	174	172	175	172.4	1.85472
<i>False Positives</i>	26	27	23	24	22	24.4	1.85472
<i>False Negatives</i>	23	22	26	35	21	25.4	5.08331
<i>True Positives</i>	174	175	171	162	175	171.4	4.92341
<i>Random Forest</i>	<i>1</i>	<i>2</i>	<i>3</i>	<i>4</i>	<i>5</i>	<i>mean</i>	<i>Std dev</i>
<i>Accuracy</i>	0.85279	0.78934	0.82234	0.80407	0.88041	0.82979	0.033
<i>Precision</i>	0.83575	0.77941	0.82902	0.79412	0.87817	0.82329	0.03457
<i>Recall</i>	0.87817	0.80711	0.81218	0.82234	0.88265	0.84049	0.03299
<i>F1 score</i>	0.85644	0.79302	0.82051	0.80798	0.88041	0.83167	0.03215

<i>ROC AUC</i>	0.93561	0.88477	0.92489	0.88861	0.93544	0.91386	0.02256
<i>True Negatives</i>	163	152	164	154	173	161.2	7.57364
<i>False Positives</i>	34	45	33	42	24	35.6	7.39189
<i>False Negatives</i>	24	38	37	35	23	31.4	6.52993
<i>True Positives</i>	173	159	160	162	173	165.4	6.28013
<i>KNN</i>	<i>1</i>	<i>2</i>	<i>3</i>	<i>4</i>	<i>5</i>	<i>mean</i>	<i>Std dev</i>
<i>Accuracy</i>	0.64975	0.63706	0.63706	0.66667	0.65394	0.64889	0.01116
<i>Precision</i>	0.65775	0.63918	0.63235	0.67742	0.66129	0.6536	0.01615
<i>Recall</i>	0.62437	0.62944	0.65482	0.63959	0.62755	0.63516	0.01108
<i>F1 score</i>	0.64063	0.63427	0.64339	0.65796	0.64398	0.64405	0.00776
<i>ROC AUC</i>	0.6632	0.68379	0.68053	0.70178	0.69517	0.68489	0.01328
<i>True Negatives</i>	133	127	122	136	134	130.4	5.1614
<i>False Positives</i>	64	70	75	60	63	66.4	5.38888
<i>False Negatives</i>	74	73	68	71	73	71.8	2.13542
<i>True Positives</i>	123	124	129	126	123	125	2.28035
<i>SVM</i>	<i>1</i>	<i>2</i>	<i>3</i>	<i>4</i>	<i>5</i>	<i>mean</i>	<i>Std dev</i>
<i>Accuracy</i>	0.533	0.58883	0.59645	0.58015	0.59796	0.57928	0.02399
<i>Precision</i>	0.6383	0.81818	0.85185	0.75807	0.83929	0.78114	0.07835
<i>Recall</i>	0.15228	0.22843	0.2335	0.23858	0.2398	0.21852	0.03336
<i>F1 score</i>	0.2459	0.35714	0.36653	0.36293	0.37302	0.34111	0.04788
<i>ROC AUC</i>	0.54719	0.63929	0.6485	0.57472	0.60212	0.60236	0.03821
<i>True Negatives</i>	180	187	189	181	188	185	3.74166
<i>False Positives</i>	17	10	8	15	9	11.8	3.54401
<i>False Negatives</i>	167	152	151	150	149	153.8	6.67533
<i>True Positives</i>	30	45	46	47	47	43	6.54217

Table S3: Example of engine output for random peptide generation and their relative AMP score as determined using local and online prediction tools.

Sequence	Score
HFEIIDAMMVGKMEI	0.140870
ENHTYWDHIVDGDTR	0.889225
GVKQQSWSDSQSVE	0.544579
WTTLNNSWRQGIIKY	0.532396
GWRYQHTFVHQLASM	0.844666
...	...
KPNDALDTWHMRETD	0.203518
PSMDWDEEGHTMFAP	0.051499
PYNLRRTTMVVKGRKM	0.419810
IPASLYIHMQQVTFN	0.047382
WATMEPRWRAMEGIH	0.054740

Table S4: Snapshot of top 16 motifs identified using design engine.

'W----T',
'PY',
'M---M',
'L--A',
'Q-----V',
'Q-----S',
'FY',
'QQ',
'YM',
'M-Y',
'YQ',
'W-----T',
'WR',
'A-M',
'G---Q',
'W-----P',

Table S5: Snapshot of final list of peptides generated using top motifs with length 15 and batch size 25 with their respective AMP predicted score.

