

Artificial Intelligence to improve clinical management of MDS

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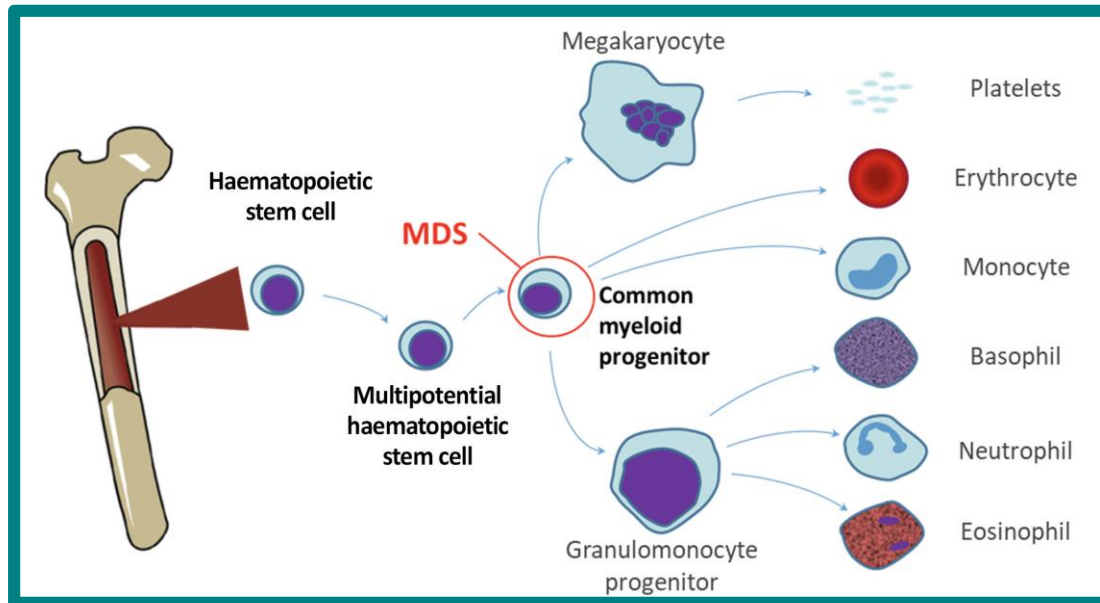
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C.A.L.R.

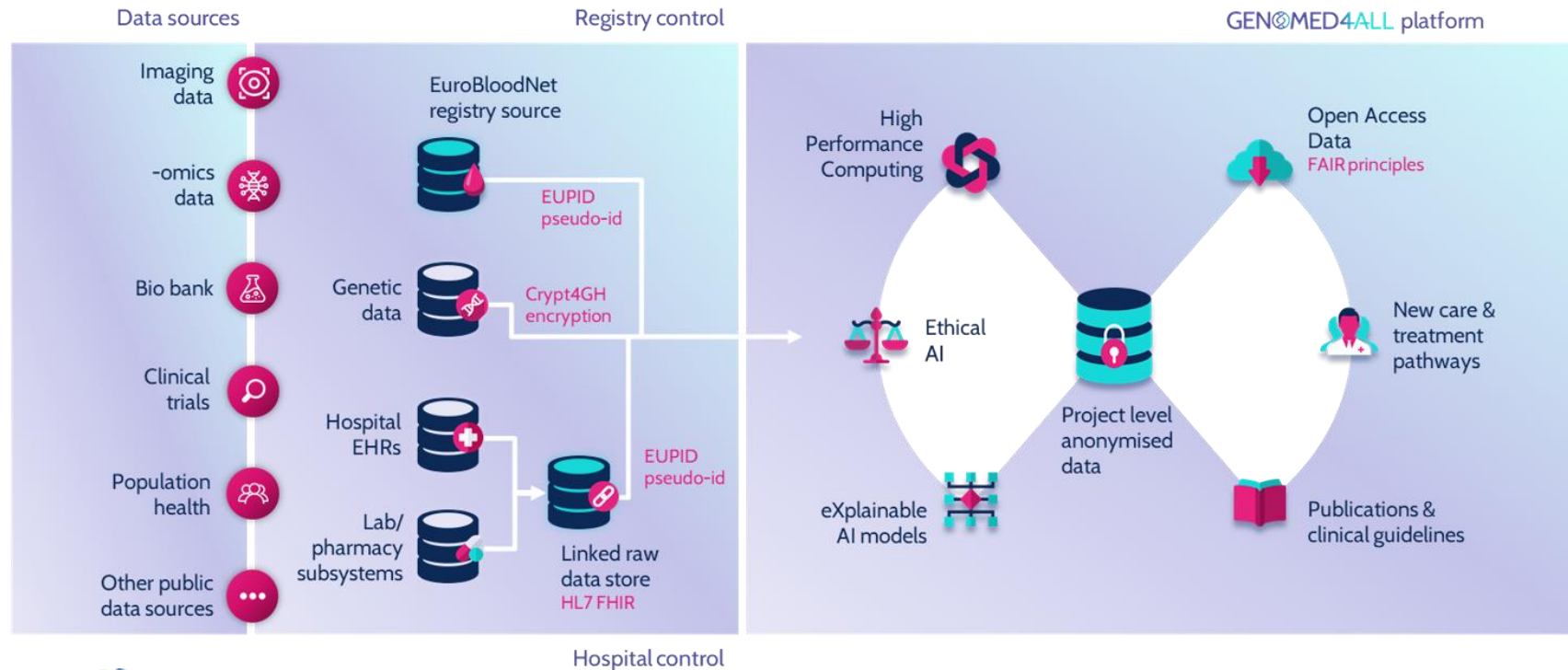


Myelodysplastic Syndromes (MDS)



- Myelodysplastic syndromes (MDS) are **heterogeneous** clonal hematopoietic disorders characterized by peripheral blood cytopenia and increased risk of evolution into acute myeloid leukemia (AML).
- MDS range from indolent conditions to cases rapidly progressing into AML, and therefore a **risk-adapted strategy** is needed

GENOMDE4ALL: Genomics and Personalized Medicine for all through AI in Rare Hematological Diseases



Artificial Intelligence (AI) for precision medicine

1- Machine Learning



2- Generative AI



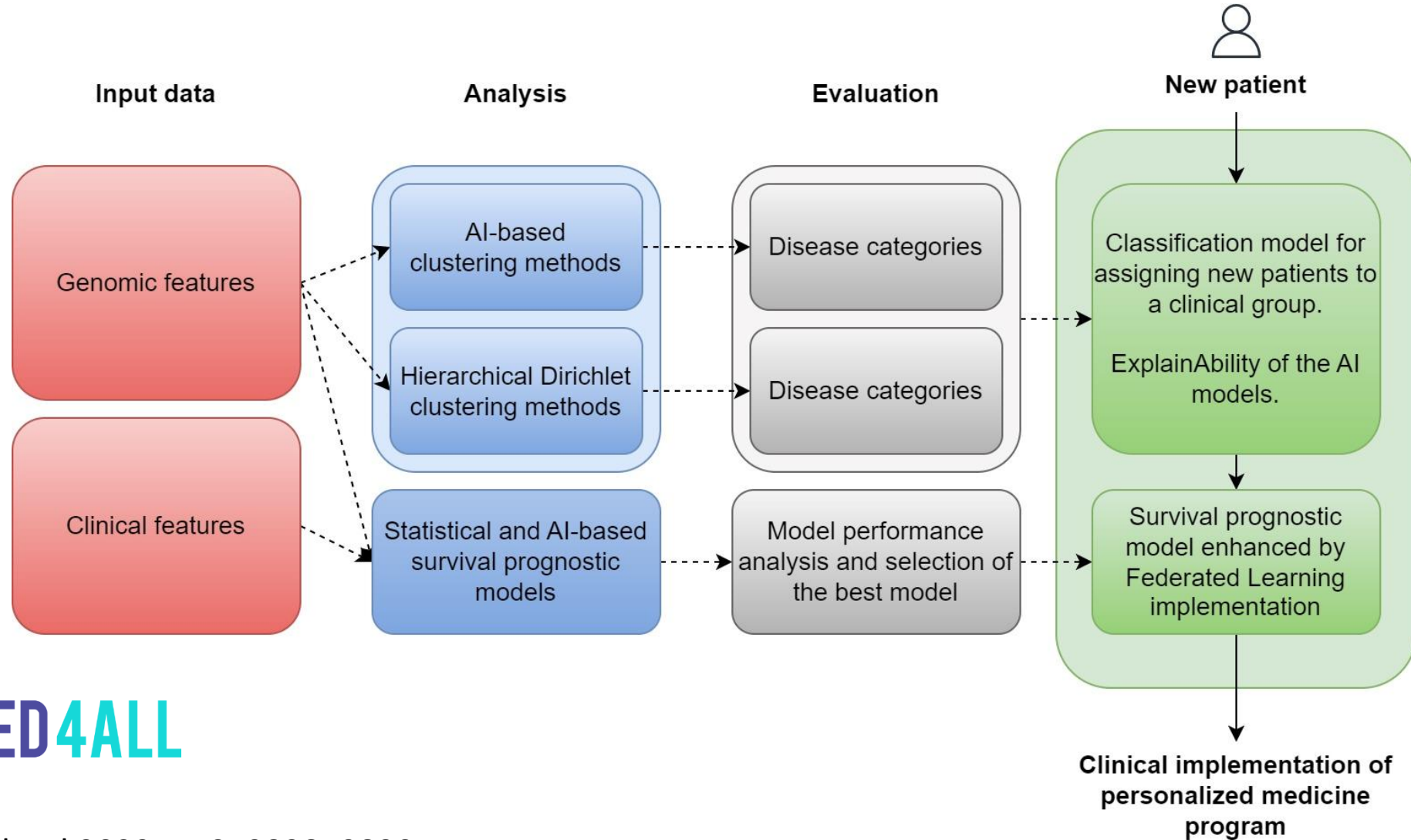
2021 WHO guidance on ethics and governance of AI for health

We have to address three important topics, deemed as essential for a **right deployment of AI in hematology**:

- **Transparency of models.** We have to provide a good understanding of the models (interpretability and explainability)
- **Reliability of models.** The main vulnerabilities of AI models are related to lack of generalizability. Therefore, extensive, independent validation of generated AI-models is required.
- **Protection of data and data sharing.** Innovative technologies such as federated learning procedures for data collection and analysis (without moving sensitive medical data from their original locations) are required to facilitate clinical implementability of AI solutions

1. *The World Health Organization. 2021 WHO guidance on ethics and governance of artificial intelligence for health. <https://www.who.int/publications/i/item/9789240029200>*

MOSAIC - Multi-Modal Analysis and Federated Learning Approach for Classification and Prognostic Assessment in Rare Cancers



2022 WHO/ICC classification of MDS

Table 1. Comparison of MDS subtype definitions in WHO 2016, WHO 2022, and ICC classification of MDS.

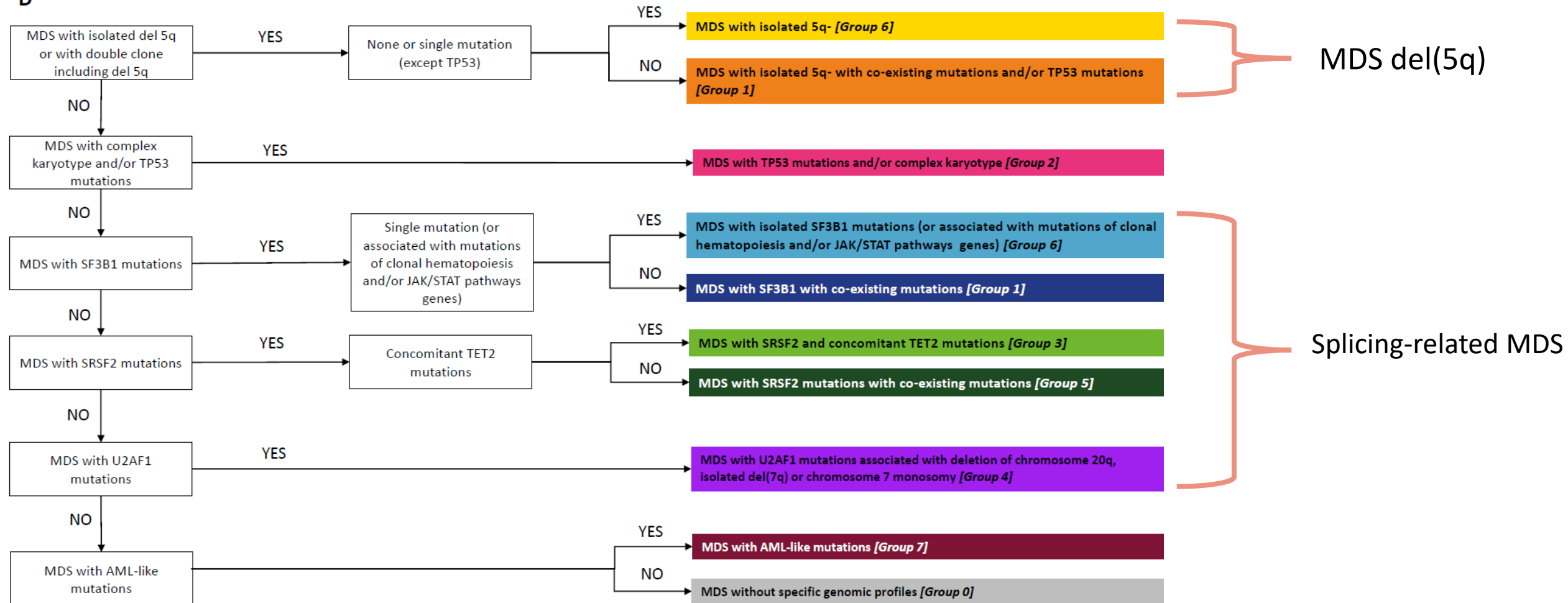
WHO 2016 [1]	WHO 2022 [3]	ICC [5]
MDS with single lineage dysplasia (MDS-SLD)	Not included MDS with low blasts (MDS-LB) < 5% BM and <2% PB	MDS, not otherwise specified with single lineage dysplasia (MDS, NOS-SLD)
MDS with multi-lineage dysplasia (MDS-MLD)	MDS with low blasts (MDS-LB) < 5% BM and <2% PB	MDS, not otherwise specified with multi-lineage dysplasia (MDS, NOS-MLD)
MDS with ring sideroblasts • With single lineage dysplasia (MDS-RS-SLD) • With multi-lineage dysplasia (MDS-RS-MLD)	MDS with low blasts and mutated <i>SF3B1</i> or MDS with ring sideroblasts (if ≥ 15% RS and <i>SF3B1</i> wild-type)	MDS with mutated <i>SF3B1</i>
MDS with isolated del(5q)	MDS with low blasts and isolated 5q deletion (MDS-5q)	MDS with del(5q)
MDS unclassifiable	Not included	Not included
Not included	Not included	MDS, not otherwise specified without dysplasia (e.g., monosomy 7/del(7q)) ^a
MDS excess blasts-1 (MDS-EB1; 5–9% bone marrow blasts)	MDS with increased blasts-1 (MDS-IB1; 5–9% bone marrow and/or 2–4% peripheral blood blasts)	MDS excess blasts (5–9% bone marrow and/or 2–9% peripheral blood blasts)
MDS excess blasts-2 (MDS-EB2; 10–19% bone marrow or peripheral blood blasts or Auer rods)	MDS with increased blasts-2 (MDS-IB2; 10–19% bone marrow or 5–19% peripheral blood blasts or Auer rods)	MDS/AML (10–19% bone marrow or peripheral blood blasts)
AML-defining genetics ^b	AML-defining genetics independent of bone marrow and peripheral blood blast count	AML-defining genetics with ≥10% bone marrow and peripheral blood blasts
AML (≥20% bone marrow and peripheral blood blasts)	AML (≥20% bone marrow and peripheral blood blasts)	AML (≥20% bone marrow and peripheral blood blasts)
Not included	MDS with biallelic <i>TP53</i> inactivation (Two or more <i>TP53</i> mutations, or 1 mutation with evidence of <i>TP53</i> copy number loss or cnLOH)	MDS with mutated <i>TP53</i> (Multi-hit <i>TP53</i> mutation, or <i>TP53</i> mutation (VAF > 10%) and loss of 17p) and MDS/AML with mutated <i>TP53</i> (Any somatic <i>TP53</i> mutation (VAF > 10%))
Not included	MDS, hypoplastic (MDS-h)	Not included
Not included	MDS with fibrosis (MDS-f)	Not included
Not included	Clonal hematopoiesis (CHIP, CCUS) ^c	Pre-malignant clonal cytopenias and CCUS ^c

