

Predicting proteasomal cleavage sites: a comparison of available methods

Patricia Saxová^{1,2}, Søren Buus³, Søren Brunak¹ and Can Keşmir^{1,4}

¹Center for Biological Sequence Analysis, BioCentrum-DTU, Technical University of Denmark, Lyngby, Denmark

²Institute of Biology and Ecology, P. J. Šafárik University, Kosice, Slovakia

³Institute for Medical Microbiology and Immunology, University of Copenhagen, Copenhagen, Denmark

⁴Theoretical Biology/Bioinformatics, Utrecht University, Utrecht, The Netherlands

Keywords: artificial neural network, cleavage site prediction, MHC class I epitope, proteasome, protein degradation

Abstract

The proteasome plays an essential role in the immune responses of vertebrates. By degrading intercellular proteins from self and non-self, the proteasome produces the majority of the peptides that are presented to cytotoxic T cells (CTL). There is accumulating evidence that the C-terminal, in particular, of CTL epitopes is cleaved precisely by the proteasome, whereas the N-terminal is produced with an extension, and later trimmed by peptidases in the cytoplasm and in the endoplasmic reticulum. Recently, three publicly available methods have been developed for prediction of the specificity of the proteasome. Here, we compare the performance of these methods on a large set of CTL epitopes. The best method, NetChop at www.cbs.dtu.dk/Services/NetChop, can capture ~70% of the C-termini correctly. This result suggests that the predictions can still be improved, particularly if more quantitative degradation data become available.

Introduction

Proteasomes are multisubunit proteases that play a central role in the degradation of proteins in the cell (1). Some degradation products of the proteasome are taken up by the transporter associated with antigen processing (TAP) and transferred into the endoplasmic reticulum. Here they can associate with newly synthesized MHC class I molecules. Recognition of such MHC-peptide complexes on the cell surface by activated cytotoxic T lymphocytes (CTL) is essential for the cellular immune responses (2).

The proteasome has at least three different catalytic activities: trypsin-like (i.e. cleavage after basic amino acids), chymotrypsin-like (i.e. cleavage after large, hydrophobic amino acids) and peptidyl-glutamyl-peptide-hydrolyzing activity (i.e. cleavage after acidic amino acids) (3). Since the overall enzymatic activity is the result of an interaction between these catalytic subunits, the cleavage-inhibiting or -enhancing motifs are quite complex. In the presence of IFN- γ , the three catalytic subunits of the proteasomes of vertebrates are replaced by their homologous subunits to form an 'immuno-proteasome' (4). The cleavage specificity of the constitutive proteasome and the immunoproteasome seems to be different (5,6), a factor that further increases the complexity of the enzymatic activity of the proteasome.

Due to the involvement of the proteasome in the generation of antigenic peptides it is of general interest to obtain additional insight into the specificity of the proteasome, and to predict which peptides will be generated from both pathogenic and human proteins. At the moment three proteasome cleavage prediction methods are publicly available on the Internet: PAPProC (www.paproc.de) developed at Tübingen University (7,8), MAPPP (www.mpiib-berlin.mpg.de/MAPPP/) developed at the Max-Planck Institute in Berlin (9,10) and NetChop (www.cbs.dtu.dk/services/NetChop/) developed at the Center for Biological Sequence analysis at the Technical University of Denmark (11).

PAPProC is a method for predicting cleavages by human proteasomes as well as wild-type and mutant yeast proteasomes. The influences of different amino acids at different positions are assessed using a stochastic hill-climbing algorithm (7) based on the experimentally *in vitro* verified cleavage and non-cleavage sites (8).

MAPPP is a method that combines proteasome cleavage prediction with MHC-binding prediction. FragPredict is the part of the MAPPP package that deals with the proteasome cleavage prediction. FragPredict consists of two algorithms. The first algorithm uses a statistical analysis of cleavage-

Correspondence to: C. Keşmir, Theoretical Biology/Bioinformatics, Utrecht University, Padualaan 8, 3584 CH Utrecht, The Netherlands. E-mail: C.Keşmir@bio.uu.nl

Transmitting editor: M. J. Bevan

Received 15 January 2003, accepted 26 February 2003

enhancing and -inhibiting amino acid motifs to predict potential proteasome cleavage sites (9). The second algorithm, which uses the results of the first algorithm as an input, predicts which fragments are most likely to be generated. This algorithm is based on a kinetic model of the 20S proteasome (10) and it takes the time-dependent degradation into account.

NetChop is a neural network-based method trained on MHC class I ligands generated by the human proteasomes. Every MHC ligand has to be generated by the proteasome, therefore the rationale behind using MHC class I ligands is that these ligands bear the closest resemblance to naturally processed *in vivo* cleavage products. However, as some of the products of the proteasome would not bind MHC molecules, MHC class I ligands represent only a subset of *in vivo* cleavage products. The MHC class I ligands used to develop NetChop were compiled from public databases (11). There are two versions of NetChop available, 1.0 and 2.0. The later version is trained with a data set that is 3 times larger.

The aim of this study is to compare the performance of the three publicly available methods mentioned above. Since there is increasing evidence that antigenic peptides result from proteasome cleavage especially at the C-terminal end [see, e.g. (12–15)], we test all the methods on a set of publicly available MHC Class I ligands. We are concerned primarily with the ability of the methods (i) to predict correctly the C-terminal of a ligand and (ii) not to predict *major* cleavage sites within the ligand. We excluded N-terminal cleavage analysis, because the majority of the T cell epitopes are trimmed at their N-terminal by other peptidases, e.g. in the endoplasmic reticulum (15).

We find that the method developed using MHC class I ligands, i.e. NetChop, predicts CTL epitope boundaries more accurately than the methods based on *in vitro* degradation data.

Methods

Performance measurement

We require that a proteasome cleavage prediction method should be able to identify the C-terminal of any natural MHC class I ligand without predicting major cleavage sites within the ligand. Thus, for each ligand we test whether (i) the proteasome cleavage prediction methods can predict the C-terminal cleavage correctly and (ii) the same methods do not predict a cleavage site within the epitope (i.e. all positions except the C-terminal residue) which is more likely than at the C-terminal.