Sequence	Score
YFDYSNRWRYGILYR	0.082126
HYPDRALYVEFMVMW	0.072012
MPPWMLWIQWVPYP	0.082951
QLMWQWVMSQKGWSV	0.141375
PLNNVFYVPYDDWMD	0.080597
VMMQYRMIVYGTAWG	0.550192
LGRMFWDQYLFVHN	0.024660
RGLKTIDPYVYSHRT	0.876791
AWYHHFLWNHQYGPY	0.097113
WEKKRMVNQLTSMPN	0.050804
HAGWNWYVDTPFKS	0.869051
VMWQWRETVAQSNG	0.030749
GWTQMMPGMTVQSF	0.936140
YQARWVQVELQVPFF	0.072117
LIWAVRLSHQTADIA	0.715828
HLQVSQAQIMWSHYI	0.128904
PMMFDQEQLMVVDVY	0.129583
WKPAKMWQWQNYVQP	0.670909
MLAVMPQSPVDFSW	0.063223
WQDLENDARSNFKYT	0.182409
MTVQWEVQKSSDTMQ	0.026656
AYWRVRLTILMQWPY	0.092487
QKRVVQFQRSSPYPS	0.622008
QFRWWGLVSTPVPW	0.758741

Table S6: List of peptides generated using algorithm and their respective antibacterial activities in terms of exact MIC value in μM .

Peptide name	Sequence	M.W.	Charge	<i>E.</i> <i>Faecalis</i> <i>ATCC19433</i>	<i>S.</i> <i>Aureus</i> <i>ATCC25923</i>	<i>K.</i> <i>Pneumoniae</i> <i>ATCC43816</i>	<i>A.</i> <i>Baumannii</i> <i>AB5075</i>	<i>P.</i> <i>Aeruginosa</i> <i>PAOI</i>	<i>E.</i> <i>Aerogenes</i> <i>ATCC13048</i>
MLWH01	LAKIVPHKIGKQLGT	1601.0111	4						
MLWH02	ALFLGGIWKKLPKAI	1653.0463	4				100	100	
MLWH03	GKIIGIKVATFLKA	1570.0303	4			200			
MLWH04	LLOKIFVVKGKAGLK	1640.0833	5						
MLWH05	GLWKIKLAFGKMFAK	1736.0293	5	25	100	25	12.5	25	25
MLWH06	GKIFGKFLALIKMW	1763.0652	4	12.5	25		25	200	200
MLWH07	GGAFFKLTKIHKKA	1648.0158	5			200	50	200	
MLWH08	VQKLLKFKIKITAKA	1727.1515	6	100		50	100	200	50
MLWH09	LLGFTAFIKGQIKTA	1605.9577	3	50	50	100	200	200	100
MLWH10	GIVNVTGAITKLIK	1537.9889	3						
MLWH11	AWLKFKKKFGLFAKI	1823.1305	6	25	100	12.5	12.5	25	12.5
MLWH12	LVLQGGIVKAVVKAIN	1563.0205	3						
MLWH13	ILKLVANKIFHAVVF	1710.0677	3						
MLWH14	WFKAIAGLGGKFIK	1661.0151	4	25	25	25	25		25
MLWH15	AVLIAKLFIHKLNIK	1734.1363	5	200		200	100	200	
MLWH16	GGCCWRCNRRVLR	1826.8514	5						
MLWH17	IAKVLNKFVFNIGIK	1702.0626	4						
MLWH18	FGNLATQVKALNG	1571.9120	3						
MLWH19	FLPLIILKFIGLLP	1723.1493	3	1.5	3		200		
MLWH20	LAFVKGILANVGKVK	1554.9944	4						
MLWH21	IKVLKLVTNKGNGI	1608.0416	4	200		200	50		200
MLWH22	TLVKAVAHIAKLKLF	1650.0677	4	100	100	50	12.5	200	200
MLWH23	KFITKFVKGGLGLTL	1620.0096	4			200	50	200	
MLWH24	TKFGQKVIKLIAQFL	1732.0731	4	6	12.5	6	3	25	6
MLWH25	ALKAKLKFPKGFLFK	1734.1040	6			100	100	200	200
MLWH26	GLLTGKKKLLGIIA	1524.0096	4				200		
MLWH27	VHKAFRKGSGAAILW	1638.9444	4			200	200		
MLWH28	FVTITLKNFAKKKVG	1692.0419	5			200	200		200
MLWH29	GIVKAQLNIATFLTK	1614.9791	3						
MLWH30	KIVFGLAKIGKFIVK	1659.0931	5						
MLWH31	FKLIKNVGALAFGT	1575.9472	3						
MLWH32	KGFAKMFRLKKLAR	1805.1306	7	12.5	6	12.5	3	25	25
MLWH33	NFKIHKVGIKLKGA	1655.0942	6				100		
MLWH34	LGAPLKWIKGKFAIP	1637.0151	4						
MLWH35	KHVFALGGKLVKLK	1683.0681	5			200	50		
MLWH36	AIVKKGWIKILAWVF	1770.1040	4	100	6	100	12	200	100
MLWH37	LGKVAHWGKLHVPKL	1681.0275	4						
MLWH38	FGLPKIKIGIKAIMI	1640.0543	4						
MLWH39	FVLTKATNGVLKALI	1585.9889	3						
MLWH40	KLKNKTAGIFKGQVK	1658.0325	6						
MLWH41	YFHLKKSFGKKLKR	1877.1483	7				100		
MLWH42	FKRHGIGKFPKYASK	1762.0126	6						
MLWH43	WFGKKFPQIACKRCC	1812.9075	5				25		
MLWH44	GKFFAKALFLAIVKT	1652.0147	4						
MLWH45	KLKGLIFAWFKGIIA	1703.0619	4	25	100	100	25	200	25
MLWH46	IGIKGAMPLKGIIKK	1565.0184	5						
MLWH48	GPLMGAFIKLLPKF	1658.0075	4	200		100	25	100	200
MLWH49	IAVNQLTKIAKVAVG	1522.9530	3						

