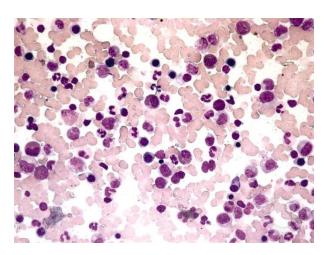
Noninvasive Diagnosis of Myelodysplastic Syndromes (MDS)

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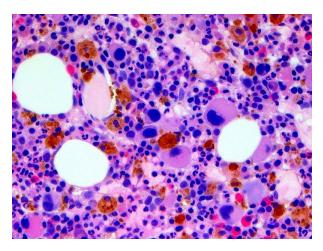
Tel Aviv Medical Center/York University/
Radboud University/EUMDS

Myelodysplastic Syndromes (MDS)

- Clonal hematopoietic stem cell disease
- Abnormal
 - differentiation, maturation, apoptosis
- Anemia common
 - often treated with EPO
- Cytopenias:
 - WBC and Platelets
- Transformation
 - Acute Leukemia (20%-60%)



Normal bone marrow



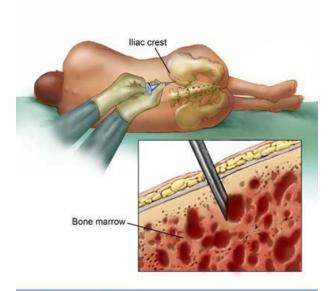
MDS bone marrow

MDS Diagnosis – gold standard

Bone marrow examination

But...

- Invasive
- Painful
- Possible bleeding (thrombocytopenia)
- Difficult for the older patients



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Aim of this study

Can we diagnose MDS noninvasively in at least a portion of the patients?

MDS patients: 48

Control patients: 63

MDS patients: 48

Control patients: 63

$$Y = \frac{e^{(B_1X_1+B_2X_2+...+B_6X_6+C)}}{1+e^{(B_1X_1+B_2X_2+...+B_6X_6+C)}}$$

- Gender (X₁)
- Age (X₂)
- Hb (X₃)
- MCV (X₄)
- WBC (X₅)
- Platelets (X₆)

MDS patients: 48

Control patients: 63

$$Y = \frac{e^{(B_1X_1+B_2X_2+...+B_6X_6+C)}}{1+e^{(B_1X_1+B_2X_2+...+B_6X_6+C)}}$$

- Gender (X₁)

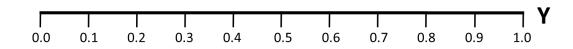
- Age (X₂)

- Hb (X₃)

- MCV (X₄)

- WBC (X₅)

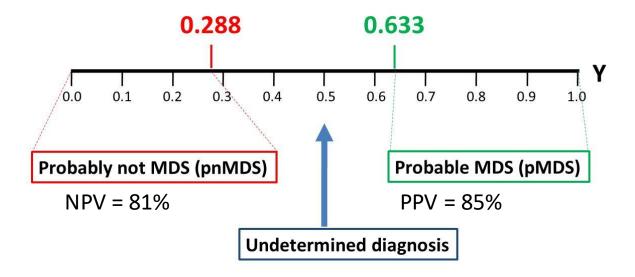
- Platelets (X₆)



Probably not MDS (pnMDS)

Probable MDS (pMDS)

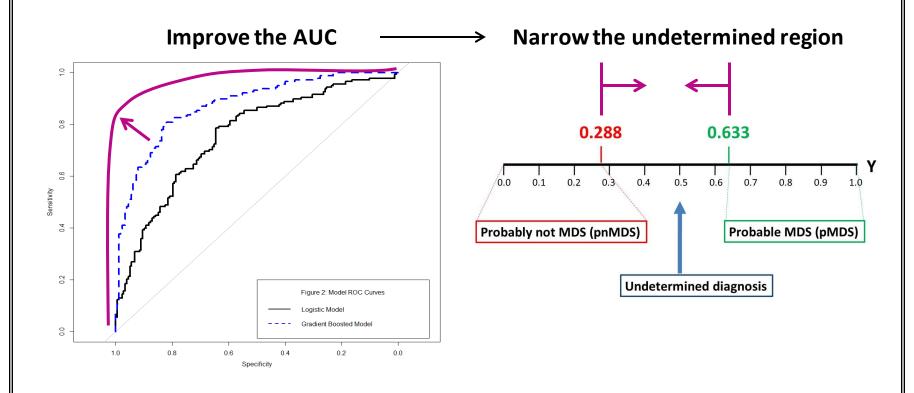
- Area under ROC curve (AUC) = 0.748
- *Two* cutoffs:



Approximately 50% could benefit from noninvasive diagnosis

(Oster et al. Leuk & Lymph 2018: 59; 2222)

Model improvement



More patients could benefit from noninvasive diagnosis of MDS

Improved model (Stages II and III)

- More patients
 - MDS:
 - Israel MDS registry (>260)
 - European MDS registry (>2600)
 - Controls: Ichilov pathology data base
- More variables
- Better model
 - Gradient Boosted Model (GBM)
 - Much more complex than LoR
 - Takes into account the interactions among variables
- Collaboration with York University

Improved model

- Stage II (ASH 2017)
 - Patients:
 - MDS: 178
 - Controls: 178
 - Gradient Boosted Model (GBM)
 - Number of variables: 6 variables
- Stage III (ASH 2018)
 - Patients:
 - MDS: **502**
 - Controls: **502**
 - Gradient Boosted Model (GBM)
 - Number of variables: 10 variables



Age →

Gender ------

 $\mathsf{Hb} \longrightarrow$

WBC ----

 $PLT \longrightarrow$

 $MCV \longrightarrow$

$$Y = \frac{e^{(B_1X_1 + B_2X_2 + \dots + B_6X_6 + C)}}{1 + e^{(B_1X_1 + B_2X_2 + \dots + B_6X_6 + C)}}$$

 X_i (i=1...6) – Each variable B_i (i=1...6) – Relative weights of X_i C - A constant