The predictions originate from scores that are compared with a threshold and they are classified as follows:

True positive (TP): if the prediction at the C-terminal, P_c , is above the threshold.

False negative (FN): if P_c is less than the threshold.

True negative (TN): if no cleavages are predicted within the epitope (excluding the C-terminal residue) or if the predicted cleavage sites within the epitope are less likely than at the C-terminal (i.e. less than P_c and the threshold).

False positive (FP): if there is at least one predicted cleavage site within the epitope which is more likely than at the C-terminal (i.e. higher than P_c).

We use the following performance measures to compare NetChop, PAPRoC and MAPPP:

Sensitivity = $TP/(TP + FN)$

Specificity = $TN/(TN + FP)$

$$CC = \frac{TP \times TN - FN \times FP}{\sqrt{(TN + FN)(TN + FP)(TP + FN)(TP + FP)}}$$

The sensitivity gives the percentage of C-terminal cleavages that are predicted correctly and the specificity gives the percentage of epitopes with no major predicted cleavage sites (i.e. cleavage sites that are more likely than at the C-terminal) within the epitope. The correlation score, CC, is a measure of how well a method performs *both* in positive (i.e. true cleavage sites) and in negative (i.e. true non-cleavage sites) examples.

Results

Organization of test data set

We focus on the prediction of the specificity of the human proteasome, and therefore we use only peptides associated with HLA-A and HLA-B molecules from the SYFPEITHI database (16) to test various methods. In October 2001 there were 977 unique ligands associated with 120 different HLA-A and HLA-B molecules in the SYFPEITHI database. These ligands are either known T cell epitopes or are naturally processed peptides eluted from MHC molecules. We discarded ligands <8 or >12 amino acids. We also excluded ligands that had already been used for developing NetChop 1.0 or 2.0. The source protein for each ligand was searched for in the SWISSPROT database (17). When an epitope was found in several homologous proteins, homologous proteins were aligned and the most representative protein was chosen unless some additional information about the source protein could be deduced from the original paper. Only epitopes originating from human proteins or from possible human pathogens were included in the data set. The resulting set of 402 peptides contained homologous ligands. In order to prevent possible biases in the analysis, the homologous ligands were excluded using the FASTA (18) and Hobohm-1 algorithms (19). The final set used in our analysis consisted of 249 unique ligands from 135 proteins. The process is described in Fig. 1. The list of ligands is given in Appendix A. Excluding overlapping epitopes, we tested each method on 231 ligands.

Comparison of the methods predicting cleavage by the human proteasome

We use three performance measures to compare the publicly available methods for predicting proteasome cleavage. The formal definitions of these measures are given in Methods. Since there is accumulating evidence that the C-termini of MHC ligands are cleaved precisely by the proteasome, each method should be able to predict the C-terminal of HLA ligands as possible cleavage sites. The sensitivity measure gives the percentage of cleavage sites predicted at the C-terminal of 231 MHC ligands. Note that while all three methods can predict proteasome cleavage sites, only FragPredict can predict fragments generated by the proteasome. In order to

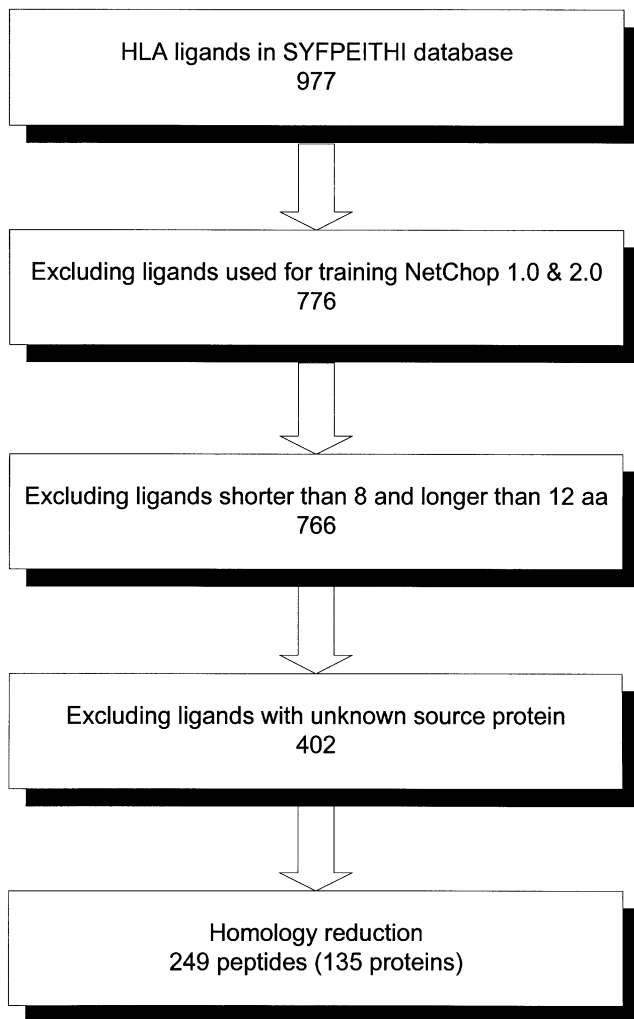


Fig. 1. Diagram summarizing the compilation of the data set used in this study.

be able to compare the FragPredict method with the other two methods, we use only the prediction of cleavage sites from FragPredict. For FragPredict and NetChop, which produce the probability scores of cleavage for each position in a protein sequence, we used a threshold of 0.5 to classify the predictions, i.e. any position in the sequence with a predicted probability >0.5 is considered as a predicted cleavage site. PProC does not allow the use of a threshold value for predictions; we assume that the sites with corresponding '+++', '++' and '+' values produced by this method are predicted cleavage sites. The performance measures of the methods for this data set are given in Table 1. FragPredict is able to predict most of the C-termini as cleavage sites, followed by NetChop 2.0. In contrast, PProC and NetChop 1.0 predict much fewer of the MHC ligand C-termini residues as cleavage sites.