Open issues:

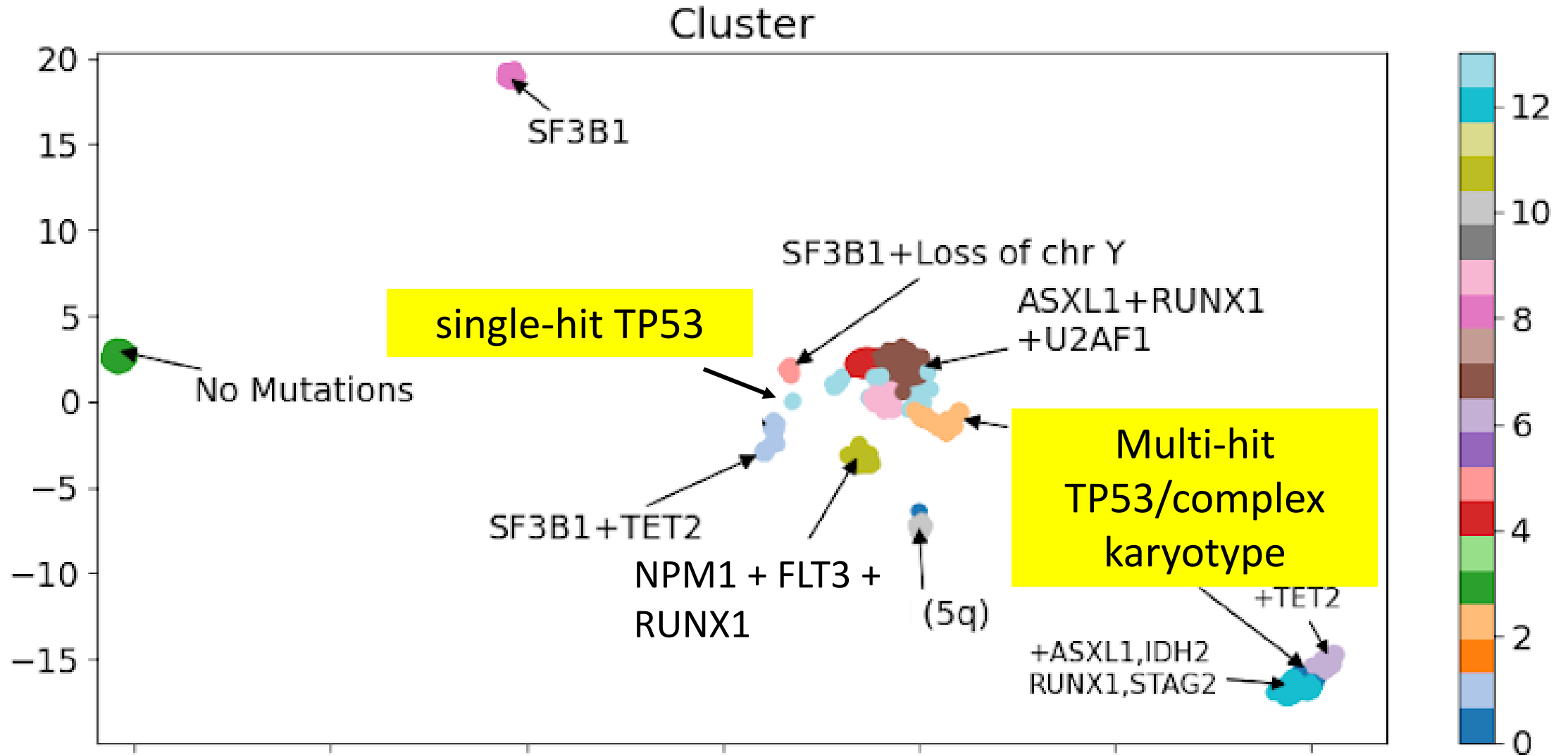
- Increasing the number of clinical entities defined according to biological features
- ICC/WHO harmonization

Molecular classification of MDS by statistical learning methods

D



Molecular classification of MDS by machine learning methods (AI)

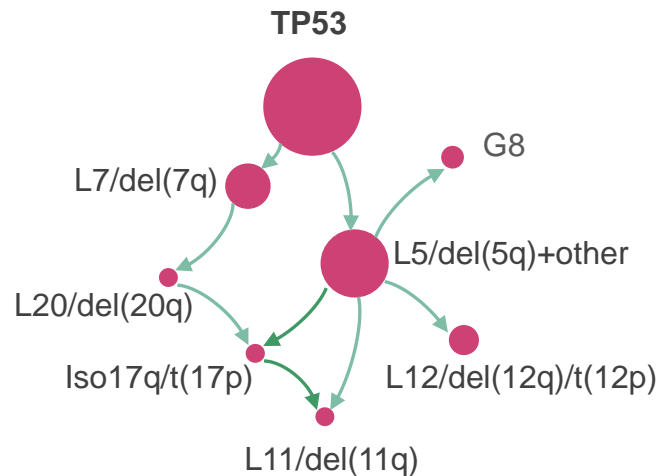


A new category of high-risk MDS is defined according to p53 dysfunction

MDS with TP53 mutations and/or complex karyotype is a distinct disease category

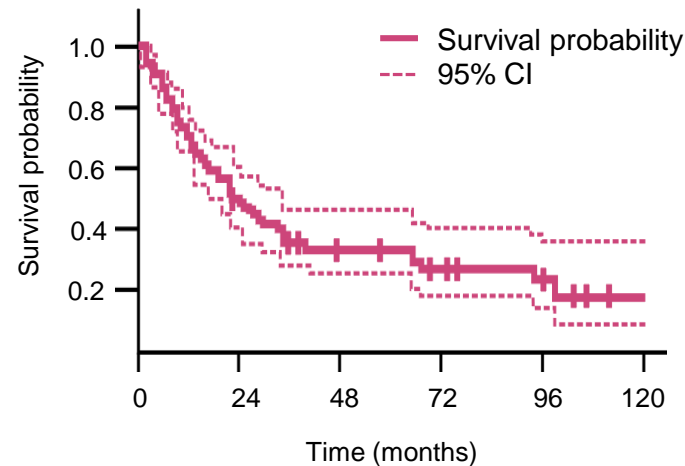
Group 2 (n=188)

MDS with TP53 mutations and/or complex karyotype

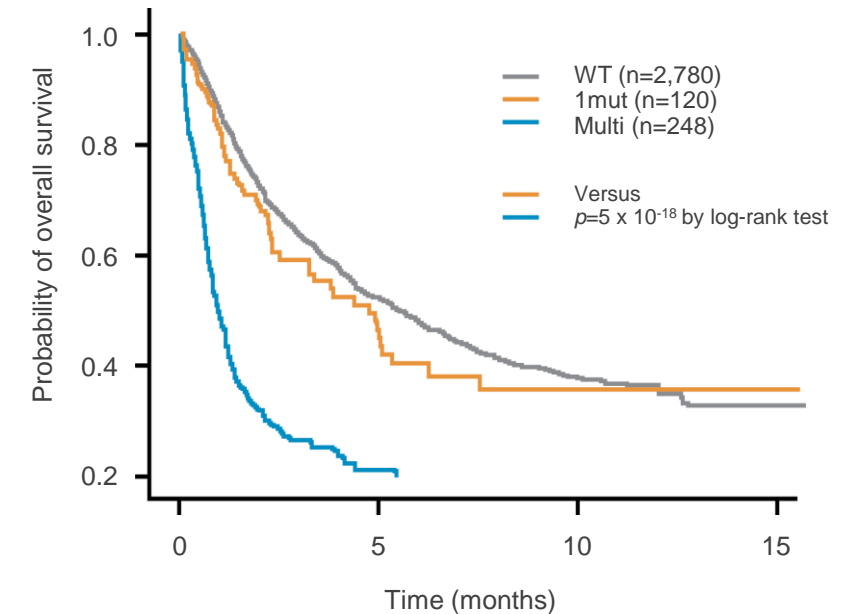


Implications of TP53 allelic state for genome stability, clinical presentation and outcomes in MDS

PB: two or more cytopenias with transfusion-dependency
BM: excess of blasts



J Clin Oncol 2021;39:1223–1233



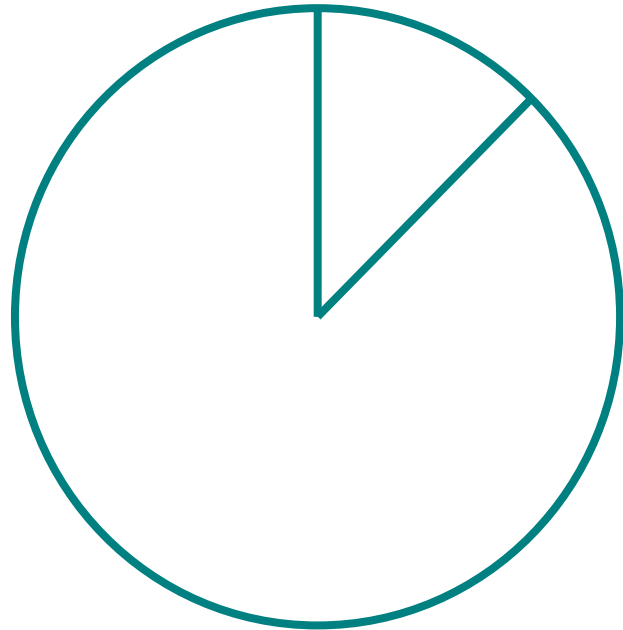
Nat Med 2020;26:1549–1556

A new category of high-risk MDS is defined according to p53 dysfunction

SHAP (Shapley_Additive_Explanations) was used to explain the classification model by computing the contribution of each feature