Table S7: List 2 of peptides generated using algorithm and their respective antibacterial activities.

Peptide name	Sequence	M.W.	Charge	<i>E. Faecalis ATCC19433</i>	<i>S. Aureus ATCC25923</i>	<i>K. Pneumoniae ATCC43816</i>	<i>A. Baumannii AB5075</i>	<i>P. Aeruginosa PAO1</i>	<i>E. Aerogenes ATCC13048</i>
MLWH51	LGVPVFVKFFKFFGK	1787.062	5	100	200	100	12.5	50	200
MLWH52	KHFSGFNGFFKKFSK	1803.955	5			200	100		
MLWH53	LVLKIKKKAVKALWE	1765.167	5			100	100	200	
MLWH54	IIKIINVPNVIWNII	1760.104	2						
MLWH55	FIQMINIIHYIHF	1913.072	1						
MLWH56	GSKGSPFFSKKFRNF	1731.918	5						
MLWH57	QGWIKIILPHIRYIK	1876.153	4	25		12.5	12.5		25
MLWH58	AKKKVVAWTKWPPKI	1865.116	6			200	50	200	
MLWH59	KHGIPYIKKITNKI	1764.111	5				200		
MLWH60	KTFIKKIGTKIKKWT	1818.157	7			200	6.25	50	
MLWH61	PSKYAKIAIKFHKQI	1770.064	5			50	50	50	50
MLWH62	KGGVWFGRILKHQSK	1739.008	5						
MLWH63	RFVKGIGGHGKKMMY	1777.993	6						
MLWH64	FKTLKHKKFKRNPW	1969.186	7			200	50	50	
MLWH65	IQVGDIAYKKIKRIK	1771.116	5						
MLWH66	SKFAKNFKKKYKNKF	1904.112	8				200		
MLWH67	KIAKFVKKIIKPSIM	1742.133	6				25	50	
MLWH68	WVPVWPKKLKLTRG	1834.143	6				50	200	
MLWH69	IVPIKKIIKYKFQK	1857.23	6	200			200	200	
MLWH70	WKGKKKFFDKKKKWF	2027.195	8				50	200	
MLWH71	RQIDWYWKRGKGYAR	2052.125	6			200	100	200	
MLWH72	IMKKILINKIHNMK	1835.133	5				50		
MLWH73	FNKGKKFFFKFIKVV	1962.136	6	25	25	12.5	12.5	25	25
MLWH74	RMFFKFKKKFKKWGQ	2032.168	8	200	50	25	25	200	100
MLWH75	GAKFFKDFKKRDKV	1868.087	6						200
MLWH76	LHHIIIGIWKAIIYT	1789.074	2						
MLWH77	KKLHHPPFKHFSQKP	1904.066	5						
MLWH78	WKIVKKFKMKSHPYH	1914.042	5			200	200		
MLWH79	KKKFKITGHTKYDYG	1812.002	5			200			
MLWH80	FMKMWKVMGKHKKF	1865.069	7				100	200	
MLWH81	IHAITHIHKFDVIA	1717.037	2						
MLWH82	FSKIIGITLKGLTII	1615.04	3						
MLWH83	GFKKEFDKFWWKKKR	2056.149	6			100	200	200	
MLWH84	IIHLVIRWIGELATM	1763.025	1						
MLWH85	AKVWKIHKLKIKP	1819.262	8			100	25	100	
MLWH86	WIDIIEITRIIVHI	1845.121	0						
MLWH87	PHFIKIIIVQIIYIIN	1822.12	2						
MLWH88	KNRIHKHILKKIYI	1915.233	6				50	50	
MLWH89	KFFKKKLKDKTKDWI	1951.174	6				200		
MLWH90	KGFGKYKKGKKIYKK	1799.127	9						
MLWH91	HPILKIIIMDKHIY	1845.103	2						
MLWH92	FNKAGFKNSKWNKI	1808.018	6						
MLWH93	NHTFQIFMRKGNKKY	1910.007	5			100	100	>200	
MLWH94	KQQKKFFKKKKFFKE	2015.216	8						
MLWH95	DKFVKKPWKKKVKKY	1948.21	8						
MLWH96	KAFKGFWKKGKHKDK	1831.07	7						
MLWH97	FKLPFNKFKKSKRKA	1865.148	8				200		
MLWH98	PKWHKNKMFKFKRKf	2048.174	8		200	50	100	200	100
MLWH99	VIYSIGLRIFNKIRP	1787.09	4	50		50	50	200	50
MLWH100	MHKVDHIKKIMKTK	1848.092	5						

