A machine learning model based on tumor and immune biomarkers to predict undetectable measurable residual disease (MRD) in transplant-eligible multiple myeloma (MM)

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Disclosures

- Puig: Amgen, Celgene, Janssen, Takeda, The Binding Site: honoraria; Amgen, Celgene, Janssen, Takeda: consulting or advisory; Celgene: speakers' bureau; Celgene, Janssen, Amgen, Takeda: research funding; Amgen, Celgene, Janssen, Takeda: travel accommodations and expenses.
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- Mateos: Janssen, Celgene, Takeda, Amgen, Adaptive, GSK, Sanofi, Oncopeptides: honoraria, membership on an entity's Board of Directors or advisory committees.
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- The remaining authors declare they have no competing interests.

The (r)evolution of current therapies Unmet need of biology-based individualized treatment



Possible models towards individualized treatment How to select and confirm the success of treatment selection?

	Treatment A				
Diagnosis	Treatment B				
	Treatment C				
	Treatment D				
	Treatment E				

Possible models towards individualized treatment

Confirmation of predicted PFS can only be done **retrospectively**

	Treatment A	Selection based on predicted PFS	PFS	5 years? 10 years?
	Treatment B			V
Diagnosis	Treatment C			
	Treatment D			
	Treatment E			

Possible models towards individualized treatment Predicting undetectable MRD can be confirmed <u>earlier</u>



Possible models towards individualized treatment Predicting undetectable MRD can be confirmed <u>earlier</u>



This idea has not been investigated previously; therefore, we sought to explore this concept and define a machine learning model **to predict undetectable MRD in newly-diagnosed transplant-eligible MM patients**, treated with a standard of care.

Methods. (I) GEM2012MENOS65 & GEM-CESAR trials used in the study

Trials differ by disease stage and treatment



Methods. (I) GEM2012MENOS65 & GEM-CESAR trials used in the study

Trials differ by disease stage and treatment



Methods. (II) Study workflow



Methods. (II) Study workflow



Methods. (II) Study workflow





- Albumin
- B2 microglobulin
- LDH levels
- International staging system (ISS)
- Revised-ISS









GEM2012MENOS65



Univariate analyses to identify variables significantly associated with MRD status





Results. (II) Logistic regression algorithm





Results. (II) Logistic regression algorithm

Variables at diagnosis	-0.6	-0.4	-0.2	0.00	0.2	0.4	_	1	0	Input h (xi)	x	Coefficient weights (w)
Myeloid precursors								> 0.21	≤ 0.21	1 or 0		+ 0.497
Intermediate neutrophils								> 36.33	≤ 36.33	1 or 0		+ 0.490
Mature B cells								> 1.75	≤ 1.75	1 or 0		+ 0.379
Eosinophils								> 1.76	≤ 1.76	1 or 0		+ 0.252
del(17p13) and/or t(4;14)								positive	negative	1 or 0		- 0.351
Plasma cell clonality (BM)										(X - 11.01) / 13.71		- 0.398
Circulating tumor cells (PB)								> 0.735	≤ 0.735	1 or 0		- 0.629
CD27neg CD38pos T cells								> 0.61	≤ 0.61	1 or 0		- 0.656
CD56bright CD27neg NK cells								> 0.04	≤ 0.04	1 or 0		- 0.685
Logistic regression intercept										1		+ 0.237
Immune microenvironment Cytogenetic abnormalities Tumor burden				\mathcal{P}_{k}	oredicte	d (Y =	+1	Xi , Wi) =	$\frac{1}{1+e^{-\Sigma}}$	$wh(x_i)$		$=\sum_{i=1}^{\infty}wh(x_i)$
$\mathcal{P}(y \mid z)$	X i, W i)	<	<u>_</u> 1	Ppredic	ted (Y	= +1	 X i	, Wi) > 0.5		► Ypredicted = U	nde	tectable MRD
			<u> </u>	Ppredic	cted (Y	ı = +1	X	i, Wi) < 0.5		► Ypredicted = p	ersi	stent MRD

Results. (II) Logistic regression algorithm

Interactive webpage to facilitate its use in clinical practice

del(17p13) and/or t(4;14)	negative	
	Ex. positive or negative	
Plasma cell clonality	0.24	
	Ex. 0.24	
Absolute number of circulating tumor cells [PB]	1.45	
	Ex. 1.45	
Myeloid Precursors	163	
	Ex. 1.63	
CD56bright CD27neg NK cells		Des l'aties and to table MDD is sould all second as this is second
	Ex. 0.23	Your prediction is undetectable MRD, with a probability of 70.1% of undetectable MRD
Eosinophils	2.34	
	Ex. 2.34	http://www.MRDpredictor.com
CD27neg CD38pos T cells	0.35	
	Ex. 0.35	
Mature B cells	2.53	
	Ex. 2.53	
Intermediate neutrophils	40.6	
	Ex. 40.6	
	Devident	

Results. **(III) The accuracy of the algorithm** Probability distributions



GEM2012MENOS65



Results. **(III) The accuracy of the algorithm** Probability distributions



Results. (III) The accuracy of the algorithm Receiver operating characteristic curves



Results. **(IV) Prognostic value of the model** PFS and OS based on <u>actual MRD outcomes</u>



Results. (IV) Prognostic value of the model

PFS and OS based on MRD **predicted** outcomes; *standard-confidence* predictions (n = 212)



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Results. (IV) Prognostic value of the model

PFS and OS based on MRD **predicted** outcomes; *high-confidence* predictions (n = 102)



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Conclusions

- We demonstrated that it is possible to predict patients' MRD status with significant accuracy, using an integrative, weighted model based on machine learning algorithms.
- These findings should stimulate other investigators to further validate this model and define new ones in other treatment scenarios.
- Finally, selecting a regimen based on probable MRD outcomes, and confirming soon after if that probability was accurate, is a possible new approach towards individualized treatment in MM.

Thank you!







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