An effective prediction method should also be capable of identifying non-cleavage sites (i.e. sites that are not likely to be used by the proteasomes). When the MHC ligands are used as a test set for proteasome cleavage predictions, it is hard to

Table 1. The performance of three publicly available methods for the prediction of proteasomal cleavage sites deduced from natural human MHC class I ligands

Method	<i>N</i>	Sensitivity	Specificity	CC
PProC	217	45.6	30.0	-0.25
FragPredict	231	83.5	16.5	0.00
NetChop 1.0	231	39.8	46.3	-0.14
NetChop 2.0	231	73.6	42.4	0.16

N corresponds to the number of natural MHC ligands tested. PProC requires a flanking region (six positions to the left and four positions to the right of a cleavage site); 14 of the ligands are found at the beginning, or end, of their source protein and could therefore not be analyzed by PProC. For each ligand, the C-terminal residue should be predicted as a cleavage site. Sensitivity shows the percentage of correct predictions out of *N* true cleavage sites. Specificity shows the percentage of *N* MHC ligands that are predicted as not containing any major cleavage sites. A threshold value of 0.5 was used to classify cleavage and non-cleavage sites. The definitions of the measures are given in Methods. Sensitivity and specificity are in percentages.

define which sites are really non-cleavage sites. Many CTL epitopes contain minor cleavage sites [see, e.g. (20,21)]. Nevertheless, an epitope should not contain a major cleavage site, i.e. a cleavage site that is more likely than the cleavage site at the C-terminal. Therefore, one can assume that if a method does not predict any major cleavage sites within an epitope, it is able to classify non-cleavage sites correctly. In other words, an incorrect prediction of a non-cleavage site (i.e. a false positive) is one where at least one internal position within an epitope has a probability of cleavage higher than both the threshold *and* the probability of the cleavage at the C-terminal. Following this definition, the total number of true non-cleavage sites becomes the same as the number of epitopes. The specificity measure in Table 1 gives the percentage of the MHC ligands with no major predicted cleavage sites within the ligand. NetChop 1.0 is the most successful method in classifying non-cleavage sites, followed by NetChop 2.0 and PProC. FragPredict predicts many major cleavage sites within ligands that would make them highly unlikely MHC ligands. The performance of this method does not change much when we use the full FragPredict package (i.e. including the fragment prediction method): 11% of MHC ligands are predicted to stay intact during the protein degradation (using the suggested value of $P > 0.9$). There are other ways of measuring the performance on non-cleavage sites and we have tried many of them, e.g. one can assume that each position within a ligand should have a cleavage probability lower than the threshold. In all cases, the ordering of the methods according to their success in classifying non-cleavage sites correctly did not change (results not shown).

The correlation coefficient (CC) is a measure of how well a method performs *both* on positive (i.e. true cleavage sites) and negative (i.e. true non-cleavage sites) examples. $CC = 0$ corresponds to random prediction and $CC = 1.0$ represents 100% correct prediction. A negative CC value means that the predictions are not correlated with the real values. Only NetChop 2.0 has a positive CC (see Table 1). This suggests that NetChop 2.0 generates the most reliable predictions.

Table 2. The performance of three publicly available methods for the prediction of proteasomal cleavage sites identified by *in vitro* degradation studies

Method	Sensitivity	Specificity	CC
PAPProC	46.4	64.7	0.10
FragPredict	72.1	41.4	0.12
NetChop 1.0	34.4	91.4	0.31
NetChop 2.0	57.4	76.4	0.32

A threshold of 0.5 was used for FragPredict and NetChop to classify cleavage and non-cleavage sites

Different threshold values can be used in FragPredict and NetChop to classify positions as predicted cleavage sites or predicted non-cleavage sites. When a low threshold is used the methods predict more cleavage sites (and *vice versa* for a high threshold). We investigate the performance measurements of both methods at the standard threshold of 0.5 and at the threshold when the methods reach a maximum correlation coefficient. However, varying the threshold did not change the ranking of the methods according to their performance (results not shown).

The better performance of NetChop may be due to the fact that it was trained using MHC ligands. MHC ligand data reflect not only proteasome specificity, but they also reflect a combined specificity of the proteasome, TAP and MHC. Thus, it cannot be ruled out that NetChop captures this combined specificity and thus performs best when the C-termini of MHC ligands are used for proteasome cleavage predictions. To see if this is the case we also tested all three methods on *in vitro* degradation data generated by the human proteasome. We collected such data from the literature (see Appendix B) excluding the data used to develop PAPProC and FragPredict. The results shown in Table 2 confirm that NetChop is able to capture the specificity of the proteasome better than the other methods.

Conclusion

We found that NetChop, an artificial neural network trained with MHC class I ligands, predicts the C-terminal of CTL epitopes more reliably. This is mainly because NetChop can predict the non-cleavage sites better than any of the other methods (see Table 1). There are two possible explanations for this. First, artificial neural networks are much more non-linear than the other two methods. Thus they might capture the complex specificity of the proteasome better. Second, both PAPProC and FragPredict are based on very limited set of *in vitro* degradation data, whereas NetChop is trained on a larger data set, i.e. with MHC class I ligands.

The C-termini of MHC ligands represent only a subset of cleavage sites occurring during *in vivo* degradation because not all cleavages would result in protein fragments that can be transferred to the endoplasmic reticulum and can bind to an MHC class I molecule. Thus, the use of MHC ligands to develop a method that can predict proteasome cleavage has been the subject of much criticism (H. Margalit, pers. commun.). However, here we demonstrate that the C-termini

of MHC ligands might even represent the specificity of the *in vivo* degradation better than the *in vitro* cleavage maps. Degradation data derived from *in vitro* experiments probably overestimate *in vivo* degradation, because the methods based on this type of data, e.g. FragPredict, predict that most of the MHC ligands in our data set will be destroyed due to major cleavage sites within the ligands.

Even the best method could predict only 73% of the C-termini of natural MHC class I ligands correctly. Moreover, only 42% of the natural MHC ligands are predicted to remain intact. The stochastic nature of degradation (22) and the differences between the immunoproteasome and the constitutive proteasome are just two of many reasons that can explain the poor performance. The use of quantitative data, i.e. concerning not only the cleavage sites used, but also how often a certain site is used, improves the prediction results significantly (C. Kesmir *et al.*, unpublished). Thus, it should be possible to improve on current prediction methods when more quantitative data become available.

