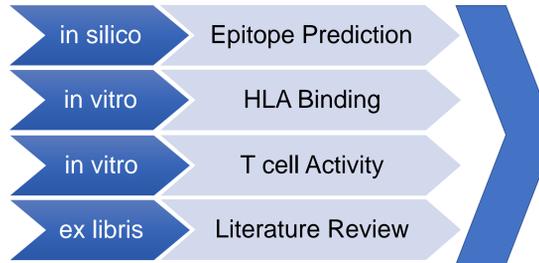


## The PANDA Platform

- In 2017, the FDA released a draft guidance that requires generic peptide drug manufacturers to provide evidence that synthesis-related impurities found in their drug substance do not increase the immunogenicity of the drug product.
- Peptide drugs can be associated with impurities that result from changes in the sequences due to failures in the manufacturing process leading to deletions, insertions, integration of incorrect amino-acids, side-chain modifications and other modifications.
- We have used both immunoinformatics-driven analysis and in vitro validation assays to perform immunogenicity risk assessment of peptide generics. This combination of in silico and in vitro tools is referred to as the PANDA process can be used to support generic peptide drug equivalency in an ANDA application.

## PANDA: Immunogenicity Risk Assessment for Synthetic Peptide Abbreviated New Drug Application Using Computational and Analytical Methods



**Statement of Immunogenicity**  
 EpiVax's expert opinion on the T cell mediated immune response to RLD vs. synthetic peptide generic

ANDAs for Certain Highly Purified Synthetic Peptide Drug Products That Refer to Listed Drugs of rDNA Origin  
 Guidance for Industry  
**DRAFT GUIDANCE**  
 This guidance document is being distributed for comment purposes only.  
 Comments and suggestions regarding this draft document should be submitted within 60 days of publication in the Federal Register of the notice announcing the availability of the draft guidance. Submit electronic comments to <https://www.regulations.gov>. Submit written comments to the Division of Dockets Management (HFA-305), Food and Drug Administration, 5630 Fishers Lane, Room 106, Rockville, MD 20852. All comments should be identified with the document number listed in the notice of availability that publishes in the Federal Register.  
 For questions regarding this draft document, contact CDER, Xiaohui Jiang at 240-402-7964.

## In Silico Evaluation of Immunogenicity

### Searching for T Cell Epitopes with EpiMatrix

EpiMatrix predicts T cell epitopes  
 HLA binding is a prerequisite for immunogenicity

EpiVax tests for binding potential to the most common HLA molecules within each of the "supertypes"<sup>3</sup> shown above.  
 These are representative of >95% of human populations worldwide.<sup>2</sup>