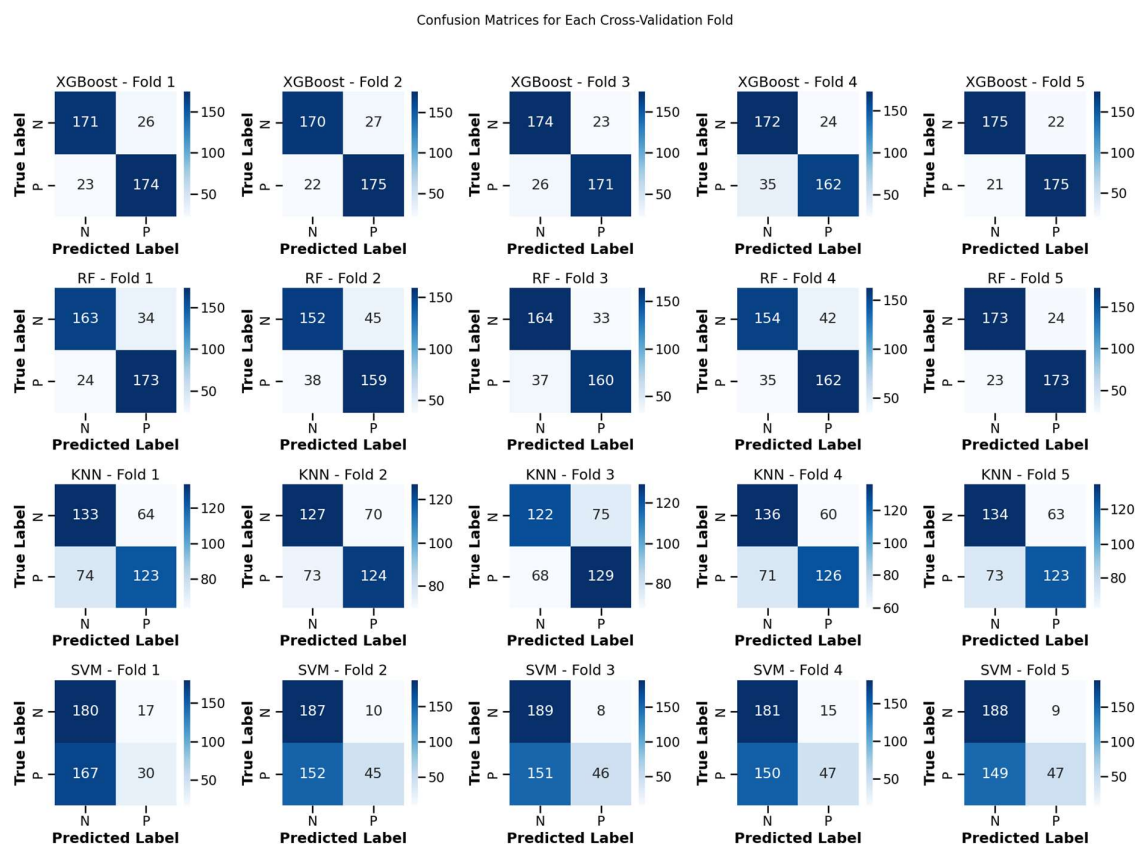


Figure S1: Confusion matrices for each Cross-Validation fold for four Models -XGBoost, RF, KNN, SVM