In a separate study we found that NetChop 2.0 can correctly discriminate the C-termini of natural MHC ligands from the rest of the protein (results not shown). Thus, NetChop can discriminate the regions that are most likely to be presented to T cells across a protein. This creates a promising future perspective to identify the immunogenic regions in the pathogenic and the human genomes.

Acknowledgements

We thank Lars Juhl Jensen for advice on the statistical evaluation of our results. We are grateful to S. M. McNab for linguistic advice.

Abbreviations

CTL	cytotoxic T lymphocyte
TAP	transporter associated with antigen processing

References

- 1 Rock, K. L. and Goldberg, A. L. 1999. Degradation of MHC class I-presented peptides. *Annu. Rev. Immunol.* 17:739.
- 2 Rammensee, H. G., Falk, K. and Rotzschke, O. 1993. Peptides naturally presented by MHC class I molecules. *Annu. Rev. Immunol.* 11:213.
- 3 Uebel, S. and Tampe, R. 1999. Specificity of the proteasome and the TAP transporter. *Curr. Opin. Immunol.* 11:203.
- 4 Tanaka, K. and Kasahara, M. 1998. The MHC class I ligand-generating system: roles of immunoproteasomes and the interferon- γ -inducible proteasome activator PA28. *Immunol. Rev.* 163:161.
- 5 Van den Eynde, B. J. and Morel, S. 2001. Differential processing of class-I-restricted epitopes by the standard proteasome and the immunoproteasome. *Curr. Opin. Immunol.* 13:147.
- 6 Toes, R. E., Nussbaum, A. K., Degermann, S., Schirle, M., Emmerich, N. P., Kraft, M., Laplace, C., Zwiderman, A., Dick, T. P., Muller, J., Schonfisch, B., Schmid, C., Fehling, H. J., Stevanovic, S., Rammensee, H. G. and Schild, H. 2001. Discrete cleavage motifs of constitutive and immunoproteasomes revealed by quantitative analysis of cleavage products. *J. Exp. Med.* 194:1.
- 7 Nussbaum, A. K., Kuttler, C., Haderl, K. P., Rammensee, H. G. and Schild, H. 2001. PAPProC: a prediction algorithm for proteasomal cleavages available on the WWW. *Immunogenetics* 53:87.
- 8 Kuttler, C., Nussbaum, A. K., Dick, T. P., Rammensee, H. G.,

- Schild, H. and Haderl, K. P. 2000. An algorithm for the prediction of proteasomal cleavages. *J. Mol. Biol.* 298:417.
- 9 Holzhtter, H. G., Frommel, C. and Kloetzel, P. M. 1999. A theoretical approach towards the identification of cleavage-determining amino acid motifs of the 20 S proteasome. *J. Mol. Biol.* 286:1251.
- 10 Holzhtter, H. G. and Kloetzel, P. M. 2000. A kinetic model of vertebrate 20S proteasome accounting for the generation of major proteolytic fragments from oligomeric peptide substrates. *Biophys. J.* 79:1196.
- 11 Kesmir, C., Nussbaum, A. K., Schild, H. and Brunak, S. 2002. Prediction of proteasome cleavage motifs by neural networks. *Protein Eng.* 15:287.
- 12 Craiu, A., Akopian, T., Goldberg, A. and Rock, K. L. 1997. Two distinct proteolytic processes in the generation of a major histocompatibility complex class I-presented peptide. *Proc. Natl Acad. Sci. USA* 94:10850.
- 13 Stoltze, L., Dick, T. P., Deeg, M., Pommerl, B., Rammensee, H. G. and Schild, H. 1998. Generation of the vesicular stomatitis virus nucleoprotein cytotoxic T lymphocyte epitope requires proteasome-dependent and -independent proteolytic activities. *Eur. J. Immunol.* 28:4029.
- 14 Paz, P., Brouwenstijn, N., Perry, R. and Shastri, N. 1999. Discrete proteolytic intermediates in the MHC class I antigen processing pathway and MHC I-dependent peptide trimming in the ER. *Immunity* 11:241.
- 15 Mo, X. Y., Cascio, P., Lemerise, K., Goldberg, A. L. and Rock, K. 1999. Distinct proteolytic processes generate the C and N termini of MHC class I-binding peptides. *J. Immunol.* 163:5851.
- 16 Rammensee, H. G., Bachmann, J., Emmerich, N. N., Bachor, O. A. and Stevanovic, S. 1999. SYFPEITHI: database for MHC ligands and peptide motifs. *Immunogenetics* 50:213.
- 17 Bairoch, A. and Apweiler, R. 2000. The SWISS-PROT protein sequence database and its supplement TrEMBL in 2000. *Nucleic Acids Res.* 28:45.
- 18 Pearson, W. R. and Lipman, D. J. 1988. Improved tools for biological sequence comparison. *Proc. Natl Acad. Sci. USA* 85:2444.
- 19 Hobohm, U., Scharf, M., Schneider, R. and Sander, C. 1992. Selection of representative protein data sets. *Protein Sci.* 1:409.
- 20 Lucchiari-Hartz, M., Van Endert, P. M., Lauvau, G., Maier, R., Meyerhans, A., Mann, D., Eichmann, K. and Niedermann, G. 2000. Cytotoxic T lymphocyte epitopes of HIV-1 Nef: generation of multiple definitive major histocompatibility complex class I ligands by proteasomes. *J. Exp. Med.* 191:239.
- 21 Morel, S., Levy, F., Burlet-Schiltz, O., Brasseur, F., Probst-Kepper, M., Peitrequin, A. L., Monsarrat, B., Van Velthoven, R., Cerottini, J. C., Boon, T., Gairin, J. E. and Van den Eynde, B. J. 2000. Processing of some antigens by the standard proteasome but not by the immunoproteasome results in poor presentation by dendritic cells. *Immunity* 12:107.
- 22 Nussbaum, A. K., Dick, T. P., Keilholz, W., Schirle, M., Stevanovic, S., Dietz, K., Heinemeyer, W., Groll, M., Wolf, D. H., Huber, R., Rammensee, H. G. and Schild, H. 1998. Cleavage motifs of the yeast 20S proteasome beta subunits deduced from digests of enolase 1. *Proc. Natl Acad. Sci. USA* 95:12504.
- 23 Ayyoub, M., Stevanovic, S., Sahin, U., Guillaume, P., Servis, C., Rimoldi, D., Valmori, D., Romero, P., Cerottini, J. C., Rammensee, H. G., Pfreundschuh, M., Speiser, D. and Levy F. 2002. Proteasome-assisted identification of an SSX-2-derived epitope recognized by tumor-reactive CTL infiltrating metastatic melanoma. *J. Immunol.* 168:1717.