#### EpiMatrix Detail Report: Teriparatide RLD

Frame	AA	Start	Stop	Hydrophobicity	Z-Score	DRB1*0101	DRB1*0301	DRB1*0401	DRB1*0701	DRB1*0901	DRB1*1101	DRB1*1301	DRB1*1501	Hits
1	SVSEQLQAH	9	23	0.29	-0.28	0.28	0.27	0.16	0.39	0.21	0.37	0	0	
2	VSEIQLMHL	10	-0.01	-0.37	-0.41	-0.04	-0.65	0.22	0.10	0.82	-0.99	1.11	0	
3	SEIQLMHL	11	-0.06	-0.02	-0.24	-0.41	-0.14	-1.10	-0.83	-0.60	0.52	-0.67	0	
4	EIQLMHLG	12	-0.01	1.00	0.83	1.15	0.28	1.77	0.72	1.78	0.27	1.31	2	
5	QLMHLGK	13	-0.06	2.47	1.71	2.88	1.67	2.01	1.62	2.89	1.69	2.42	8	
6	LMHLGK	14	-0.91	-1.16	-0.46	-0.44	0.20	0.37	0.12	0.01	-0.02	-0.29	0	
7	AMHLGK	15	-0.1	2.27	1.06	1.26	2.17	1.17	1.44	1.18	1.26	1.41	2	
8	MHLGK	16	-0.91	1.41	1.26	0.84	0.64	1.84	0.95	1.93	1.49	1.21	2	
9	HLGK	17	-1.21	0.38	1.07	1.11	-0.04	0.55	-0.10	1.17	0.75	1.45	0	
10	LKHLGK	18	-0.64	-0.85	0.93	-1.12	0.03	0.21	0.35	0.28	0.59	-0.24	0	
11	LKHLGK	19	-0.64	0.06	0.67	0.66	1.00	0.71	0.12	-0.32	2.08	0.30	1	
12	GKHLGK	20	-1.57	1.00	0.78	1.05	0.33	1.38	0.36	1.02	0.06	1.30	0	
13	KHLGK	21	-1.06	0.29	0.34	0.16	0.47	-0.05	0.00	0.25	-0.12	-0.34	0	
14	HLGK	22	-1.01	-1.07	0.26	-1.12	-0.23	-0.12	0.26	-0.13	-0.83	-1.38	0	
15	LSDHVVV	23	-0.76	1.38	1.33	0.20	1.54	0.91	0.80	1.09	1.16	0.91	0	
16	NSMVEVVK	24	-0.76	0.35	-0.03	0.31	0.41	-1.17	-0.73	-0.61	-0.70	-1.75	0	
17	SMVEVVK	25	-0.87	-1.07	-0.90	-2.16	-0.92	-0.79	-1.56	-0.55	-0.38	-0.58	0	
18	MVEVVK	26	-1.21	0.00	0.13	0.68	0.90	-0.03	-0.43	0.71	0.49	1.27	0	
19	SEVVK	27	-1.86	-0.55	-0.29	-0.25	-1.04	-0.77	-0.95	-0.55	-0.96	-1.27	0	
20	VVVK	28	-1.04	-0.05	0.10	-0.47	0.98	-0.22	-0.05	0.23	1.30	0.67	0	
21	SMVEVVK	29	-0.93	1.23	1.09	0.96	0.86	2.34	0.23	2.51	1.51	1.38	2	
22	EVVK	30	-1.79	-0.64	-0.68	-1.47	-0.92	1.47	-0.88	0.09	0.54	-0.07	0	
23	LVVK	31	-0.93	0.71	1.03	0.16	1.65	2.04	0.88	1.42	0.27	0.48	2	
24	LVVK	32	-1.19	0.19	0.39	-0.25	-0.14	1.05	0.40	0.61	0.32	-1.21	0	
25	RKLVVK	33	-2	0.29	-0.02	0.82	-0.04	0.62	-0.44	-0.07	0.20	1.15	0	
26	KLVVK	34	-1.19	0.18	-0.46	0.64	0.00	-0.13	0.10	0.35	1.20	-1.35	0	

Summarized Results: Maximum Single Z-score: 2.88, Maximum Z-score: 2.88, Sum of Significant Z-scores: 4.74, Count of Significant Z-scores: 2, Scores Adjusted for Tregitope: --, Hydrophobicity: -0.67, EpiMatrix Score: 16.03, EpiMatrix Score (w/o flanks): 16.03, EpiMatrix Score (w/o flanks): 16.03

\*7 frames contain putative T cell epitopes (Z-scores > 1.64, medium and dark blue shading)  
 EpiBar = promiscuous binding motif  
 Teriparatide has a total of 19 EpiMatrix Hits  
 High EpiMatrix Score (scores above 10 indicate significant immunogenic potential)

### Searching for Human-like epitopes with JanusMatrix

The EpiBar in frame 5 has a high JanusMatrix Human Homology Score suggesting it is a potential regulatory epitope (Tregitope) and will be tolerated or even actively tolerogenic\*  
 \*confirmed with unpublished in vitro studies

### Immunogenicity Quadrant Plot

Observed Teriparatide and Salmon Calcitonin Impurities

Salmon Calcitonin and its impurities fall into the Epitope Sparse, Non-human quadrant. Observed immunogenicity to SCT can be attributed to foreign epitopes within the sequence.  
 Teriparatide and many impurities fall into the Epitope Dense, Human-Like quadrant. Two observed impurities create non-human epitopes and are predicted to be immunogenic.

## Evaluating Risk of (nearly) all possible peptide-related impurities with the What-if-Machine (WhIM)

### The "What-if Machine" (WhIM) is a computer algorithm that:

- Mimics the process of synthesizing polypeptide drug products;
- Records all possible product impurities created through known failures in the synthesis process<sup>4</sup>;
- Scores each potential impurity for T cell epitope content (EpiMatrix) and human cross-reactive potential (JanusMatrix);
- Weights each impurity based on an assumed probability of occurrence;
- Summarizes the scores of all potential impurities in order to calculate an impurity risk profile.