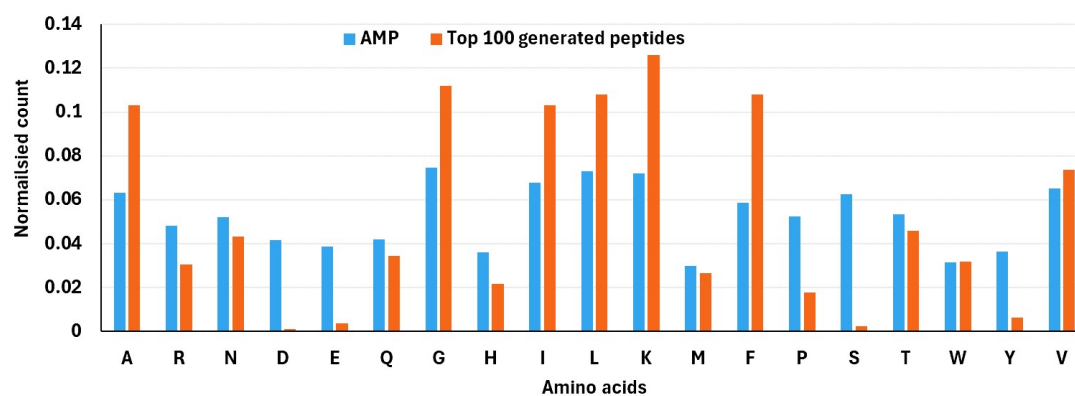


Figure S2: The comparison of amino acid composition of top 100 generated and predicted peptides to the training AMP dataset.

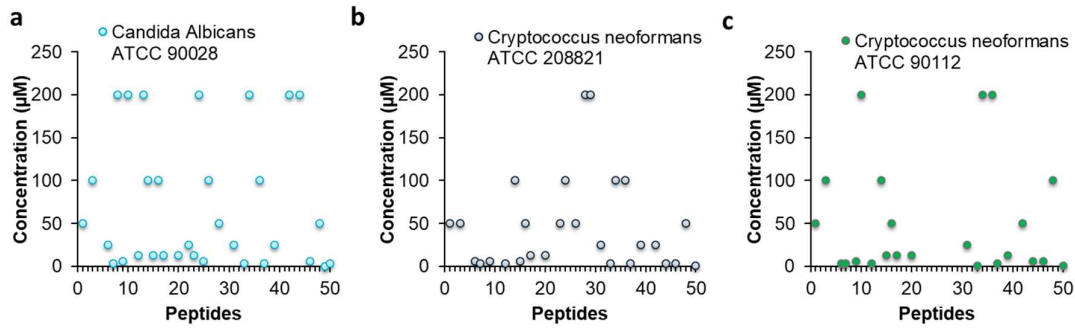


Figure S3: Antifungal activity of the peptide generated using algorithm. The scatter plot showing MIC values of various peptides against 3 different fungal strains, a) *Candida Albicans* ATCC90028, b) *Cryptococcus neoformans* ATCC208821, and c) *Cryptococcus neoformans* ATCC90112. The MIC testing experiments were performed in duplicates.