Appendix A

Table 3. The list of peptides (including the flanking regions) used in our study

QVPLRPMTYKAAVDLSHFLKEKGGLEGLIHSQ	NEF_HV1PV	73	WQKLETFWAKHMWNFISGIQYLAGLSTLP	POLG_HCV1	1754
PAATLEEMMTACQGVGGPHKARVLAEAMSQ	GAG_HV1BR	338	TKILEPFRSQHPDIVIYQYMDLTVGSD	POL_HV1U4	319
EAIRFIGRAMADRGLLRDIKAKTAYEKIL	VNUC_INBAA	253	ATPPGSVTVPHPNIEEVALSTTGEIPFYG	POLG_HCV1	1349
LLGLMLICSAAENLWVTVYGGVPVWKDATT	ENV_HV1S3	20	RPPPGRRPFHPVGEADYFEYHQEGGPDGEP	EBN1_EBV	397
VLEWRFSRLAFHHVARELHPEYFKNC	NEF_HV1BR	180	EFIWMCMTVRHRCAIRKPKLPIVKQRRW	EBN4_EBV	139
MELAALCRWGLLLALLPPGAAS	ERB2_HUMAN	1	IKGGRHLIFCHSKKKCDELAAKLVALGIN	POLG_HCV1	1385
THTVPIYEGYALPHAILRLDLGRDLTDY	ACTB_HUMAN	160	TLIGANASFSIALNFPGSQKVLDPGQVIWV	PM17_HUMAN	76
TAPPAHGVTSAPDRPAPGSTAPPAHV	MUC1_HUMAN	131	GYIKGIVKDIHDPGRGAPLAKVVFDPYR	RL8_HUMAN	39
TALLKIEGVYARDETEFVLGKRCAYVYKA	R35A_HUMAN	25	EAFSKNLKLGIHEDSTNRRRLSELLRYHTSQ	HS9B_HUMAN	430
QAPSNRVMIPTATIGTAMYKLLKKSRRVAY	BRL1_EBV	124	FLLSLRGAGAIAKADHVSTYAAFVQTHRPT	HA2Q_HUMAN	22
VFDNKFHIGAVGIGIIVVMIFGMIFSMI	CD9_HUMAN	187	SVGLGKVLIDILAGYGAGVAGALVAFKIM	POLG_HCV1	1841
LVVSVFVGGGLAVILPPLSPYFKYSVMINKATP	NI9M_HUMAN	19	PAAEHRLREELAKFLHWMMSVYVVELLR	TERT_HUMAN	530
LAAGWPMGYQAYSSWMSYSDTHTPTTFV	EBN3_EBV	34	TRVESENKVVILDSFDPLVAEEDEREISV	POLG_HCV1	2242
YGGISLLSEFCRVLCCYVLEETSVMIAKR	VIE1_HCMVA	299	WDVLKGSRVSILFGHENRVSTLVRVSPDGT	GBB5_HUMAN	310
GGIGRFYIQMCTELKLSYDEGLIQNSLT	VNUC_IAPUE	34	RPILSPLTKGILGFVFTLTVPSERGLQRRR	VMT1_IAPUE	49
CGIAVGTITVDADKYAVTVETRLIDERAA	OM1E_CHLTR	357	PVGEIYKRWIILGNLKVIRMYSPSILDIRQ	GAG_HV1BR	256
IGKMRYSVSRDFKGVKVLIDIREYWMDE	P15_HUMAN	65	FQNLQVIRGRILHNGAYSLTLQGLGISWL	ERB2_HUMAN	425
EELFDLHARDHCVAHKLFNNLK	UCRH_HUMAN	69	SSIVYEAADAILHTPGCVPCVREGNASR	POLG_HCV1	210
ATLCSALYVGDLCGVSFVLVQGLFTFSPPR	POLG_HCV1	269	AELLAENREILKEPVHGVYDDPSKDLIAE	POLG_HV1BR	466
DVDNASLARLDLERKVESLQEEIAFLKKL	VIME_HUMAN	208	DLTFLARSALILRGVSAHKSCLPACVYGP	VNUC_IAPUE	255
VIDTLTCGFADLMGYIPLVGPALGGAARA	POLG_HCV1	122	GPLCIRMDQAIMDKNILKANFSVIFDRLE	VNS1_IAPUE	113
SALSEGATPQDLNTMLNTVGGHQAAMQML	GAG_HV1BR	172	LRGTALTEVIPLTEEAELAELENREILK	HA2Q_HUMAN	438
YLEYRQVPDSDPARVEFLWGPRAEAESY	MAG1_HUMAN	248	VQACRAIRHIPPRIROGLERILL	ENV_HV1BR	838
IVKNIDDGTSDDRPYSHALVAGIDRYPRK	RL27_HUMAN	24	GVVAGGGVALIRAASAITAAGLKGDNDQ	CH60_YEREN	410
RDYFEYEGKIDITIEHTDRQSGKRRGFGF	ROA2_HUMAN	129	GVDIRHNKDRKVRKPEKKSQDI	RL18_HUMAN	1
KEALLDTGADDTVLEEMNLPGKWPKMIG	POL_HV1RH	75	KISKGANPVEIRRGVMLAVDAVIAELKKQ	CH60_HUMAN	130
YARKRSHTNDVKQLTEVYVQKVSITESVI	POL_HV1U4	508	AIGCVRNKQIVDCLTEMYMGTAITTCE	FAYF_HUMAN	1511
IKKQLGSLVSDYCNVNLKEFTAGSVEITLR	BRL1_EBV	18	LHPDKWTVQPIVLPKEDSWTVNDIQKLVG	POL_HV1BR	401
LVKTGTITTFEHAHNMRVMKFSVSPVVRV	EF2_HUMAN	479	KQGGQWQTYQIQEPFKNLKTGYARTRGA	POL_HV1BR	498
ATLYCVHQRIEIKDTKEALDKIEEQNKNS	GAG_HV1BR	82	LNAWVVKVVEEKAFSPEVIMFMSALSEGATPQ	GAG_HV1BR	151
AMKAYINKVEELKKKYGI	ACBP_HUMAN	69	IGVYAGYGTVYKARDPHSGHFVALKSVRVNG	CDK4_HUMAN	12
RGRERFEMFRELNEALELKAQAGKEPGG	P53_HUMAN	333	IPYWDWRDAEKCDICTDEYMGQHQHTPNPN	TYRO_HUMAN	233
LVKLVYQLEKEPIVGAETFYVDGAASRETK	POL_HV1BR	589	ANIQEFAGCKKIFGSLAFLPESFDGDPAS	ERB2_HUMAN	359
AQONNVHEKLVKLVADARVVKCKLGDVNFEE	RS7_HUMAN	161	CGHEALTGTTEKLIETYFSKNYQDYEYLN	MYPR_HUMAN	34
QLEKEPIVGAETFYVDGAANRETKLKGAGYV	POL_HV1A2	583	LWDQSLKPCVKLTPLCVTLNCTNVNGTAV	ENV_HV1MA	110
GHQAAAMQMLKETINEEAEWDRVHPVHAGP	GAG_HV1BR	192	QVRIKPGSANKPKDELVDYENDIEKKICK	CSP_PLAFA	358
VCMFLASKLKETSPLTAEKLCIYTDNSIKP	CGD2_HUMAN	104	VAAGMNPMDLKRGDKAVIAAVEELKKLS	CH60_YEREN	107