### Risk Profile for Salmon Calcitonin & Teriparatide Impurities

Measure	Salmon Calcitonin	Teriparatide
RLD	1.99	16.03
Max	13.39	46.63
Upper Q	1.06	10.33
Median	0.87	8.38
Lower Q	-4.90	7.54
Min	-18.42	-9.26
Weighted Avg	0.32	10.07
Total	1648	1853
High Scoring (Count)	9	488
High Scoring (percent)	0.55%	26.34%

### Highest-Scoring WhIM-generated Impurities: Teriparatide

Peptide Sequence	IMPURITY_TYPE	PROBABILITY	EMX SCORE	JMX SCORE	WEIGHT DELTA
SVSEIQLMHLGKHLNSMERVEVWLRKLLQDVHNF	RLD	n/a	16.03	4.74	N/A
SVSEIQLMHLNLYEINSMERVEVWLRKLLQDVHNF	DELETION	1.0E-04	46.63	1.19	1.39%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	DUPLICATE_AA	5.0E-05	29.72	2.29	-2.76%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_AA	2.5E-06	29.72	2.29	-2.76%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_AA	5.0E-06	29.72	2.29	-2.76%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_AA	2.5E-06	29.72	2.29	-2.76%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_AA	2.5E-06	24.21	1.09	-2.78%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_AA	5.0E-06	25.13	2.09	-2.12%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_AA	5.0E-06	25.13	2.09	-2.12%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_AA	2.5E-06	23.39	1.73	-2.76%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_AA	2.5E-06	24.17	0.82	-3.34%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_AA	5.0E-06	22.70	2.00	-2.41%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_AA	5.0E-06	22.46	1.33	-2.41%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_BETA_AA	5.0E-07	21.64	1.86	-4.15%
SVSEIQLMHLNLYGKHLNSMERVEVWLRKLLQDVHNF	INSERTION_AA	5.0E-06	20.82	0.75	-2.76%

## In Vitro Confirmation Assays

### In Vitro Class II HLA Binding Assays

Class II HLA Binding Results for Teriparatide and high-risk WhIM-identified Impurities:

Peptide	DRB1*0101	DRB1*0301	DRB1*0401	DRB1*0701	DRB1*0901	DRB1*1101	DRB1*1301	DRB1*1501
Teriparatide_RLD N-term (1-18)	Negligible	Negligible	Low	Non-Binder	Negligible	High	Moderate	Moderate
Teriparatide_RLD C-term (18-34)	Non-Binder	Non-Binder	Non-Binder	Non-Binder	Non-Binder	High	Moderate	Negligible
WhIM_DES-GLY12_Teriparatide (1-21)	Moderate	Moderate	Moderate	Non-Binder	Non-Binder	High	Moderate	High
WhIM_ENDO-LEU11_Teriparatide (1-21)	Very High	Negligible	High	Negligible	Non-Binder	High	Moderate	Very High

### In Vitro T cell Assays – In Vitro Immunization Protocol (IVIP)

IVIP Results: Teriparatide RLD and Two WhIM-Identified High-Risk Impurities

Peptide	Percentage of Responding Donors
Teriparatide	n/a
WhIM_DES-GLY12_Teriparatide	10%
WhIM_ENDO-LEU11_Teriparatide	10%

## Conclusions

- It is important to assess the potential immunogenicity of not only peptide drug candidates, but also their synthesis-related impurities in early stages of drug development.
- In the recent FDA guidance, peptide drug manufacturers must provide proof that synthesis-related impurities do not increase the immunogenicity of the drug substance.
- In silico tools such as EpiMatrix and JanusMatrix can provide a quick and cost-effective method to screen peptides for immunogenicity.
- When impurities are unknown, the What-if-Machine can quickly screen all plausible peptide-related impurity sequences and identify potentially immunogenic impurities.
- Combining these in silico tools with in vitro HLA binding and T cell assays is referred to as the PANDA process can be used to support generic peptide drug equivalency in an ANDA application or in the immunogenicity screening of novel peptide therapeutics.

## References

- ANDAs for Certain Highly Purified Synthetic Peptide Drug Products That Refer to Listed Drugs of rDNA Origin Guidance for Industry; Draft Guidelines issued by Office of Generic Drugs (OGD) in the Center for Drug Evaluation and Research (CDER), Food and Drug Administration Federal Drug Agency. <https://www.fda.gov/oc/ohrt/andas-certain-highly-purified-synthetic-peptide-drug-products-refer-listed-drugs-rdna-origin>.
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