

Therapeutic Targets Prediction of Tumor Specific Antigen for Vaccine Design

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Abstract—The difficulty of treating cancer lies in their similarity with normal cells in terms of antigen expression. Cancer cells are transformed cells that gain immortality by mutagenic transformation of normal cells. The cross reactivity of surface antigens on tumor cells and normal cells makes them difficult to be identified by immune system. However, T cytotoxic cells, NK cells and dendritic cells play an important role in immune response to tumor cells. T cell epitopes of tumor specific antigens were predicted as vaccine candidates so as to increase immunity against tumor cell. Tumor cells express tumor specific antigen peptide on their surfaces which are recognised by T-cytotoxic cells and hence give immune response against tumor. Dendritic cell phagocytose the tumor cell and present their peptides to the T-helper cell and hence result in tumor rejection. MHC class I binding prediction tool and CD8⁺ immunogenicity prediction tools were used to predict class I epitopes. MHC II binding prediction tool and CD4⁺ immunogenicity prediction tool were used to predict class II epitopes. Population coverage of the epitopes of both MHC I and MHC II binding peptide was calculated in Indian population. A set of most population covered epitope was selected for each antigen. Epitopes were predicted for ACTN4, BRAF, CAMEL, CASP5, CASP8, CDC27, CDK4, CDKN2A, CTNBN1, EEF2, FLT3, GPNMB, HSPA1B, KRAS, MUM1, PAPOLG, TPI1, UBXD5. A combinatorial epitope prediction would be effective in designing multifunctional immune stimulation response against tumor specific antigen.

Introduction

Tumor is the abnormal mass of tissue when uncontrolled cell division takes place. Tumor antigen is the protein product of the tumor cell, which is identified by immune cell. By identification of tumor antigen immune cell respond to the tumor cells and try to eliminate these cells. Tumor antigen is classified into three different type, these are tumor specific antigens, shared antigen, unclassified[1]. Tumor specific antigen are further classified into substitutional mutation, alternative ORF, intron encoding, chromosomal translocation, internal tandem repeat[2]. There are no. of antigen in these like in substitution mutation ACTN4, BRAF, CASP8, CDC27 etc., in alternative ORF CASP5, CDKN2A, OGT etc., in intron encoding MUM1, in chromosomal translocation ABL-BCR, BCR-ABL, DEK-CAN etc., internal tandem repeat FLT3[3][4].

The tumor specific antigens are the main target of the immunogenic response against the tumor cell. These are the

tumor specific antigens (not present on the normal cell) so if immunotherapy using these as targets for immune cells, then there is no harmful effect of it on the normal cell[5].

MHC class I presents the short peptide on the surface of tumor cell. T cytotoxic cell recognise this peptide and MHC I complex and give immune response against it[6]. MHC class II present the short peptide on the surface of tumor cell. T helper cell recognise this peptide and MHC II complex[7]. Helper T cell never give direct immune response instead of that they help in the activation of cytotoxic T cell and macrophages to give immune response to target cell. Or the T helper cell stimulate B cells to secrete antibodies. APCs ingest microbes, degrade them, and export small part of it(antigen) to the cell surface with the help of mhc II. This MHCII and peptide complex then recognise by the receptors present on the helper cell surface known as CD4⁺ receptors[7]. But this is not the only signal which is required for the activation of T cell. It required one more signal for its activation. It must be given by one of these two signal either by stimulation by a cytokine or by a stimulatory signal that is between B7 and CD28. B7 found on the surface of APCs. Receptor protein CD28 found on the surface of helper T cell. So, when the first signal and one of second signal are received then the helper T cell becomes activated[8][9].

So, Here we have to predict the epitopes, which can bind to receptors of both T helper cell as well as T cytotoxic cell. There are lots of tool which can use to predict MHC I and MHC II binding peptides[10][11][12].

They give an accuracy of about 90-95%. Then we have to select those alleles which cover most of the population frequency. So the vaccine we want to form can be bind to the most HLA alleles. We can collect these data from different papers in which population frequency of different HLA alleles given[13].

Once the epitope selection takes place, then we can select a set of those epitope which cover atleast 80% of the population. So the vaccine we formed from these epitopes can be effective on the large population size.

METHODS

DATA COLLECTION

Accession id.oftumor specific antigen and their isomers were retrieve from TANTIGEN database. There were 45 tumor specific antigens and Consensus sequences were find by multiple sequence alignment of all the isomers of that tumor specific antigens. Seventeen antigens consensus sequence were selected which could be selected for the further prediction. For all other protein we either not able to get full length sequence or they have too short sequence so we can't predict targets from them.

In some cases the complete sequence of one isomer was selected that is known as reference sequence. The peptide part of the sequences were selected which is common in most of the isomers. From these sequences we have to predict the mhcI binding peptide andmhcII binding peptide.

Epitope prediction of T-cytotoxic cell

The antigen sequences are firstdevided into small peptide which binds to MHC1 then these MHC1-peptide complexes bind to the CD8⁺ receptor of T-cytotoxic cell. So we first predicted MHC1 binding peptide then CD8⁺ receptors binding.