PSLRILYMTDEVNDPSLTIKSIGHQWYWTY	COX2_HUMAN	79	KGKGDKAQIEKRIQEIIEQLDVTTSEYEK	CH60_HUMAN	359
FIMESGAKGCEVVYSGKLRQAKSMKVFV	RS3_HUMAN	125	EDQKIGIEIHKRTLKIPAMTIAKNAGVEG	CH60_HUMAN	459
EGQELSDDEDEVYQVTYVQAGRSDDTDF	MDM2_HUMAN	264	FSVPLDEDFRKYTAFTIPSINNTEPGRIYQ	POL_HV1BR	283
KTTDGYLLRFLCVGFTKKRNNQIRKTSYA	RS3A_HUMAN	127	GKRTEQKQEVLEKARGSTYGTTPRPVVKP	EBN3_EBV	264
SLDKLKEVKEFLGENISNFLSAGNTYQLT	APL1_HUMAN	232	KYAMQLEIITLIVIGILISVILYFIFCR	E311_ADE03	20
AIKWEYVVLFLLLADARVVKLCSLWMMLLI	POLG_HCV1	713	NLPGCSFSIFLLALLSCLTVPASAHQVRNS	POLG_HCVH8	52
GPLLVLQAGFFLLTRILTIPOSLDSWWT	VMSA_HPBVW	173	SYLKGSSGGPLLCPAGHAVGIFRAAVCTR	POLG_HCV1	1159
MPAWGALFLLWATAEATKDCPSPTCT	GPIX_HUMAN	1	SAHPFGFGQSLFLFGYVYVFGDCVQGDWC	TAT_HTLIA	6
EFGATVELLSFLPSDFFPVSRDLDTVSAL	CORA_HPBVJ	8	TPGTQSPFFLLLLLTVLTVTGSGHAST	MUC1_HUMAN	2
SGWGSIEPEEFLTPKKLQCVDLHVISNDVC	KLK3_HUMAN	155	EVCNDQVDLYLLMDCSGSIRRHNVVNHAV	TRAP_PLAFA	41
VPNSDPPIYQFLWGPRAAYETTKMKVLEF	MGB1_HUMAN	260	LNLTMFLMLLWTLVLLICSSCSSCPL	LMP2_EBV	319
QYRNFWLKFFRLLKSKLEDNIRLRAL	APL1_HUMAN	119	YKCVDRDLKVLMIIPLNVTFISSDREV	VGLH_EBV	532
LDIRQGPKEPFRDYVDRFYKTLRAEQASQE	GAG_HV1BR	282	RDGNNEDEKLRKPKHKKLQKPDGPNPDP	CSP_PLAFA	95
RAFTEEGAVGEISPLPSLPGHTDEDVKN	VNS1_IAMAN	148	QFLSLQCLQALYVDSLFFLRGRDLQLLRH	MAPE_HUMAN	291
HHNLLVCSVSGFYPPGSIIEVFRNQGEEK	HB2F_HUMAN	140	MLLSVPLGLLGLLAVAEPA	CRTC_HUMAN	1
TVLIKSLRSGHDPRAQGT	HA2Q_HUMAN	241	MMRKLAILSVSSFLVEALF	CSP_PLAFA	1
KNTMMRKAIRGHLENNPALEKLLPHIRGN	RLA0_HUMAN	57	NNQGNGQLGHMNPNDPNRNVDENANANNV	CSP_PLAFA	290
DLNTMLNTVGGHQAAAMQMLKETINEEAAE	GAG_HV1BR	182	MVDGTLLLSSEALALTQT	HLAE_HUMAN	1
PHHERCSDSDGLAPPQHLIRVEGNLRVEYLD	P53_HUMAN	177	RHMQDAEMFTNAACMALNIWDRFDVFC	OM1E_CHLTR	111
DARMQAIQNAAGLCTLVAMLEETIFWLQEI	IE63_EBV	249	ATMEELQREINAHGQLVIARQKVRDAEK	NCAP_HANTV	2
IKARAACRAAGLQDCTMLVCGDDLVVCE	POLG_HCV1	2717	NWMTETLLVQNANPDKCTILKALGPAATL	GAG_HV1BR	314
NSASILPEMEGLSEFTEYLSSEVVPSPF	MAPB_HUMAN	215	QPGYWPPLYGNEGLGWAGWLVSPRGRPN	POLG_HCVTW	78
LLVPFVQWFVGLSPTVWLSVIWMMWYWGSP	VMSA_HPBVJ	338	VGTLEEIIDDHNAIVSTSVGSEHYVSLS	PRS4_HUMAN	109
AMPHLLVSSGLSRVYARLSSNRINHQ	DPOL_HPBVJ	443	ALINVSANCPNHFEHGYQYKSPVEDNHK	DUS1_HUMAN	202
VPVKLPGMDGPKVKQWPLTEEKIKALVE	POL_HV1BR	175	DKTVALWDLRNLKLLHTFESHKDEIFQV	RBB7_HUMAN	294
VGGVYLLPRRGPRLGVRATRKTSTERSQPR	POLG_HCVH	31	PPWQAGILARNLVPVMTVQGGQNLKYQEFF	PP65_HCMVA	485
PPVLQIQVMGQGGSPATAAASAVQAPT	EBN4_EBV	821	TKILEPFRKQNPDIVIYQYMDLTVGSD	POL_HV1BR	332
RPQDVKFPGGQIVGGVYLLPRRGPRLG	POLG_HCVH	18	VLKIITFTKNNQFQALLQYADPVSAQHAK	PTB_HUMAN	210
NRFGMDKIYEGQVEVTEGDEYNVESIDGQPG	RL5_HUMAN	110	IWGKTPKFKLPIQKETWETWWTEYWFATWI	POL_HV1BR	549
IGYSEKDRFQGRDVEYKVS	ATNB_HUMAN	283	TSSSPQKPKLDEGYFTLQIRGRERFEM	P53_HUMAN	312
DAVKVTLGPKGRNVLDKFSGPTITKDG	CH60_YEREN	25	QEIQGWMTSNPPIPVGDIYKRWIILGLNK	GAG_HV1MA	249
LIVTRIVELLGRRGWALKYWNNLLQYWSQ	ENV_HV1BR	731	KMPATSRPTAPPSSKGGNYVYVQIGSNY	GAG_SIVSP	117
IFHKDLCAQAGVALQTMKQEFNLNVKQK	FETA_HUMAN	582	MKQQAAGIGILLALTAICWGAL	YHBE_ECOLI	1
SAGATVGMIGVLVGVALLI	CEA5_HUMAN	684	VHFKNTRETAQAIKGMHIRKATKYLKDVT	RL17_HUMAN	24
EETVEELGVMGVYDGRHETVYGEPRKLLTQ	MAG4_HUMAN	220	VQNIQQLMVAHQALSPTLNWVVKVVEEKAFS	GAG_HV1BR	134
YYAMLAKTGVHHSYGNEHLETACGKYRVR	RL30_HUMAN	61	DRFYKTLRHEQASQSEVKNWMTETLLVQNA	GAG_HV1N5	297
MNHLGNVKYLVIVLFIFFDLF	TRAP_PLAFA	1	ATRDGKLPATQLRRHIDLTVGSATLSALY	POLG_HCV1	247
TLPALSTGLIHLHQNIVDVQYLVGVGSSI	POLG_HCV1	681	QAAADTGHSSQVSNYPIVQNIQGMVHQ	GAG_HV1BR	116