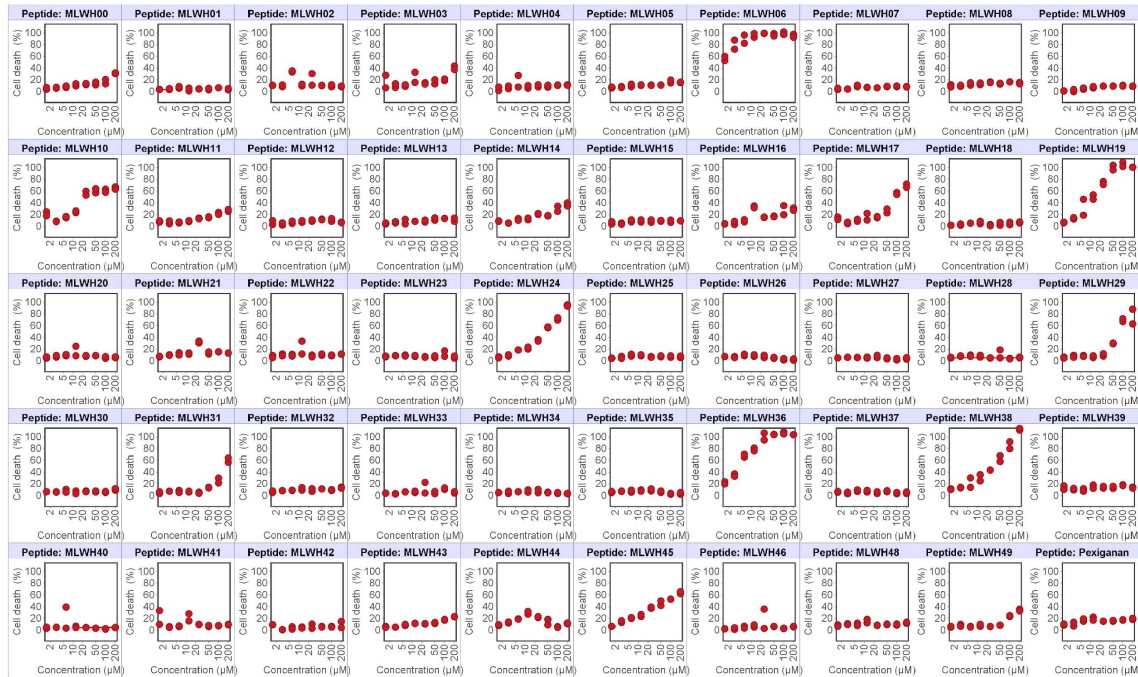


Figure S4: The graphs represent the hemolysis assay data for various peptides after incubation with whole human blood as a function of concentration. The experiment was performed in duplicates

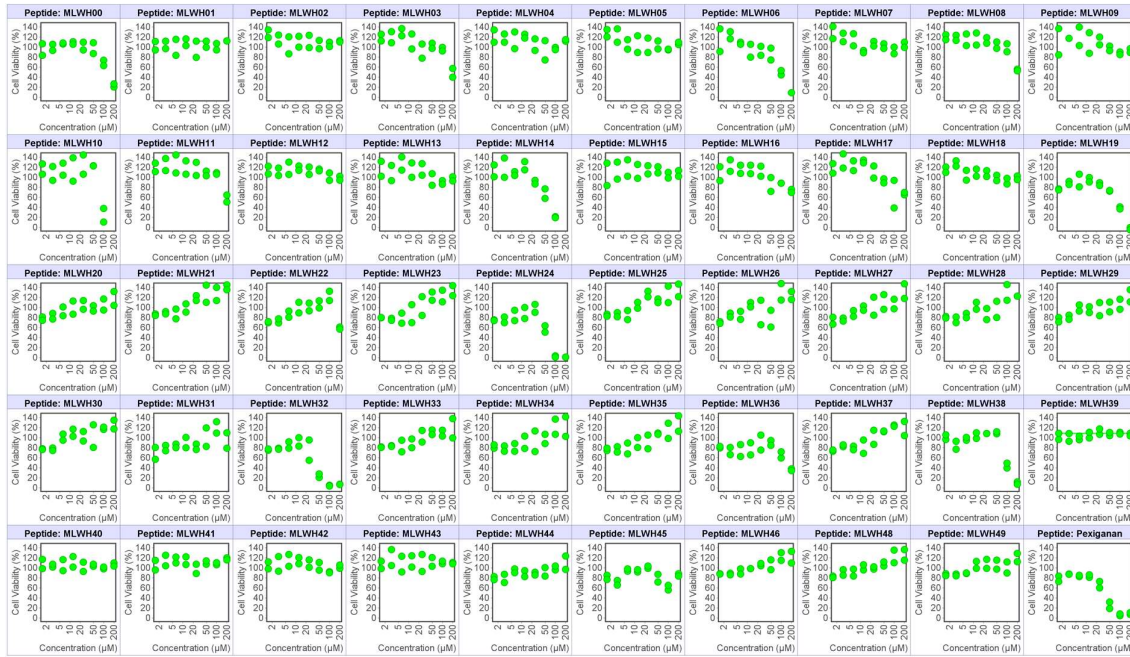


Figure S5: Cell viability of HEK-293 cells after 24 h incubation with various peptides as a function of concentration. The experiment was performed in duplicates