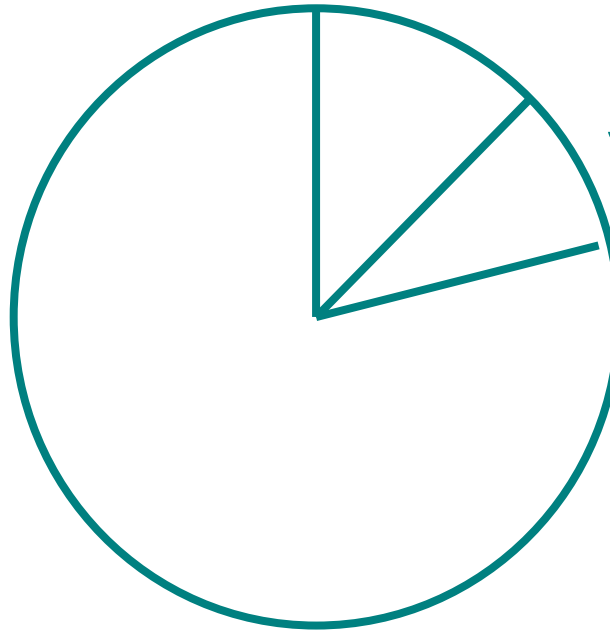
MDS with TP53-mutations

10-12%

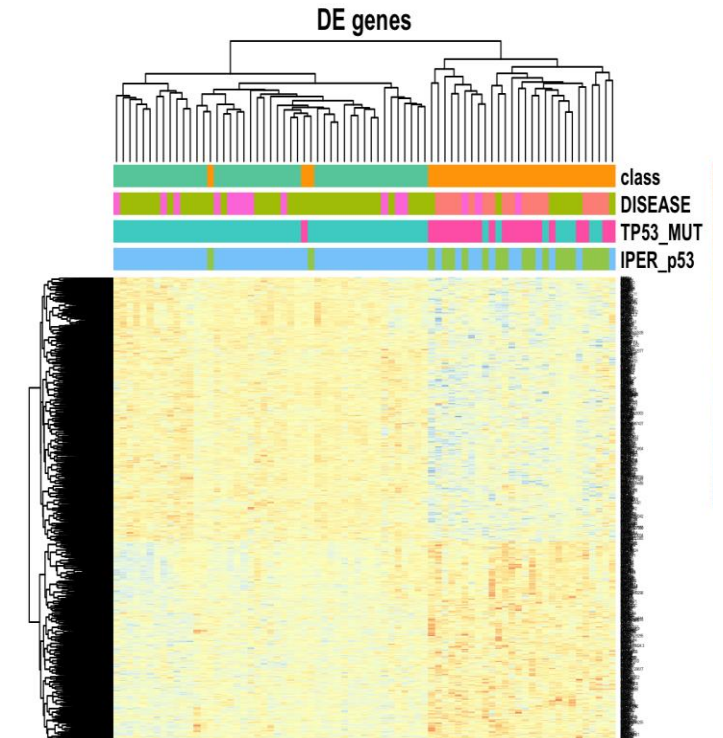


**MDS with p53 dysfunction
without TP53 mutations**

5-7%



RNA sequencing

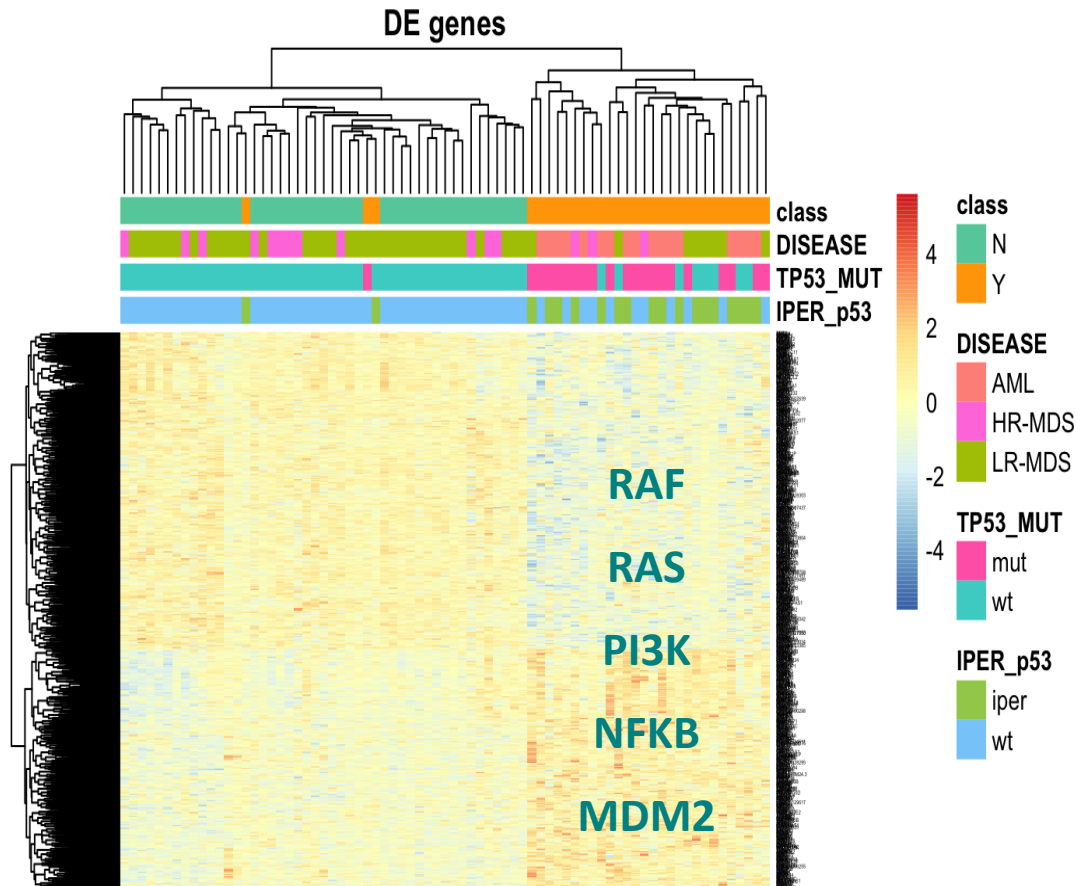


Bersanelli et al. J Clin Oncol 2021

Riva et al. Blood 2022;140:4001–4; Zampini et al, manuscript in preparation

A new category of high-risk MDS is defined according to p53 dysfunction

RNA sequencing of CD34+ progenitor cells from
236 MDS patients

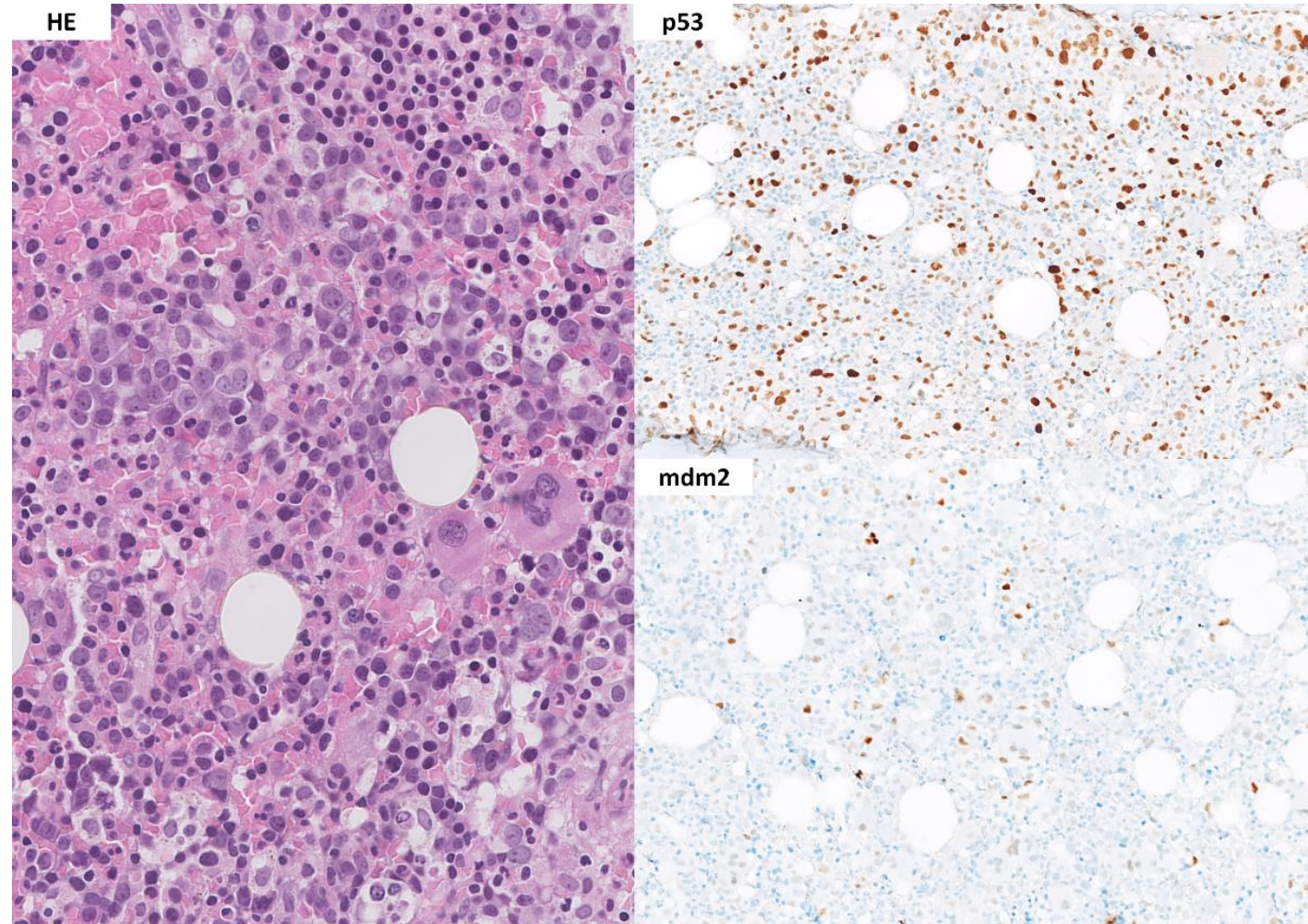


Evidence of impaired T cell and NK maturation and
function in MDS with p53 dysfunction

- Immune checkpoint overexpression (PD-L1) at the stem cell level
- Reduced numbers of cytotoxic T cells
- Expansion of myeloid-derived suppressor cells (MDSCs)
- Expansion of regulatory T cells (Tregs).
- **Impaired NK maturation and function**

Sallman DA Blood (2020) 136 (24): 2812–2823

A new category of high-risk MDS is defined according to p53 dysfunction



2022 WHO/ICC harmonization: an international project by GenoMed4all and icMDS

Aim

To create a hierarchical and harmonized classification of MDS based on a large comprehensive patient dataset using advanced statistical and machine learning methods of inference

Methods

- MCC/Genomed4all merged cohorts (n=7117)
- Hierarchical Dirichlet processes applied to define clusters capturing broad dependencies among gene mutations and cytogenetic abnormalities
- Patients are classified by neural networks
- SHAP was then used to explain the classification model by computing the contribution of each feature



Consensus Phase (panel of international experts - icMDS)

WHO – ICC harmonization proposal

DIAGNOSIS of MDS

- Presence of biTP53

YES

MDS with biallelic TP53 inactivation

NO

- 5q deletion alone, or with 1 other abnormality other than monosomy 7
- <5% BM blasts

YES

MDS with del(5q)

NO

- Presence of SF3B1 mutations
- Absence of del(7q), abn3q26.2, or complex
- Absence of RUNX1 mutations
- <5% BM blasts

YES

MDS with mutated *SF3B1*

NO

- Blast count $\geq 5\%$

YES

MDS with increased blasts

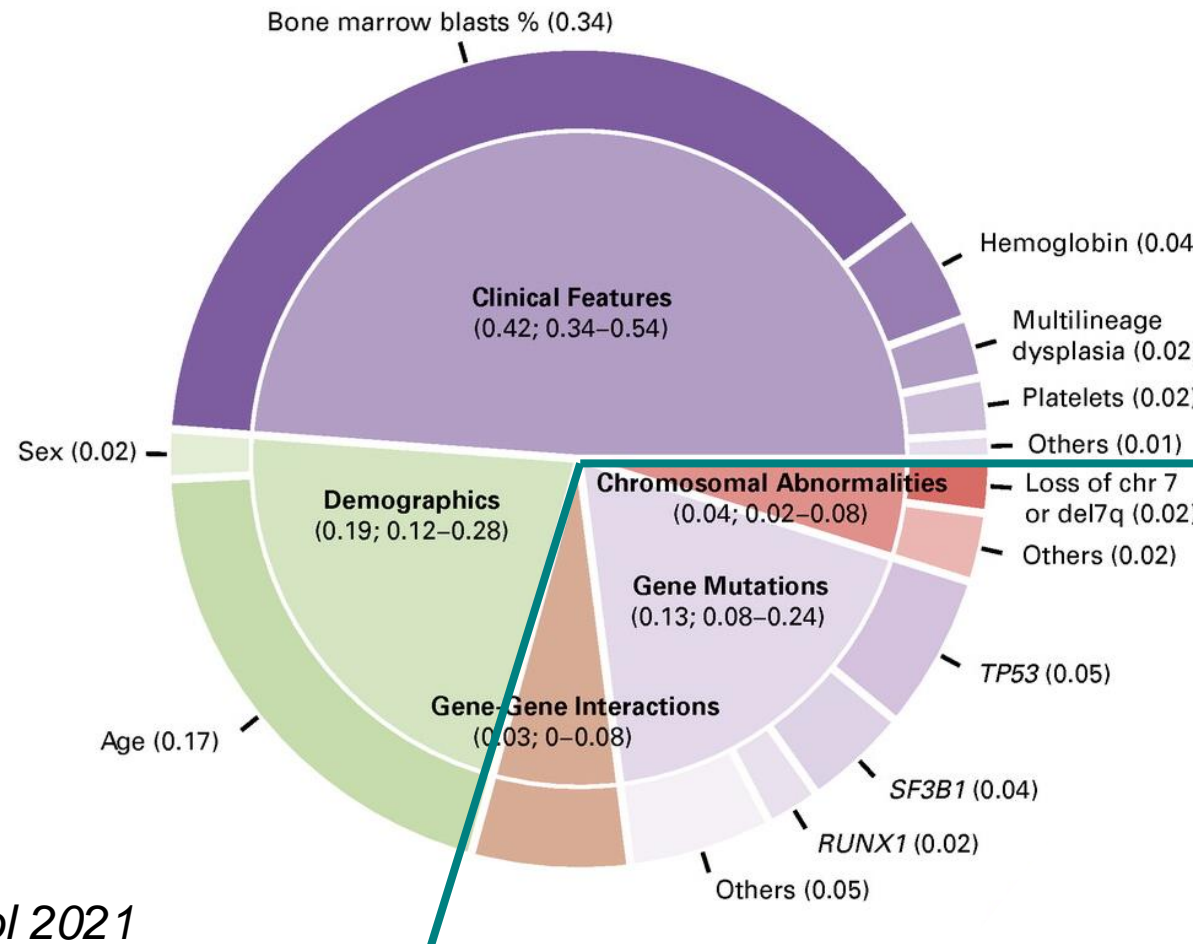
NO

MDS with low blasts

MDS with defining genetic abnormalities

MDS morphologically defined
(labels are to be refined by icMDS experts)

Contribution of molecular information to the prognostic assessment of MDS



Bersanelli et al. *J Clin Oncol* 2021
Maggioni et al. *Lancet Haematol* 2022
Sauta et al. *J Clin Oncol* 2023

Molecular IPSS (IPSS-M)

A new MDS prognostic model (**IPSS-M**) has been proposed by Bernard E et al*, expanding the original IPSS-R based on:

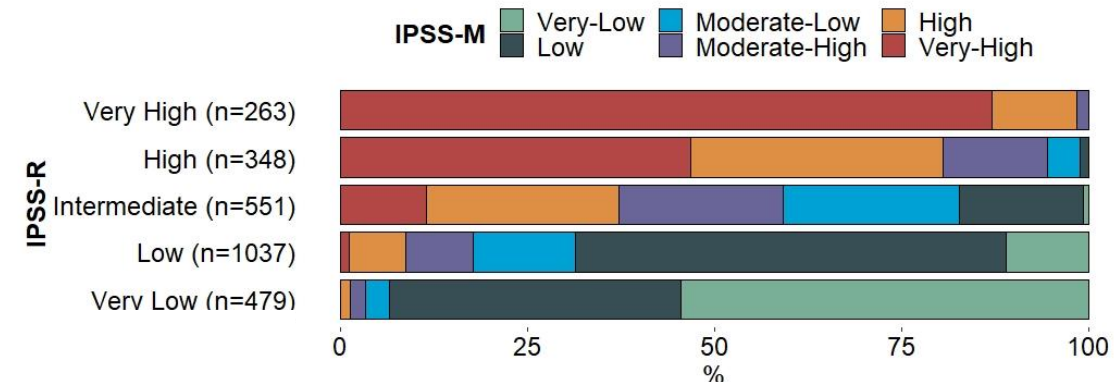
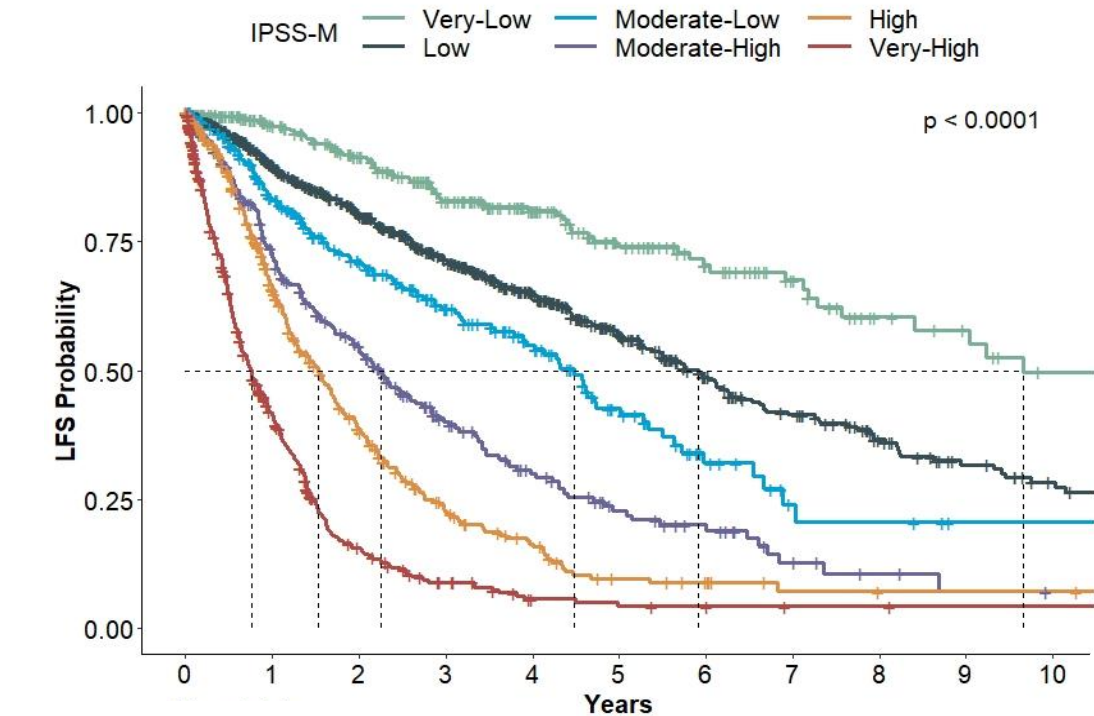
- Hematological parameters (hemoglobin, bone marrow blasts and platelets)
- Cytogenetic features

by including:

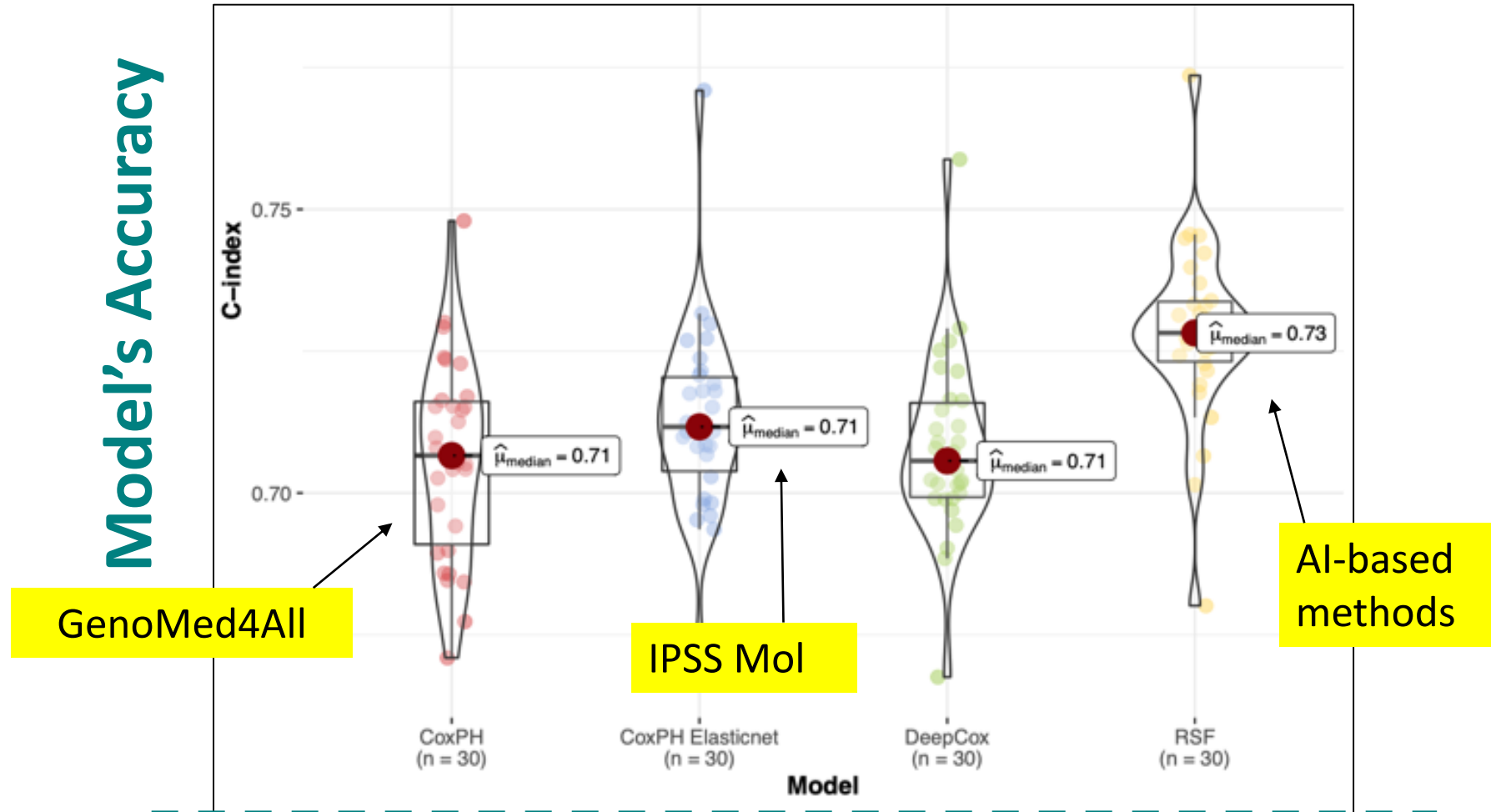
- **Somatic mutations on 31 genes (16 “Main Effect” genes and 15 “Residual Genes”),** weighting their impact on prognosis and clinical outcomes

*Bernard E et al. *NEJM Evid* 2022

Sautu E et al. *J Clin Oncol*. 2023 Mar 17;JCO2201784. doi:10.1200/JCO.22.01784.



Personalized prediction of overall survival in MDS. Statistical learning vs. Machine Learning methods



Artificial Intelligence (AI) for precision medicine

1- Machine Learning

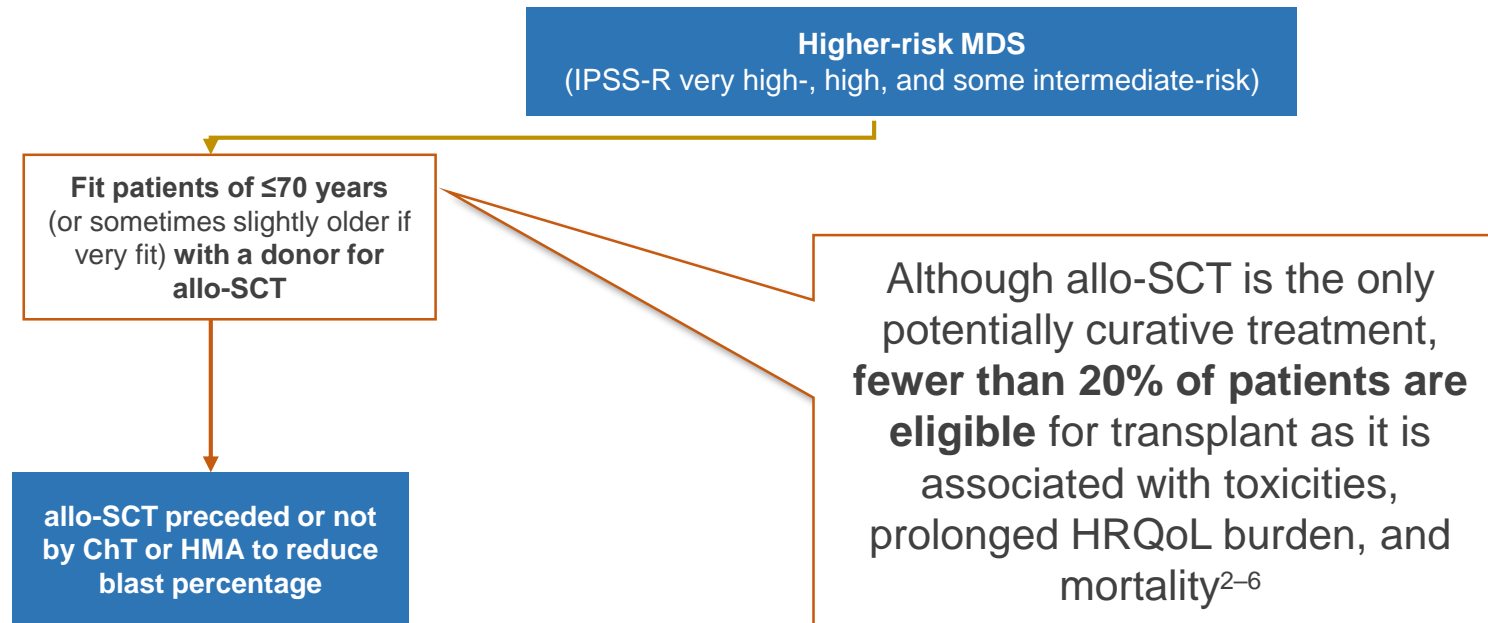


2- Generative AI



Current treatment options in high-risk MDS include HSCT and Hypomethylating agents (HMAs)

Higher-risk MDS carries a major risk of progression to AML and short survival; treatment should aim to modify the disease course, with options including allo-SCT and HMAs*¹

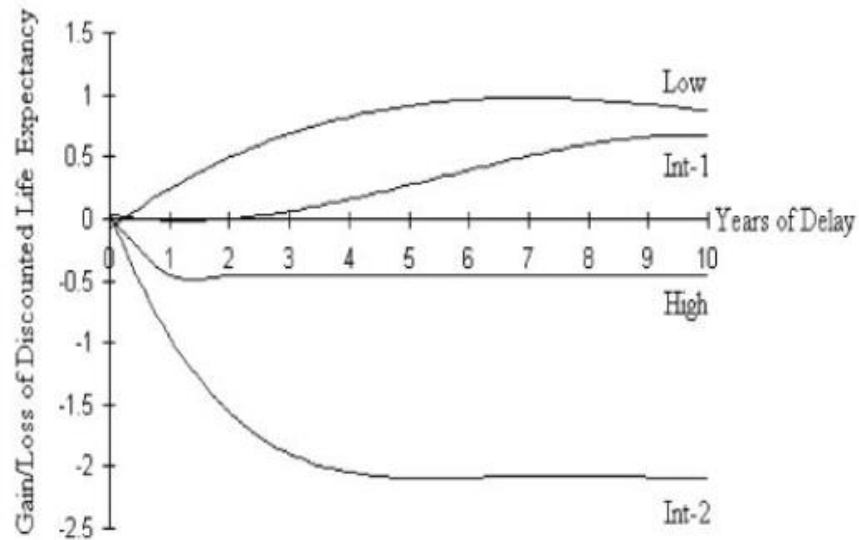


*According to ESMO Clinical Practice Guidelines

1. Fenau P, et al. Ann Oncol 2021;32:142–156; 2. Karopongse E, et al. Expert Rev Clin Immunol 2012;8:373–381; 3. Della Porta MG, et al. Blood 2014;123:2333–2342; 4. Malcovati L, et al. Blood 2013;122:2943–2964; 5. Sekeres MA, et al. J Natl Cancer Inst 2008;100:1542–1551; 6. Bhatt VR & Steensma DP. J Oncol Pract 2016;12:786–792

Transplantation strategy according to IPSS or IPSS-R

IPSS



Cutler CS et al. Blood 2004;104(2):579-85

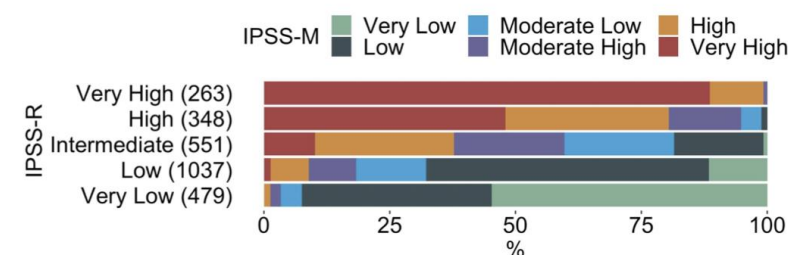
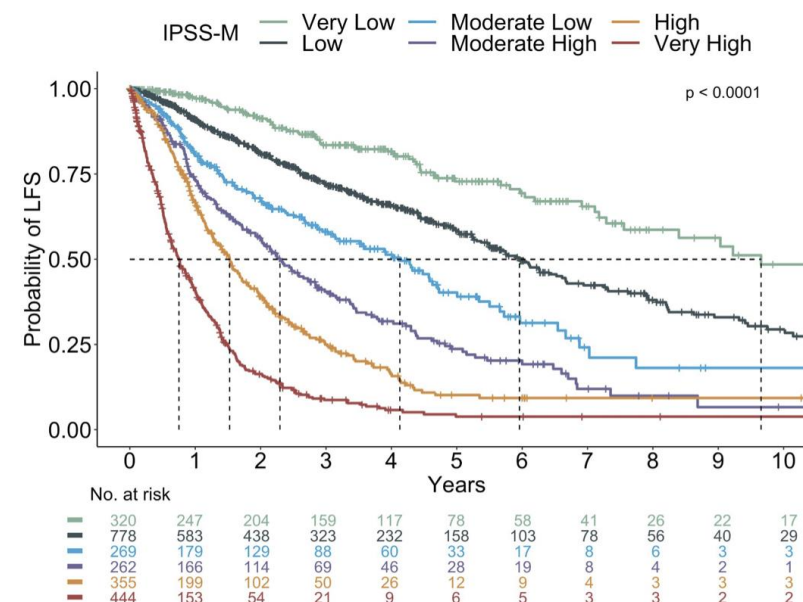
IPSS-R

	delay time (months)	40	50-55	>60
Years of life expectancy under policy 1: IPSS-R Low	0	16.4	16.1	15.1
	12	17.3	16.8	15.4
	24	17.9	17.3	15.6
	48	18.5	17.7	15.7
	60	18.7	17.9	15.7
Years of life expectancy under policy 2: IPSS-R intermediate	0	19.3	18.1	15.9
	12	17.9	17.1	14.9
	24	17.1	16.4	14.5
	48	16.3	15.7	14.2
	60	16.0	15.5	13.9