MHC I binding peptide prediction

IEDB prediction tool Tepitool was used for the prediction of MHC I binding peptide. As this is a user friendly tool. And Different attributes were manully selected.

Alleles selection

A set of alleles selected on the bases of their population frequencies. Thirtyeight most abundant alleles set of HLA class I were used which covers almost 99% population of the world.

IC⁵⁰ value selection

The ic⁵⁰ value of all these alleles are more than 100 nm , so we select a threshold of less than or equal to 100 nm for best prediction results.

Prediction method selection

NetMHC1pan method was used for the prediction of mhcI binding peptides. This is the only method which predict peptide on the bases of IC⁵⁰ value.

CD8⁺ immunogenicity prediction

IEDB CD8+ immunogenicity prediction tool was used with the default setting. In this tool automatically masking of 1,2 and C-terminal of amino acid take place. Prediction score was calculated for all mhc I binding peptide and only those peptide were selected whose prediction score is more than zero.

Population coverage of T-cytotoxic epitope

POPULATION ANALYSIS tool was used for the calculation of population coverage of selected peptide and those peptide were selected which cover most of the population.

T-helper cell epitope prediction

Tumor antigen are the internal peptides so presented by mhcI but these tumor cell phegocytosis by dendritic cells which are antigen presenting cells . They present these antigens peptides to the T-helper cell.

MHC II binding peptides would predict followed by their CD4⁺ immunogenicity prediction.

MHC II binding peptide prediction

IEDB prediction tool Tepitool was used for the prediction of MHC II binding ppeptide and different attributes would select here.

Alleles selection

Twentysix most abundant alleles of MHC II were used. These alleles IC⁵⁰ value threshold cutoff selected here is 1000.

Prediction method selection

A set of methods was used for prediction of most promisiouse binding of HLA II alleles. Only those peptide were selected which bind to atleast 50% of alleles from these Twentysix alleles.

CD4⁺ immunogenicity prediction

IEDB prediction tool were used for the prediction of immunogenicity of the mhclI binding peptides. Immunogenicity score of all the peptide were predicted and only those peptide were selected which have immunogenicity score more than zero.

Population coverage

IEDB population coverage prediction tool was used for the analysis of population coverage in India. Those peptidewere selected which cover most population.

Combined population coverage of MHC1 and MHC11

Combined population coverage was calculated of the selected peptide of both mhcI and mhc11 by population analysis tool.

RESULTS

A set of epitopes was predicted for different antigen which have maximum population coverage in Indian subcontinent.

Antigen Name	Predicted epitopes of T-Helper and T-Cytotoxic cell	Total population coverage in Indian subcontinent
ACTN4	KALDFIASK, MTYVSSFYH, VQNFHISWK, MTLGMIWTI, LAFNALIHR, SSFYHAFSGAQAET, ALDFIASKGVKLVSI, LEINFNTLQTKLRLS, LQTKLRLSNRPAFMP, SNHIKLSGSPYTTV	93.51