В

Age ───── Gender ────

 $Hb \longrightarrow$

WBC -----

 $PLT \longrightarrow$

MCV -----

Neutrophils -----

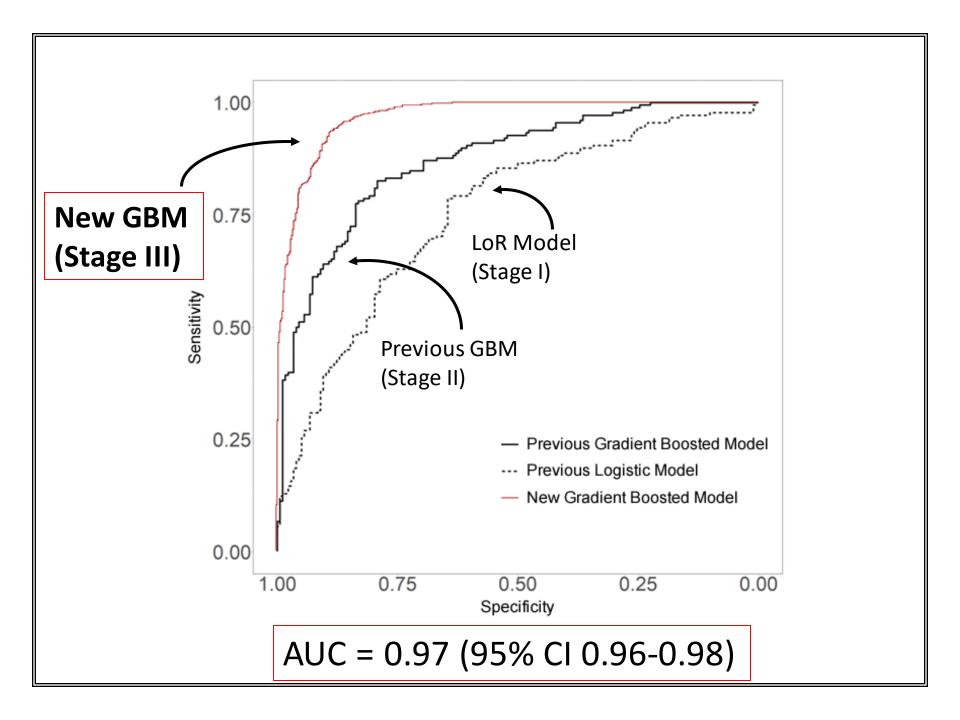
Monocytes ------

Glucose ———

Creatinine -----

GBM algorithm computation

→ GBM Score





MDS Predictive Modelling

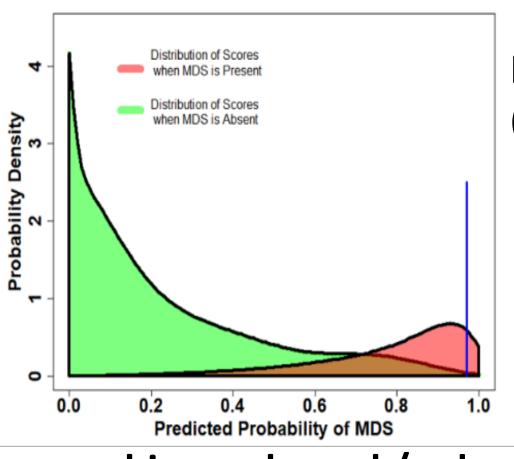
Disclaimer This web application is experimental, and should not be used in the diagnosis of any medical condition. What is the age of the patient? (In years) Sex (M/F) Male 75 O Female Haemaglobin Count? White Blood Count? 10.5 3.1 Mean Corpuscular Volume? Platelet Count? 125 101 Neutrophil Count? Monocyte Count? 0.9 0.2 **Blood Glucose Concentration?** Creatinine? 100 1.1 Calculate



MDS Predictive Modelling

Age = 75Sex = MaleHemoglobin = 10.5WBC = 3.1Platelets = 125MCV = 101Neutrophils = 0.9Monocytes = 0.2Glucose = 100Creatinine = 1.1

Neutrophii Count?		Monocyte Count?	
0.9		0.2	
Blood Glucose Concentration?		Creatinine?	
100		1,1	

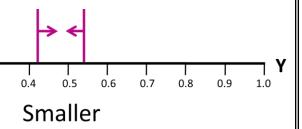


Probable MDS (pMDS)



Model Quality

- One cut-off (AUC = 0.97): improved!
 - Sensitivity = 88%
 - Specificity = 95%
- Two cut-offs
 - 90% PPV (above upper cut-off)
 - 95% NPV (below lower cut-off)
 - Indeterminate region: improved!
 - 14% of the patients
 - (50% in our earlier model)



indeterminate region

MDS Predictive Modelling MDS Predictive Modelling Disclaimer Disclaimer This web application is experimental, and should not be used in the diagnosis of any medical condition. This web application is experimental, and should not be used in the diagnosis of any medical condition. What is the age of the patient? (In years) Sex (M/F) What is the age of the patient? (In years) Sex (M/F) O Male O Male Female Female Haemaglobin Count? Haemaglobin Count? White Blood Count? 12 13 Platelet Count? Mean Corpuscular Volume? Platelet Count? Mean Corpuscular Volume? 220 180 Neutrophil Count? Monocyte Count? Neutrophil Count? Monocyte Count? 0.7 0.5 2.0 **Blood Glucose Concentration?** Creatinine? Blood Glucose Concentration? Creatinine? 1.4 120 100 1.1 Calculate Calculate Probably not MDS (pnMDS) Indeterminate Distribution of Scores Distribution of Scores 4 when MDS is Present when MDS is Present Distribution of Scores Distribution of Scores Probability Density 2 3 when MDS is Absent when MDS is Absent robability Density 0 0.0 0.2 0.6 1.0 0.0 0.4 0.8 1.0 Predicted Probability of MDS Predicted Probability of MDS



Future Work

- Complete/improve (internet, cellphone...) and publish
- Improve the model stage IV
 - Increase numbers of patients and variables
- Incorporate additional peripheral blood parameters
 - Flow cytometry
 - Genetic info
- Use the model for other purposes
 - Prognosis
 - Follow up

Summary and Conclusions

• **Stage 1**:

- 6 variables, (48 and 63 patients), LoR model
- Diagnose/exclude MDS in 50% of patients.

• **Stage 2**:

- GB model, 178 and 178 patients
- Model improvement

Stage 3:

- 10 variables, 502 and 502 patients, GB model diagnosed
- PPV = 90%; NPV = 95%
- Diagnose/exclude MDS in 86% of patients.
- <u>Conclusion</u>: For most patients, MDS can be diagnosed or ruled out noninvasively without a bone marrow examination.

Thank you

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Thank you