Della Porta MG et al Leukemia. 2017;312449-57

IPSS-M

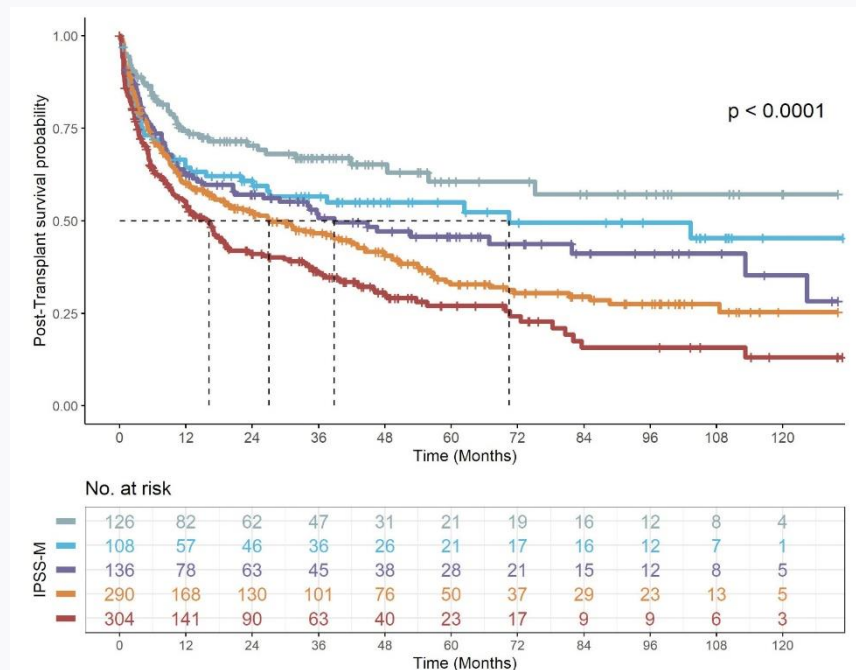
- Study Population: 2,957 patients
(validation on 718)
- The IPSS-M risk score consisted of:
 - hemoglobin, platelets and bone marrow blasts
 - IPSS-R cytogenetic category
 - 31 mutated genes
- No information on patients receiving disease modifying treatments



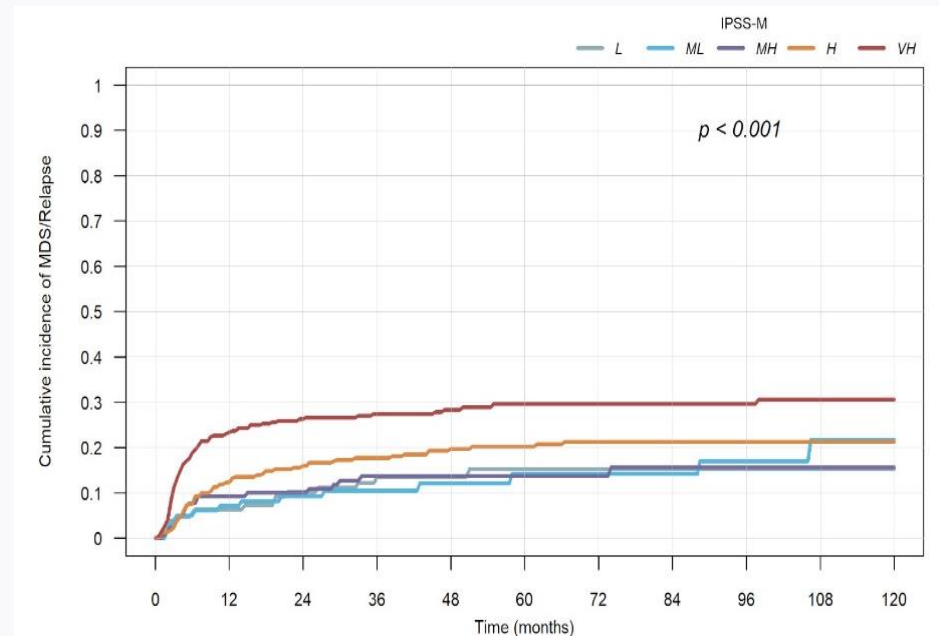
Molecular International Prognostic Scoring System for MDS (IPSS-M): Real World Validation by GENOMED4ALL (n=2,876)

IPSS-M on 964 patients who underwent allo-HSCT

Probability of overall survival



Probability of relapse



Median OS (months)	
L	-
ML	70.5
MH	38.6
H	27
VH	16.2

Sauta E et al. J Clin Oncol. 2023;41:2827-2842

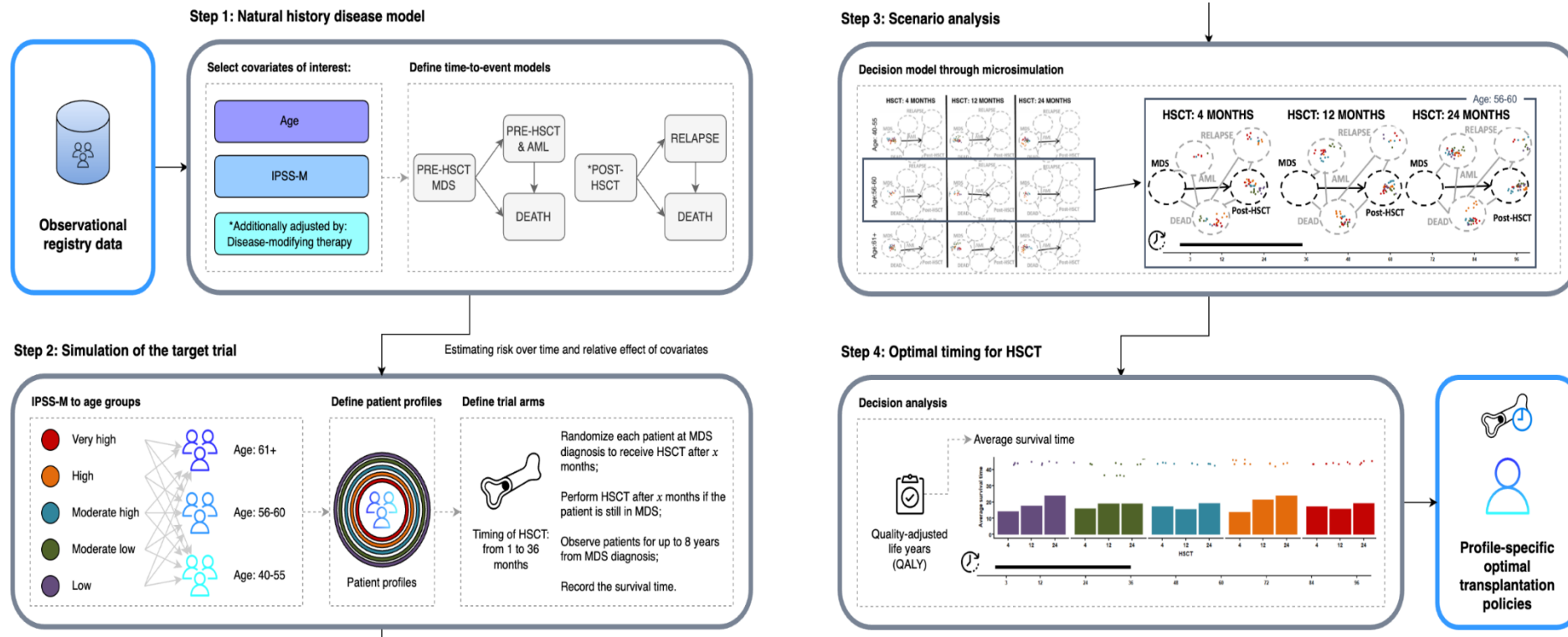
Controversy: Is the new IPSS-M helpful for decision making regarding allo HSCT?

- Mutation screening provides relevant prognostic information at individual patient level
- The implementation of IPSS-M is expected to result in a more effective selection of patients with early disease stage who are candidate to allo-HSCT
- Mutation screening may affect clinical decision making in transplantation (specific mutations are associated to a high probability of disease relapse)
- The definition of a minimum set of relevant genes may facilitate the clinical implementation of the score

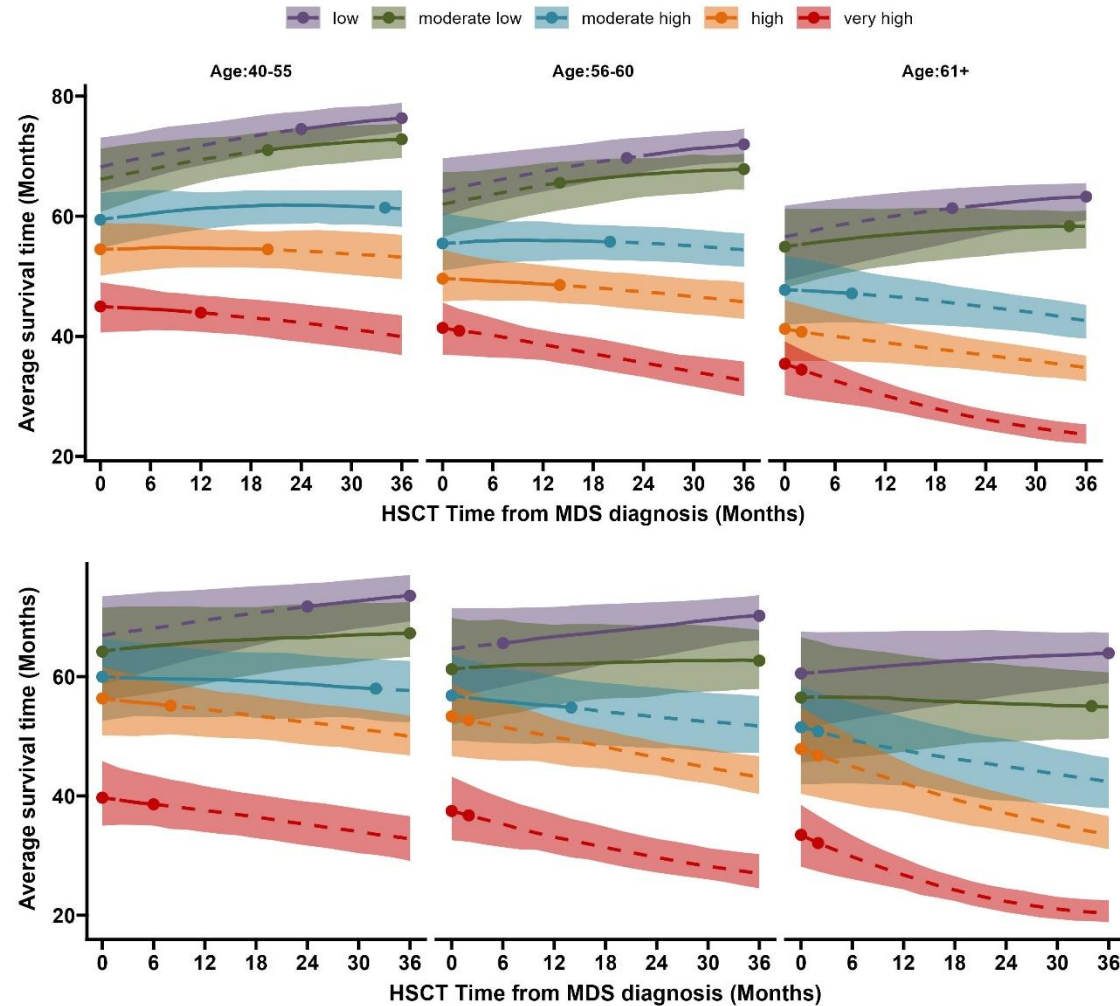
Clinical and genomic-based Decision Support System to define the optimal timing of HSCT in MDS

- Here, we aimed to develop and validate a Decision Support System to define the optimal timing of HSCT for MDS patients based on clinical and genomic information as provided by the IPSS-M.
- Patients and methods: retrospective population of 7118 patients, stratified into training and validation cohorts.
- A decision-strategy was built to estimate the average survival over an 8-year time horizon (Restricted_Mean_Survival_Time, RMST) for each combination of clinical and genomic covariates, and to determine the optimal transplantation policy by comparing different strategies.