CAMEL1	FLMAQGAML, RPWKRSWSA, CTSRCLSR, AVPLLRMEGAPAGP, FLMAQGAMLAAQERR, MLMAQEALAF LMAQ	RMAVPLLR, LMAQEALAF,	87.32	FLT3	RMPEAAPPV, CTYSPALNK, TAKSVTCTY, APPQHLIRVEGNLRV, RAMAIYKQSQHMTEV, TYQGSYGFRGLFLHS, VTCTYSPALNKMFCQ	KTYQGSYGF, RVRAMAIYK,	86.13
CASP5	SVLRAFAAR, LLYDTIFI, NVSWDRDR, GAHYDIVGMKRLQ, STFLVLMHSHGILEGI, RDMESVLRFAARPE, DFIAFCSTPHNVSW, LWVRDSPASLAVISS	KSFEVPQAK, SSTPHNVSW,	88.64	GPNMB	TIVEGILEV,ITFAVNNLIF, YVLNGTGS, YVFHTLGQY, TVISLLVYKKHKEYN, DPASPLRMANSALIS, LSVFLNRAKAVFFPG, DVDEMCLLTVRRTFN, GNVVRSKGLSVFLNR	RTFNGSGTY,	89.68
CASP8	IINNHFAK, ILTEVNYEV, MLEESNSLF, DALMLFQRLQEKRLM, RAQISAYRVMLYQIS, RGYCLIINNHFAKA, DLASLKFLSLDYIPQ	TVNNCVSYR, LSFLKELLF,	87.36	HSPA1B	LLLLDVAPL, RTTPSYVAF, VIAGLNVL, GLNVLRIINEPTAAA, QNKRAVRRRLTACER, TKDAGVIAGLNVLRI, GGVMTALIKRNSTIP, VGVFQHGKVEIAND	AEAYLGYPV,	86.34
CDC27	HSGHFVALK, RIYSYQMAL, LTPVVVTLW, TTLWHLQKDVALSVL, QTFKFTSLQNFNSCL, AAIWQALNHAYARDA, YAYAYTLGHEFVLT	CIFAEMFRR, NPHKRISAF,	91.5	KRAS	KSFEDIHY, LVREIRQYR, QYMRTEGEGF, ALTIQLIQNHVDEY, GFLCVFAINNTKSFE	CVFAINNTK, CLLDILDTA,	87.68
CDK4	HSGHFVALK, CIFAEMFRR,RIYSYQMAL, NPHKRISAF, PHSGHFVALKSVRVP, RISAFRALQHSYLHK, SYQMALTPVVVTLWY, DFGLARIYSYQMALT, RQFLRGLDFLHNCI	LTPVVVTLW,	94.41	MUM1	YAADISYPV, KMKGFTVSL,RSFEVGMLV, GMLVWHKHK, PLEELAYRRSLRVAL, IGWCVSLITDYRVRL, GFTVSLKSLKHFDC, KGAESHILRAILKSRK, LEYYAADIS YPVRKS	LSSSFTCEK,	91.56
CTNNB1	IMRTYTYEK, RLHYGLPVV, ATYAAAVLF, GQYAMTRAQVRRAAM, EKLLWTSRVLKVL, QALVNIMRTYTYEKL, WPLIKATVGLIRNLA, GLIRNLALCPANHAP	TTAPSLSGK, YAAAVLFRM,	91.62	PAPOLG	HLMPIITPA, KIFTFGSYR, YPNAAASTL, FRLTLRAVKLWAKRR, KYRHYIVLTASASTE, VESKIRVLVGNLERN, LRSLDIRCIRSLNGC, LGIIFRRVENAESVN	CTIPTVVGR, RSDFFQSFF,	87.11
EEF2	IMRTYTYEK, RLHYGLPVV, ATYAAAVLF, PLMMYISKMVPTSDK, GRVFSGLVSTGLKVR, GGIYGVLNRKRGHVF, CITIKSTAIISLFYEL, NRLYMKARPPDGLA	TTAPSLSGK, YAAAVLFRM,	85.41	TPI1	HLMPIITPA, KIFTFGSYR, YPNAAASTL, GELIGTLNAAKVPAD, ESDELIGQKVAHALA, FHFAALYISGQWPRL, VVAIGT, LRGWLKSNVDAVAQ	CTIPTVVGR, RSDFFQSFF,	86.73
				UBXD5	RTLEPIPLK, RLYPNGVPF, MAFMTRKLW, IPLKLYRNGIMMFDG, FPSELQRLYPNGVPF, KAALLLRARRAPKSS, QPDNTIG, DASAFAIFSTFPPTL	SAFEIFSTF, MTAEKFLNR,	88.45

Conclusion

Therapeutic target were predicted for vaccine formation. These can be further tasted in vitro then could be used in personalised medicine.

References

- [1] H. M. Zarour, A. DeLeo, O. J. Finn, and W. J. Storkus, "Categories of Tumor Antigens," 2003.
- [2] R. F. Wang, "Tumor antigens discovery: perspectives for cancer therapy.," *Mol. Med.*, 1997.
- [3] N. Renkvist, C. Castelli, P. F. Robbins, and G. Parmiani, "A listing of human tumor antigens recognized by T cells.," *Cancer Immunol. Immunother.*, 2001.
- [4] L. Novellino, C. Castelli, and G. Parmiani, "A listing of human tumor antigens recognized by T cells: March 2004 update," *Cancer Immunology, Immunotherapy*. 2005.
- [5] K. M. Mahoney, P. D. Rennert, and G. J. Freeman, "Combination cancer immunotherapy and new immunomodulatory targets," *Nature Reviews Drug Discovery*. 2015.
- [6] A. Williams, C. A. Peh, and T. Elliott, "The cell biology of MHC class I antigen presentation," *Tissue Antigens*. 2002.
- [7] P. Cresswell, "Editing peptide presentation to T cells," *Science*. 2017.
- [8] J. Neefjes, M. L. M Jongsma, and P. Paul, "Towards a systems understanding of MHC class I and MHC class II antigen presentation," *Nat. Rev. Immunol.*, 2011.
- [9] J. Neefjes, M. L. M. Jongsma, P. Paul, and O. Bakke, "Towards a systems understanding of MHC class I and MHC class II antigen presentation," *Nat. Rev. Immunol.*, 2011.
- [10] C. Lundegaard, O. Lund, S. Buus, and M. Nielsen, "Major histocompatibility complex class I binding predictions as a tool in epitope discovery," *Immunology*. 2010.
- [11] N. Rapin, O. Lund, M. Bernaschi, and F. Castiglione, "Computational immunology meets bioinformatics: The use of prediction tools for molecular binding in the simulation of the immune system," *PLoS One*, 2010.
- [12] M. Atanasova, A. Patronov, I. Dimitrov, D. R. Flower, and I. Doytchinova, "EpiDOCK: A molecular docking-based tool for MHC class II binding prediction," *Protein Eng. Des. Sel.*, 2013.
- [13] K. Cao, J. Hollenbach, X. Shi, W. Shi, M. Chopek, and M. A. Fernández-Viña, "Analysis of the frequencies of HLA-A, B, and C alleles and haplotypes in the five major ethnic groups of the United States reveals high levels of diversity in these loci and contrasting distribution patterns in these populations," in *Human Immunology*, 2001.