SCHAASNPPAQYSWFVNGTFQOQSTQELFIP	CEA5_HUMAN	258	AYRPPNAPILSTLPETTVVRRRGRSPRRRT	CORA_HPBVJ	131
YKNRVASRKCRAKFKQLLQHYREVA AAK	BZLF_EBV	180	TSVPAAPPASTNRQSGRQPTPLSPLRD	VMSA_HPBVJ	75
GISIKLQEEEEERRDNYVPEVSALDQEI	RS17_HUMAN	67	KTCPVQLWVDSTPPPGTRVRAMAIYKQSQ	P53_HUMAN	139
NNTRKSIQRQGRGPAFVITIGKIGNMRQAH	ENV_HV1BR	306	NGKRLPEPNWASVKKDLISYGGGWRLSAQW	POLG_DEN3	1534
SAPLPHTTERIETRSARHPWRIRFGAPQ	POLG_HCV1	254	PKMFAKGTETIHAVVIKLLNEILQARGKK	IF38_HUMAN	315
EGSDTITLPCRQIKQIINMWQVKGKAMYAP	ENV_HV1H2	409	QLQAQHLSHATHGPPVQLPPHPSGLQPP	TL3_HUMAN	127
LRSLCLFSYHRLRDLIVTRIVELLGRRGW	ENV_HV1BR	765	HHCKLTQVLNTHYVAPRLLLTGTPLQNK	SN24_HUMAN	889
GGELDRWEKIRLPPGGKKYKLVKHIWASR	GAG_HV1BR	9	YPYRLWHYPTINITYTIFKIRMYVGGVEH	POLG_HCV1	611
MHGRLVTLKDIVLDLQPPDPVG	VE7_HPVI1	1	VPLAHSSSAFTITDQVPPFSVSVQLRALDG	PM17_HUMAN	198
PPSQASSGQARMFPNAPYLPSCLESQPAI	WT1_HUMAN	116	ALEGFDDKADGTLDSQVMSLHNLVHSLNG	TYR2_HUMAN	350
VSTVQCTHGIRPIVSTQLLNGSLAEEV	ENV_HV1A2	245	FQPLHTVMRETLFIGSHVVLRELRLNVT	VGLH_EBV	410
SGCPERLASCRPLTDFDQGWGPISYANGSG	POLG_HCV1	450	GCLLDRKAAGTTPAGGGFPRRHVTLPSK	TISB_HUMAN	33
LNQSVENCTRPNNNRKSIQRQGRGPAF	ENV_HV1BR	293	PGFQALSEGCTPYDINQMLNCGVDHQAAM	GAG_HV2BE	172
QEEEEVGFVPRQVPLRPMTYKAALDISHFL	NEF_HV1A2	65	QEILDWYHTQGYFPDQWNYTPGPGIRYPLT	NEF_HV1A2	111
QKIETAFLMARRARSLSAERYTLFFDLVSSG	EBN4_EBV	233	EPRGSDIAGTTSTLQEQIGWMTNPPPIPVG	GAG_HV1BR	229
ELEVECATQLRRFGDKLNFRQKLLNLISK	APR_HUMAN	20	KNQVAMNPTNTVFDKRLIGRRFFDAVQSD	HS7C_HUMAN	56
GSDSPTLDNSRRLPIFSRLSISDD	TISB_HUMAN	315	PAGLKKKSVTVLDVGDAYFSVPLDEDFR	POL_HV1BR	264
GGLEGLIHSQRRQDILDWYHTQGYFPDW	NEF_HV1PV	95	IAKITPNNNGTYACFVSNLATGRNNSIVK	CEA5_HUMAN	642
QNPVPGNIYRRWIQLGLQKCVRMYNPTNI	GAG_HV2D2	250	PVSPGDQLPGVFSGDGRVACAPVPAPAGPI	EBN3_EBV	349
RLVFPDLGVRVCEKMALYDVVTKLPLAV	POLG_HCV1	2578	PGRGEPRIAVGYVDDTQFVRFDSDAASQ	1A01_HUMAN	39
MRVKEKYQHLWRWGWRWGTM	ENV_HV1H2	1	INEEAAEWDRVHPVHAGPIAPGMREPRGS	GAG_HV1BR	204
MRVMAPRALLLLSGGLALT	1C11_HUMAN	1	LHGMDPEREVLWRVFSRLAFHHVARELH	NEF_HV1BR	170
QLQARILAVERYLKDQQLGIWGCSGKLI	ENV_HV1BR	580	TMVAGAVWLTVMNSNTLLSAWILTAGFLIFL	LMP2_EBV	432
VGNIVQSCNPRYSIFFDYMAIHRSLTKI	EBN3_EBV	104	TIIDDDRVRQEVPKLKVICALRVTSRARSRI	RL18_HUMAN	84
VDDLRAIAEESDEEIAIAYTLATAGVSSSDS	VIE1_HCMVA	368	VLDVGDAYFSVPLDKDFRKYTAFTIPSINN	POL_HV1A2	263