Clinical and genomic-based Decision Support System to define the optimal timing of HSCT in MDS



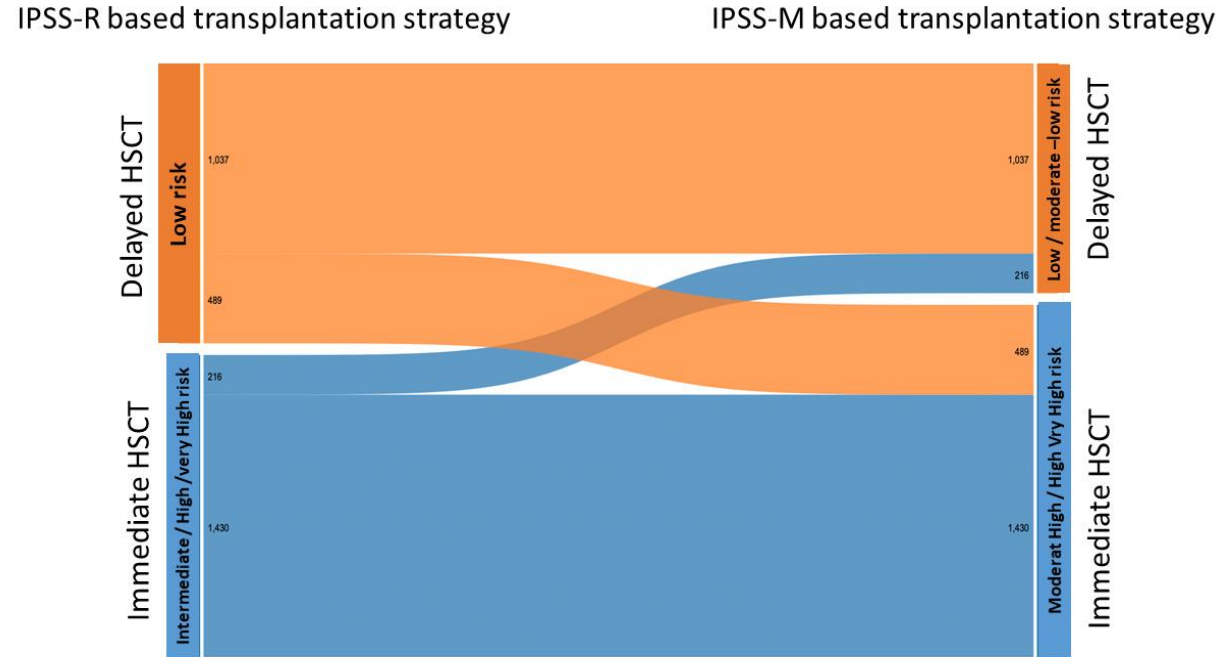
IPSS-M based transplantation strategy



TRAINING
n=3854

VALIDATION
n=2075

IPSS-M vs. IPSS-R based transplantation strategy



- 13% of candidates to be immediately transplanted under an IPSS-R based policy would benefit from a delayed strategy under an IPSS-M based policy, while 32% of candidates to delayed transplantation by IPSS-R, would benefit from immediate HSCT by IPSS.
- The comparison of the average conditional survival time for the optimal transplantation policies obtained using different scoring systems (IPSS-R/IPSS-M) resulted in a significant gain of life expectancy under an IPSS-M based policy across all age groups ($P=0.002$)

IPSS-M vs. IPSS-R based transplantation strategy

- Under an IPSS-M based policy, patients with either low and moderate-low risk benefit from a delayed transplantation policy, while in those belonging to moderate-high, high and very high risk categories immediate transplantation was associated with a prolonged life expectancy.
- Modelling decision analysis on IPSS-M vs. conventional IPSS-revised (IPSS-R) changed transplantation policy in a significant proportion of patients (45%)
- The grate majority of these patients would benefit for earlier transplantation under a IPSS-M vs. IPSS-R based policy, resulting in a significant gain-in-life expectancy under an IPSS-M based policy.
- These results provide evidence for the clinical relevance of including genomic features into the transplantation decision making process, allowing personalizing the hazards and effectiveness of HSCT in patients with MDS.

Artificial Intelligence (AI) for precision medicine

1- Machine Learning



2- Generative AI





CENTER FOR ACCELERATING LEUKEMIA/LYMPHOMA RESEARCH

*Artificial Intelligence and real world data analysis to improve
patient care and advance medical research in hematology*



GENOMED4ALL



Horizon2020
European Union Funding
for Research & Innovation



AI people

- Saverio D'Amico
- Elisabetta Sauta
- Gianluca Asti
- Victor Savevski
- Marilena Bicchieri
- Caterina Gregorio
- Marta Spreafico
- Francesca Ieva

Clinical team

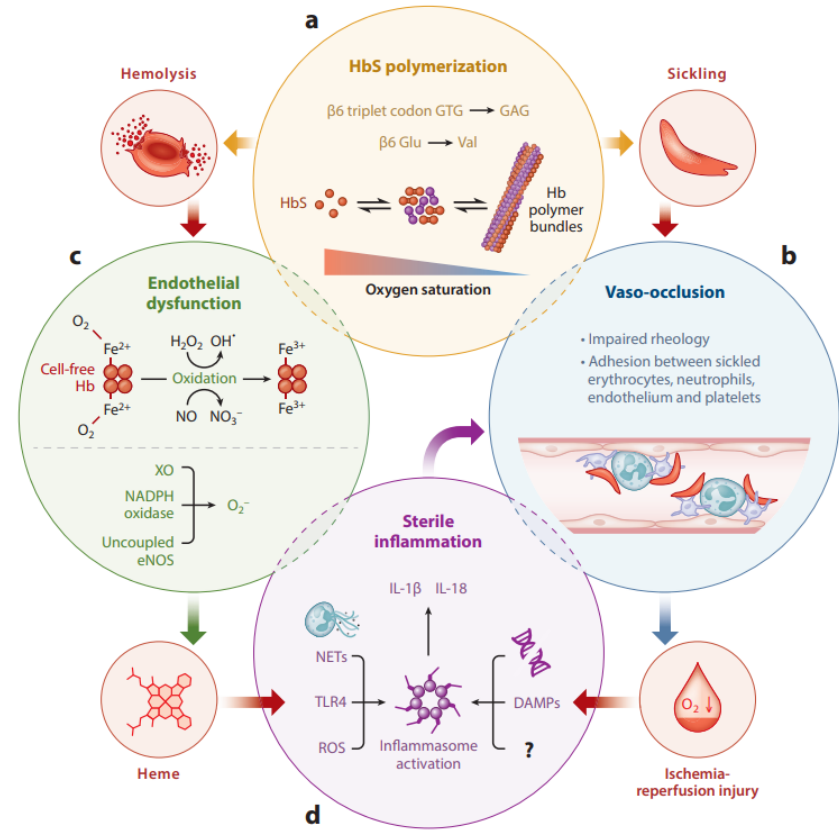
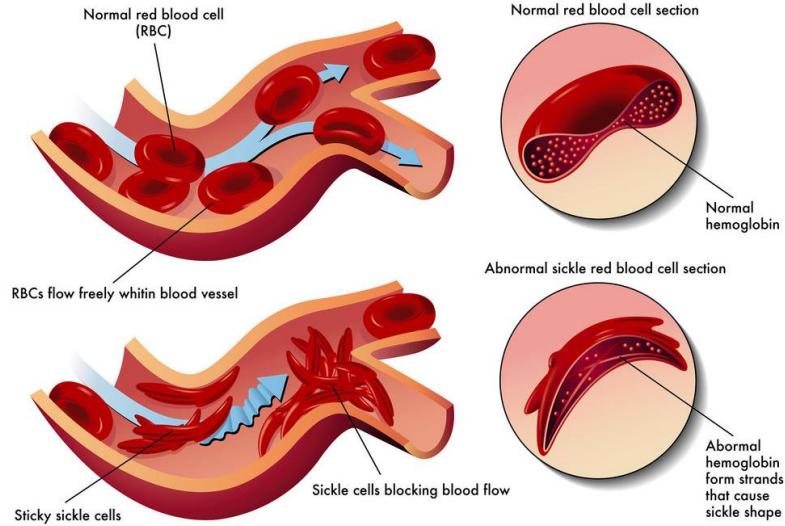
- Luca Lanino
- Giulia Maggioni
- Erica Travaglino
- Alessia Campagna
- Marta Ubezio
- Antonio Russo
- Gabriele Todisco

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| - Gastone Castellani | - Cesare Rollo |
| - Piero Fariselli | - Inigo Prada Luengo |
| - Anders Krogh | - Patricia Alonso |
| - Santiago Zazo | - Lin-Pierre Zhao |
| - Claudia Sala | - Pierre Fenaux |
| - Lorenzo Dall'Olio | - Torsten Haferlach |
| - Daniele Dall'Olio | - Maria Diez-Campelo |
| - Tiziana Sanavia | - Uwe Platzbecker |

Sickle Cell disease (SCD)

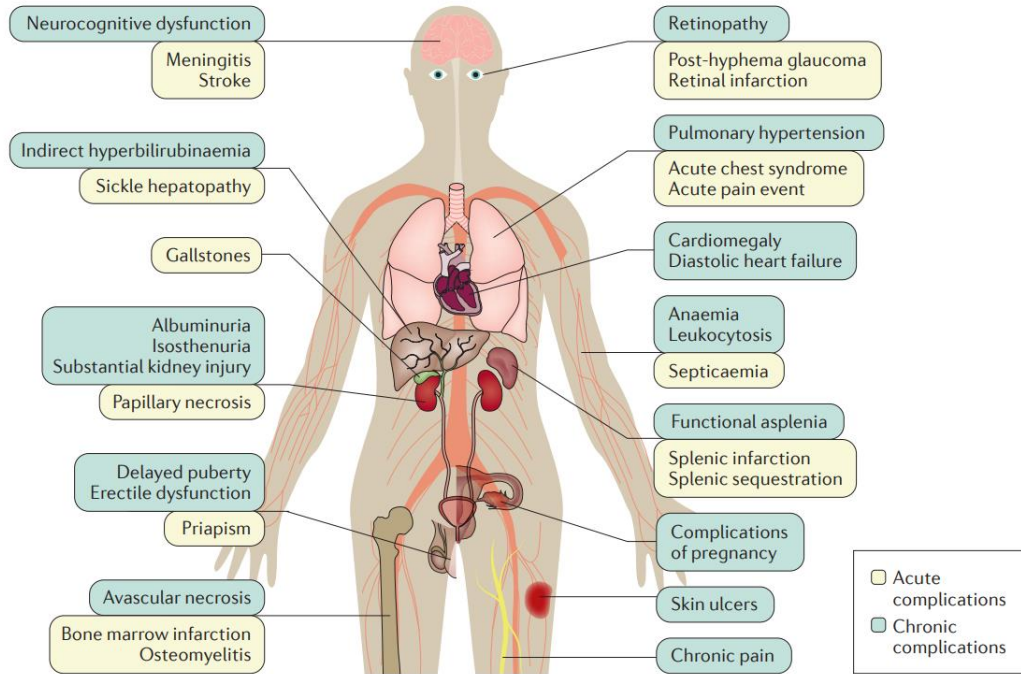
María del Mar Mañú Pereira, Raffaella Colombatti

Sickle cell disease



Kato, Gregory J., et al "Sickle cell disease." *Nature Reviews Disease Primers* 4.1 (2018): 1-22.

Sickle cell disease



- ❑ Chronic disease
- ❑ Multi systemic involvement
- ❑ Decreased life expectancy
- ❑ Low QoL

Kato, Gregory J., et al "Sickle cell disease." *Nature Reviews Disease Primers* 4.1 (2018): 1-22.

Sickle cell disease

Silent infarct

- Significant MORBIDITY:
 - affect 25% of children by the age of 6 and 40% by the age of 18 with consequences on cognition, schooling, working capacity and quality of life.
- BARRIERS to screening in the Real World Setting:
 - Screening of neurological complications and chronic organ damage is suboptimal in Europe
 - Not all centers have expert neuroradiologists to identify SCI
- PATHOPHYSIOLOGY, RISK FACTORS, RISK of PROGRESSION are incompletely understood

Disease-modifying and curative therapies

Hematopoietic Stem Cell Transplantation (HSCT)/gene therapy

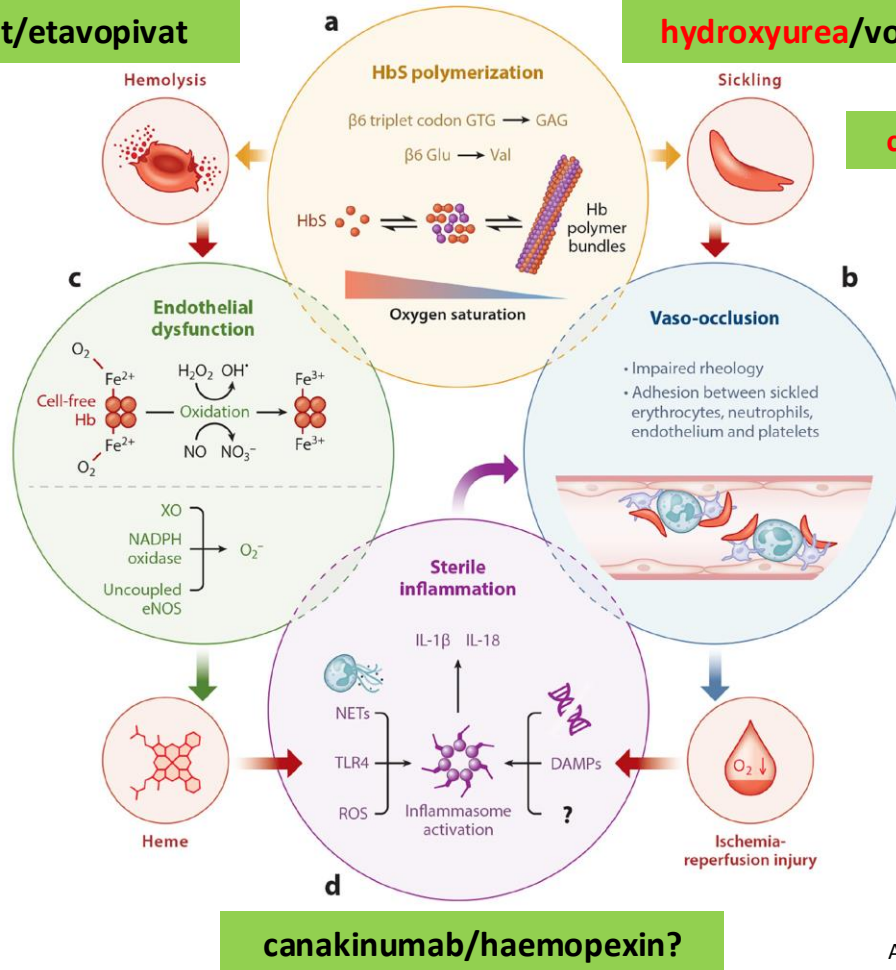
voxelotor/mitapivat/etavopivat

hydroxyurea/voxelotor/mitapivat/etavopivat

chronic transfusion regimen

hydroxyurea/crizanlizumab /decitabine

L-glutamin



Sickle cell disease

Research needs

- ❑ Monogenic disorder autosomal recessive BUT different genotypes and many disease modifiers 800+
- ❑ Complex and interconnected physio pathological pathways → high heterogeneity of clinical outcomes
- ❑ New treatments in the market
- ❑ Lack of functional tests for:
 - Disease classification
 - Prognosis
 - Response to treatment

Lack of large amount of high quality clinical and research data in SCD

Aim

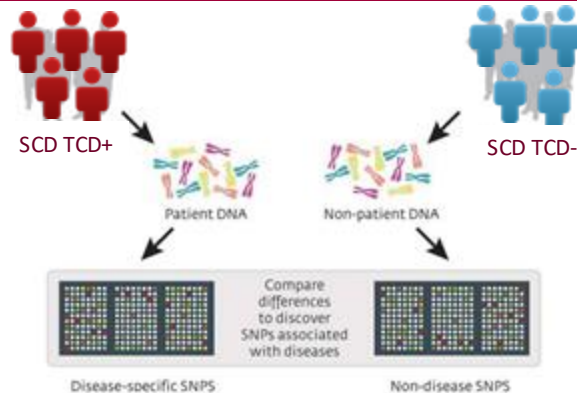
- ❑ Development of AI based scores combining clinical and research data to predict the risk and time of occurrence of severe complications:
 - ❑ Recurrent Vaso occlusive crisis
 - ❑ Acute chest syndrome
 - ❑ Silent infarcts
 - ❑ Stroke (hemorrhagic and ischemic)
 - ❑ Renal impairment
 - ❑ Hepatic failure
 - ❑ Retinopathy

Transcranial doppler
(TCD) ultrasound
screens for stroke risk

ERN-EuroBloodNet
Survey on TCD

What are the **genetic**
modifiers protecting or increasing
the risk for stroke in SCD?

GWAs SCD TCD+  vs SCD TCD- 

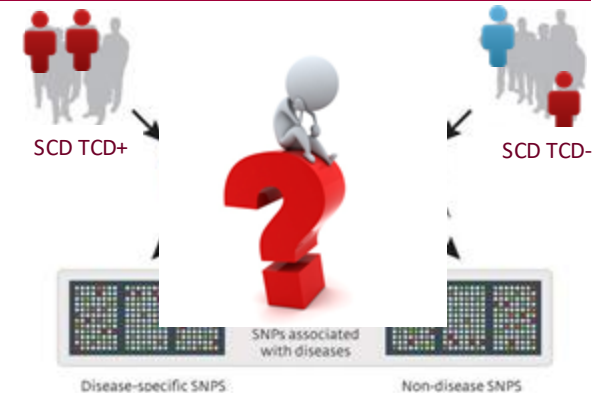


ERN-EuroBloodNet

>70% of SCD are not
correctly screened through
TCD for stroke risk

*Limited Access to Transcranial Doppler Screening and Stroke Prevention for Children with Sickle Cell Disease in Europe:
Results of a Multinational ERN-EuroBloodnet Survey*
D Cuzzubbo, V Gutierrez-Valle, M Casale, V Voi, C McMahon, M Mañá Pereira, M de Montalembert, B PD Inusa, R Colombatti
Blood (2021) 138 (Supplement 1): 915.

GWAs SCD TCD+  vs SCD TCD- 

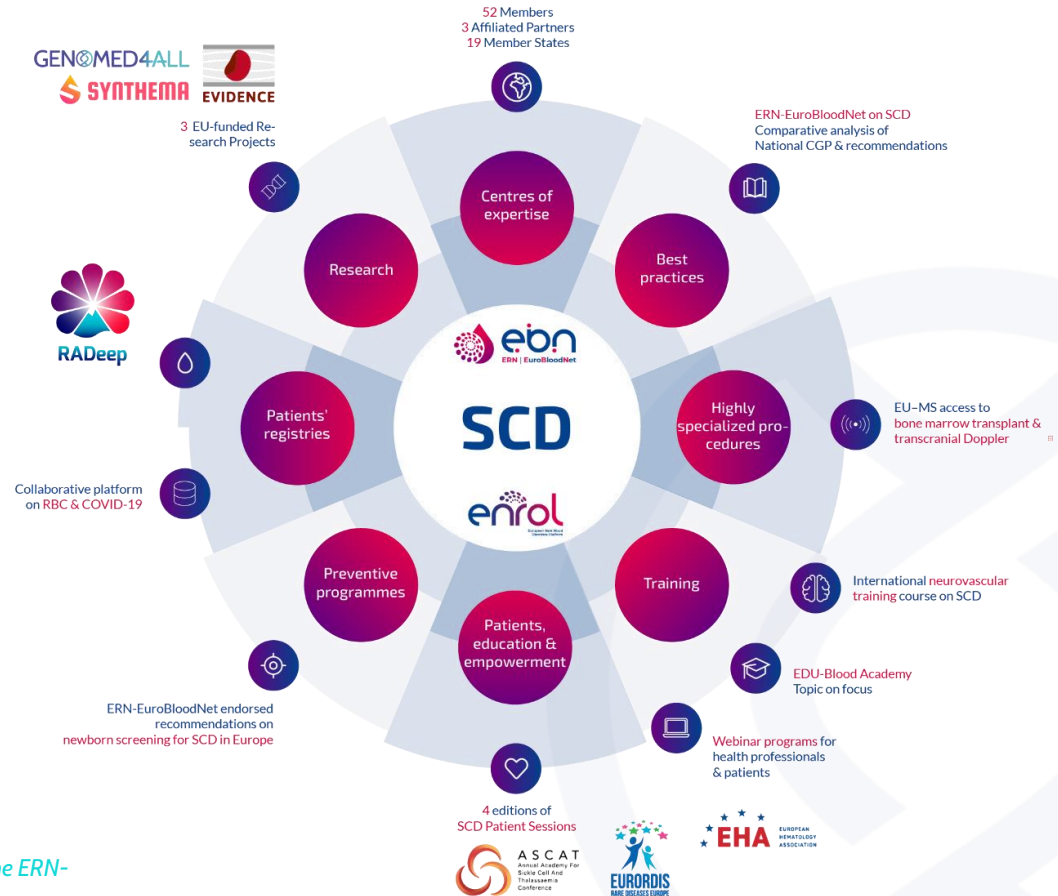


Sickle cell disease

Landscape and challenges in EU ERN-EuroBloodNet & RADeep

- 12 MS: Belgium, Cyprus, Denmark, France, Germany, Greece, Ireland, Italy, Portugal, Spain, Sweden and The Netherlands.
- Between 28,821-48,928 SCD patients based on experts' opinion (published and available data from existing registries, RADeep and data from national health insurance systems)
- 330 of the 447 million inhabitants of the EU, will be screened for the presence of SCD through national newborn screening (NBS) programmes in 2023. The average birth prevalence is in the order of 1:3000.

Mañú Pereira MDM et al. Sickle cell disease landscape and challenges in the EU: the ERN-EuroBloodNet perspective. Lancet Haematol. 2023 Aug;10(8):e687-e694.



Sickle cell disease

RADeep

- ❑ RADeep is an initiative endorsed by ERN-EuroBloodNet for pooling data from patients affected by a Rare anemia disorders (RADs)

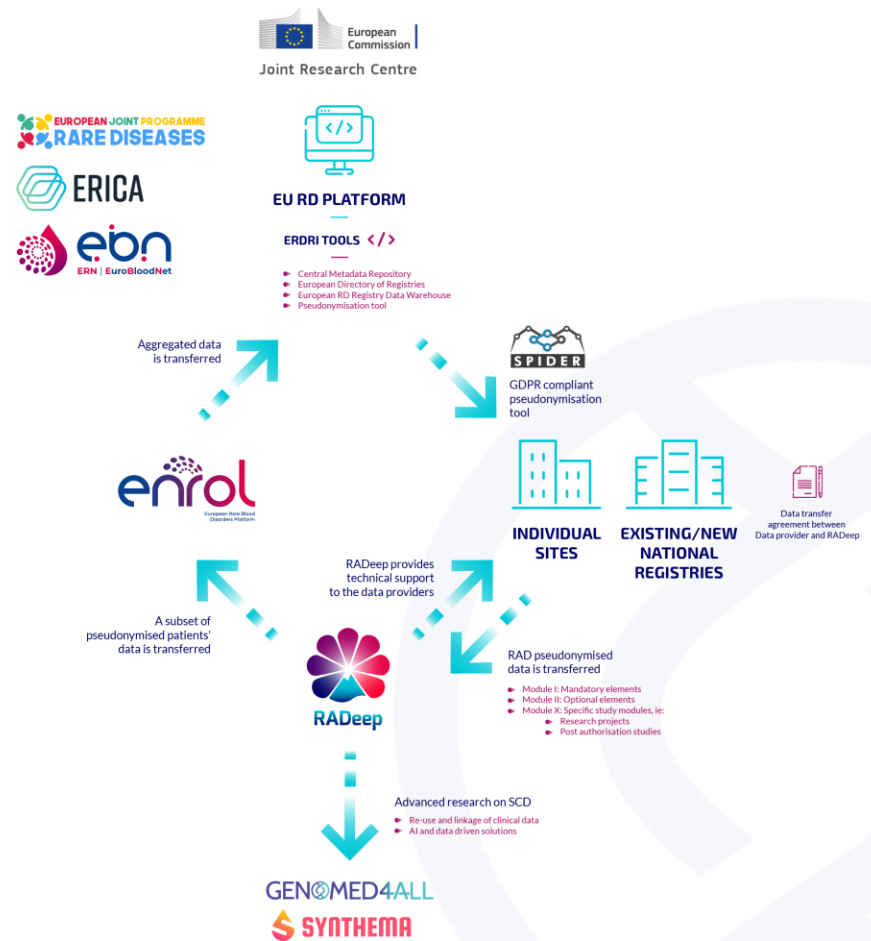
RADeep is built in line with the EU Rare Disease Platform recommendations for patients' registries on rare disorders and the [European Rare Blood Disorders Platform \(ENROL\)](#) (an umbrella for both new and already existing registries on rare hematological disorders).



Sickle cell disease

Clinical & Lab & Research Data

- ❑ Integrated with RADeep
- ❑ 400 Parameters
- ❑ Different types of data:
 - Demographics
 - Chronic organ damage
 - Organ damage assessment tests
 - Acute complications
 - Treatments
 - Offspring
 - Lab data
 - -omics: radiomics, metabolomics, GWAS
 - Oxygenscan
- ❑ Different formats of data = several ontologies (ORPHA, HPO, SCDO, LOINC, ATC...)



Sickle cell disease

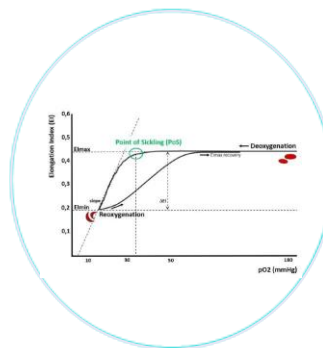
1,000 SCD Patients - Types of data



Genetic modifiers
800+ GWAS



Metabolomics
1900+ metabolites



Functional test
Rheology
LoRRca - Oxygenscan



MRIs



Clinical and lab
data

Sickle cell disease

ERN-EuroBloodNet Clinical sites

FR - AP-HP Henri-Mondor Paris – Pablo Bartolucci

Target: 300

FR - AP-HP Necker Paris – Mariane de Montalembert

Target: 150

IT - Università di Padova Padova - Raffaella Colombatti

Target: 150

ES - VHIR-HUVH Barcelona - Mar Mañú Pereira

Target: 250

H Gregorio Marañón - Elena Cella

H Sant Joan de Deu – Anna Ruiz

H Virgen del Rocío – Salvador Payán

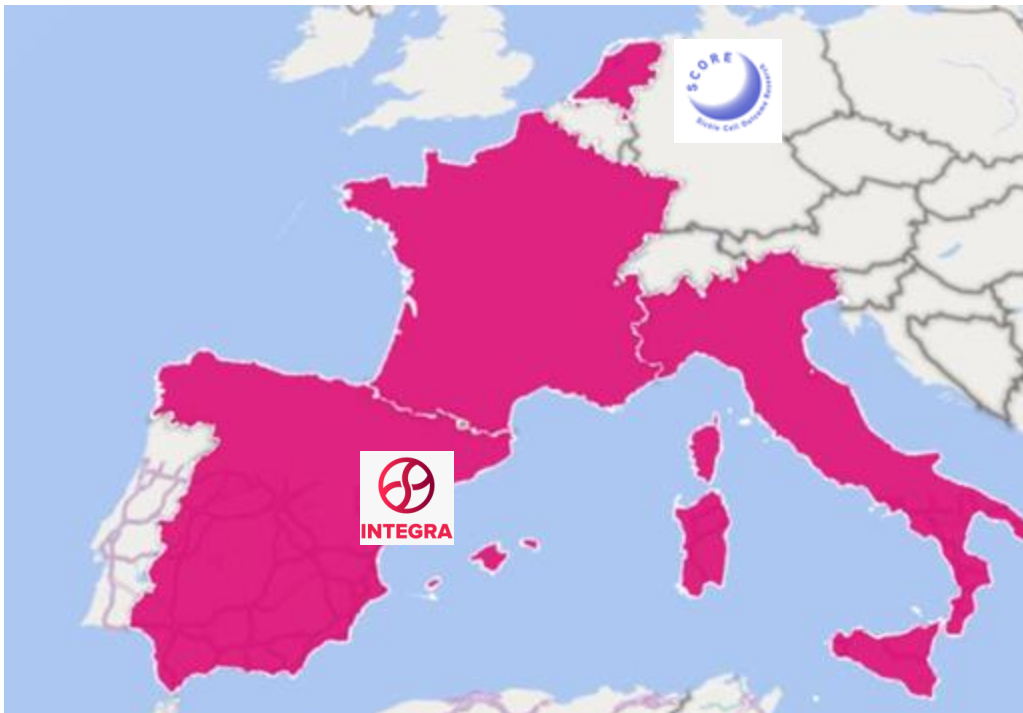
NL - UMCU Utrecht, Netherlands - Edward Van Beers