HYREVA AAKSSENDRLRLLLKQMCPSLDV	BZLF_EBV	199	FPSTAQAQAAVQGPVGTDFKPLNSTPATT	Z207_HUMAN	286
SGGDPEIVTHSFNCGGEFFYCNSTQLFNS	ENV_HV1H2	365	EQTRSKAGLLVSDGGPNLYNIRNLHIPEV	RRP1_IAPUE	581
PGYAGMLGNSSHIPQSSYCSLHPHERLS	ITF2_HUMAN	236	TEARDLHCLLVNPHTDWKSGLVEVAS	G45B_HUMAN	112
EKVTWTEAAGSIRDGVRAAYTALHYLSHLS	QORL_HUMAN	115	KKKYKLVKHIWASRELERFVAVNPGLET	GAG_HV1BR	25
SGSGTYCLNVSLADTNSLAVVSTQLMPGQ	PM17_HUMAN	560	GKWSKSSVIGWPTVRRMRRAEPAADGV	NEF_HV1W	3
TVKTNVPMNSLDQSVVELYTDTFASWSV	OM1E_CHLTR	167	LAAMLRLQAQYHAKDPNNLFMVRLAQLGT	PSD2_HUMAN	741
SSTQASLEIDSLFEGIDFYTSITRARFEEL	HS71_HUMAN	276	TPPLITDYREYHTDTTVKVVKMTEEKLA	TP2A_HUMAN	950
SAGHTVSGFVSLAPGAKQNVQLINTNGSWH	POLG_HCV1	391	LGFLQRTDLSYIKSFVSDALGTTSIQTPW	EBN3_EBV	148
LSISSCLQQLSLLMWITQCFLPVFLAQPPSG	CTG1_HUMAN	147	FPVFSKASEYLQVFGIEVVEVVPISHLY	MAG2_HUMAN	147
STLPGNPAIASLMAFTAAVTSPITTSQTL	POLG_HCV1	1779	PQPICTIDVYIMVVKCWIMIDSECRPRFRE	ERB2_HUMAN	942
SLTSAQSGDYSLVIVTTFFVHYANFNHYFV	VGLH_EBV	215	DVGAGVIDEDYRGNVGVVLFNFGKEKFEV	DUT_HUMAN	183
WGVLAGIAYFSMVGWAKVLVLLLFAGV	POLG_HCV1	353	FLTSLIDRYTPTISRERAVELLRKCLE	PSB2_HUMAN	137
RCALGVFRKFSRFEALRLALMLNDMELV	PSD2_HUMAN	250	KKFIRHQSDRVYKIKRNWRKPRGIDNRV	RL32_HUMAN	17
TMESSTLELRSRVWAIRTRSGGNTNQRA	VNUC_IAPUE	373	DPASRELVSYVNVNMGLKIRQLLWFHIS	CORA_HPBVO	78
AYLTLAKHTISSDYVIPIGTYGQMKNGSTPM	TYRO_HUMAN	136	MSWRGRSTYYWPRRRYVQPPMIGPM	GGE4_HUMAN	1
PAHLLQDDISSYTTTTITAPPVGLQN	ACOD_HUMAN	2			

The peptides are shown in boldface. The SWISSPROT accession number of the proteins and the start position follow the sequence.

Appendix B

Table 4. Samples of peptide degradation by the human constitutive proteasome *in vitro*

Cleavage map	Reference
D↓WQND↓Y↓TPGPGVVR↓V↓PL↓TF↓GW↓CY↓K↓L↓V↓PVEPDK	20
TGSTAV↓PYGSF↓KH↓V↓DT↓RLQ	21
MNGD↓DAF↓ARR↓PTV↓G↓A↓QIPEIK↓K↓A↓FD↓DIAKYFSKEEWEKMK↓SEKIFYV↓Y↓M↓KRKYEAMT↓K↓GF↓K↓A↓T L↓PPFM↓CN↓KRA↓EDFQGNL↓DNDPNRGNQVER↓PQM↓T↓F↓G↓RL↓QGISP↓K↓MPKPAEEGNDSEEVPEAS↓GPQND G↓KEL↓CPPGKPTTEKIHE↓R↓SGPKRGEHAW↓TH↓RL↓RE↓R↓KQ↓L↓VIY↓E↓EISDPEEDDE	23

Data have been collected from literature to test the performance of three publicly available methods for the prediction of proteasomal cleavage sites; an arrow represents the observed cleavage site.