Target: 250

Amsterdam – Bart Biemond

Erasmus MC – Marjon Cnossen

Radboud UMC – Saskia Schols



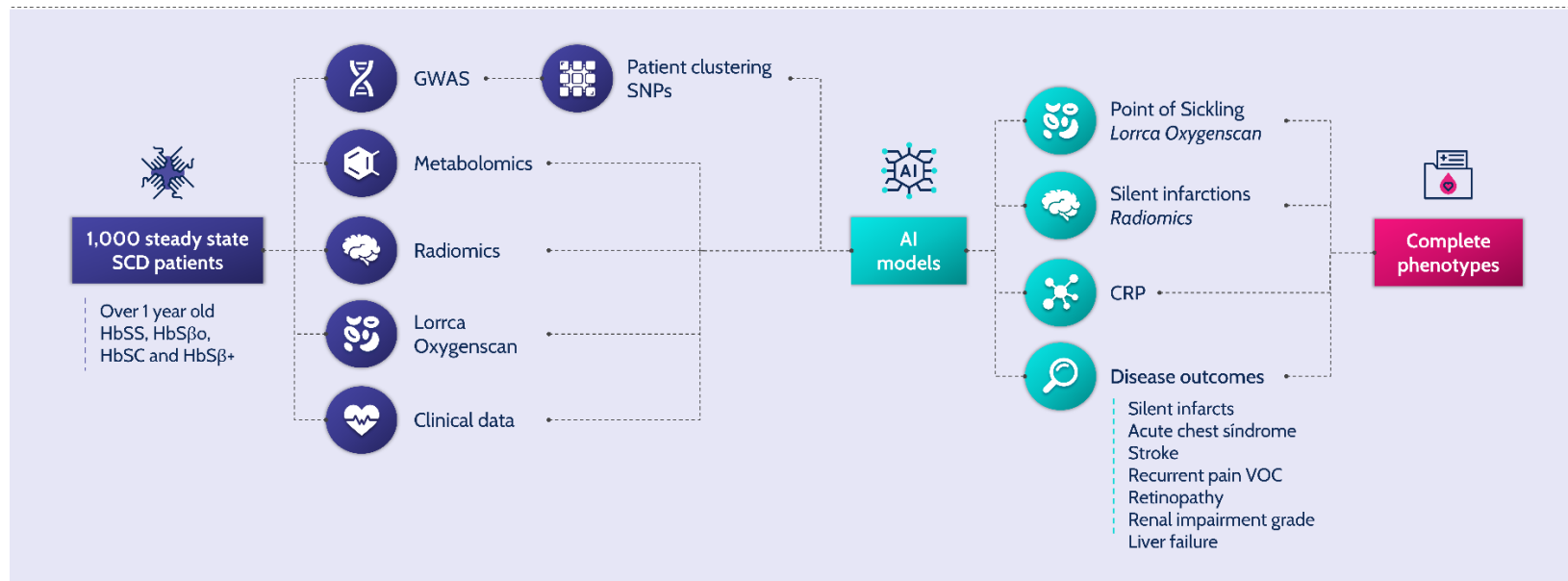
Sickle cell disease

AI Aims in Genomed4All

- ❑ Aim 1: To allocate SCD patients in different groups according to their genomic profile and assess correlation with inflammation related markers, point of sickling and phenotype.
- ❑ Aim 2: To allocate SCD patients in different groups according to their metabolomic profile and assess correlation with inflammation related markers, point of sickling and phenotype
- ❑ Aim 3: Develop a probability score using AI-based brain MRI image analysis (radiomics) to:
 - ❑ Diagnose already existing lesions (silent infarcts) that might go undetected due to inter-observer variability or lack of expertise on site.
 - ❑ Predict the risk of developing silent infarcts in the future: Correlation between occurrence of silent infarcts and clinical, genomic, metabolomic and functional laboratory features.
- ❑ Aim 4: Based on results from objectives 1 to 3, to develop predictive risk scores for the appearance of most prevalent and severe clinical outcomes (e.g., stroke, vaso-occlusive crisis, acute chest syndrome).
- ❑ Aim 5: Based on results from objectives 1 to 3, to develop predictive risk scores over time for the appearance of most prevalent and severe clinical outcomes (e.g. stroke, vaso-occlusive crisis, acute chest syndrome).

Sickle cell disease

Study design



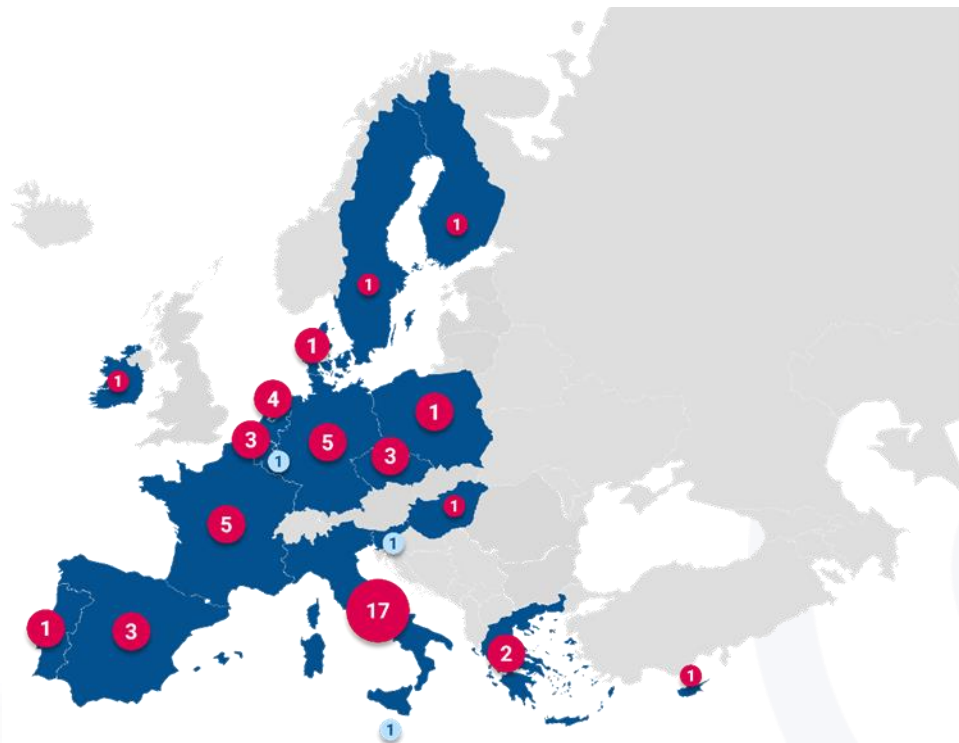
Sickle cell disease


Clinical sites



RBC subnetwork

Members
Affiliated Partners





Data driven solutions pooling EU standardized clinical data and research data is crucial to advance research in Sickle cell disease



Genomed4all & ERN-EuroBloodNet & RAdDeep



Thanks!
Any questions?

GenoMed4All & ERN-EuroBloodNet

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Acknowledgements



**European
Reference
Network**

for rare or low prevalence
complex diseases



Network

Hematological
Diseases (ERN EuroBloodNet)



**Co-funded by
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This project is supported by the European Reference Network on Rare Haematological Diseases (ERN-EuroBloodNet)-Project ID No 101085717. ERN-EuroBloodNet is partly co-funded by the European Union within the framework of the Fourth EU Health Programme.

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Educational Program
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GenoMed4All & ERN EuroBloodNet for precision medicine in hematology

Federico Álvarez

GenoMed4All Coordinator – UPM

Use cases challenges: MDS, SCD and MM

Multiple Myeloma

The need to have information about CLINICS and BIOLOGY

Dr Marina Martello, PhD

*DIMEC - Dipartimento di Scienze Mediche e Chirurgiche -
Università di Bologna*

*IRCCS Azienda Ospedaliero-Universitaria di Bologna - Istituto di
Ematologia "Seràgnoli"*

Multiple Myeloma

The need to have information about CLINICS and BIOLOGY

CLINICS



ARTIFICIAL
INTELLIGENCE?



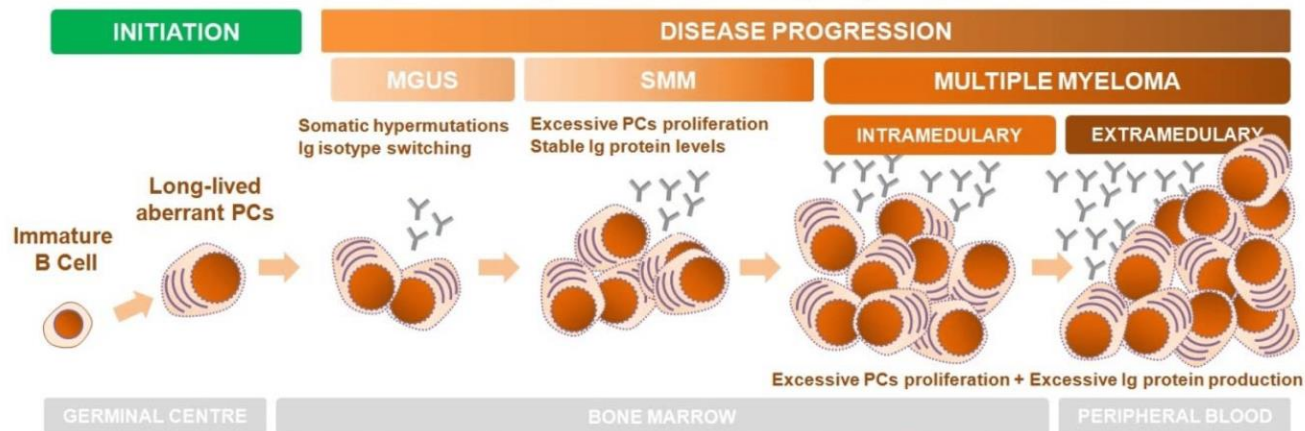
BIOLOGY



Multiple Myeloma(S)

By name and by nature!

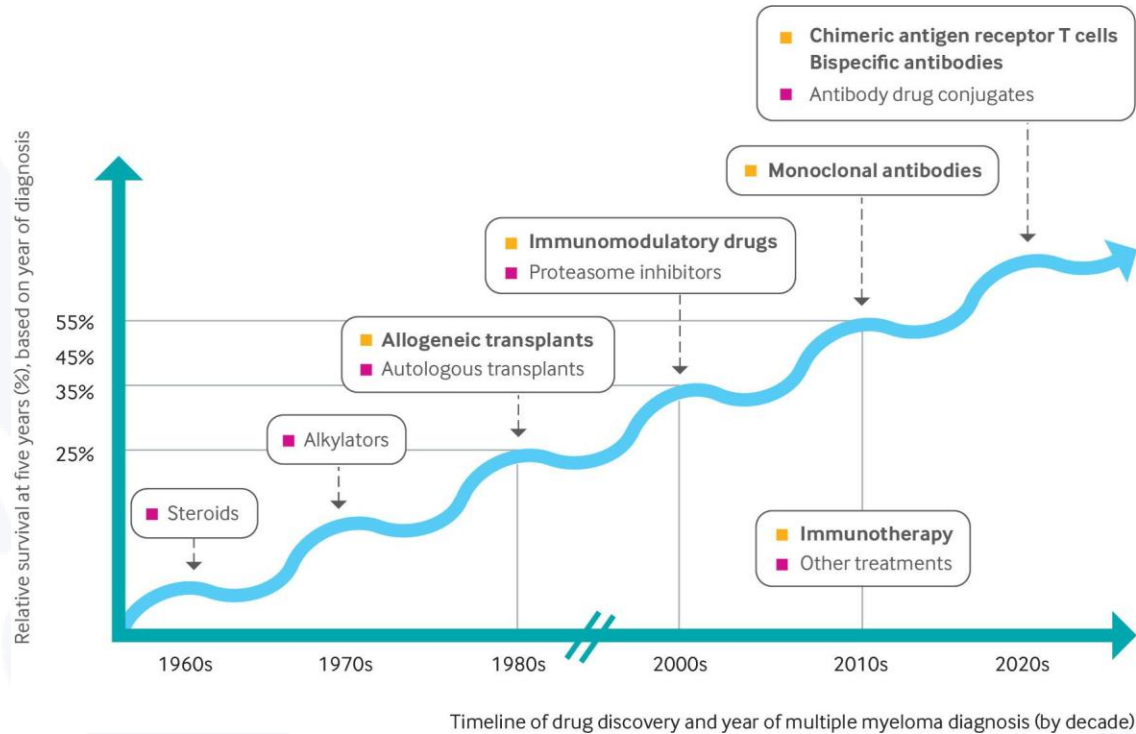
- **PLASMA CELLS** disease: inside bone marrow, but also outside
- **MULTI-STEP** evolution: from MGUS to MM
- Disease **DISTRIBUTION**: presence of multiple lytic lesions



LIQUID BIOPSY

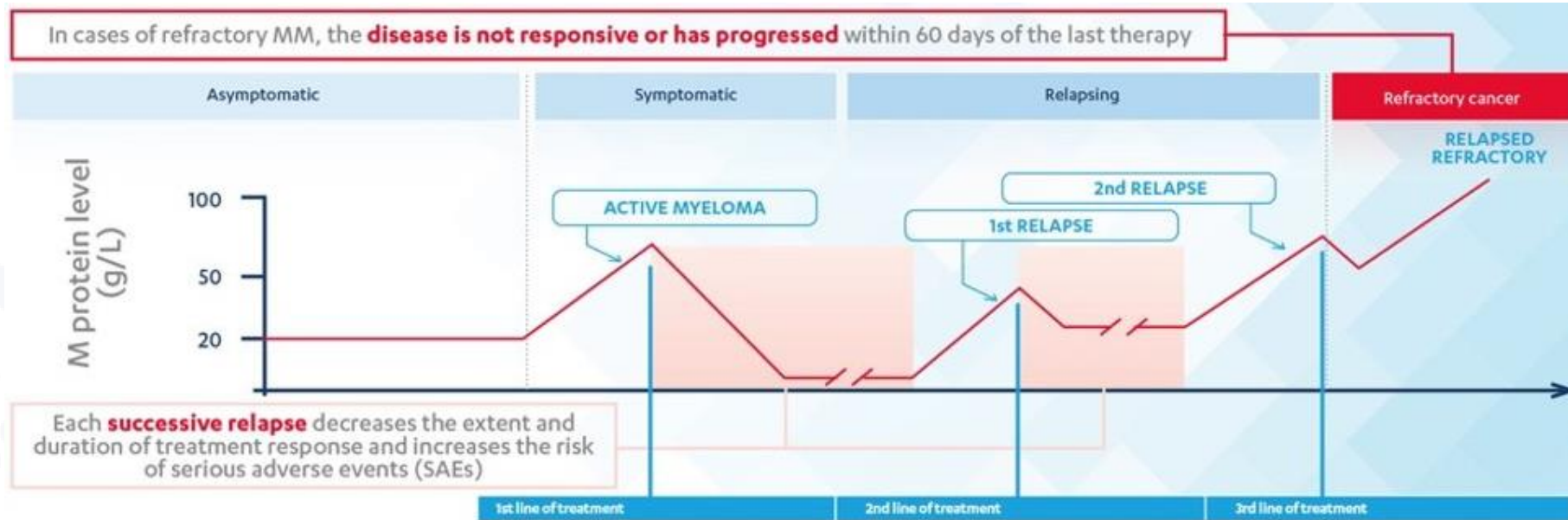
Multiple Myeloma CLINICS

State-of-the-art



Multiple Myeloma CLINICS

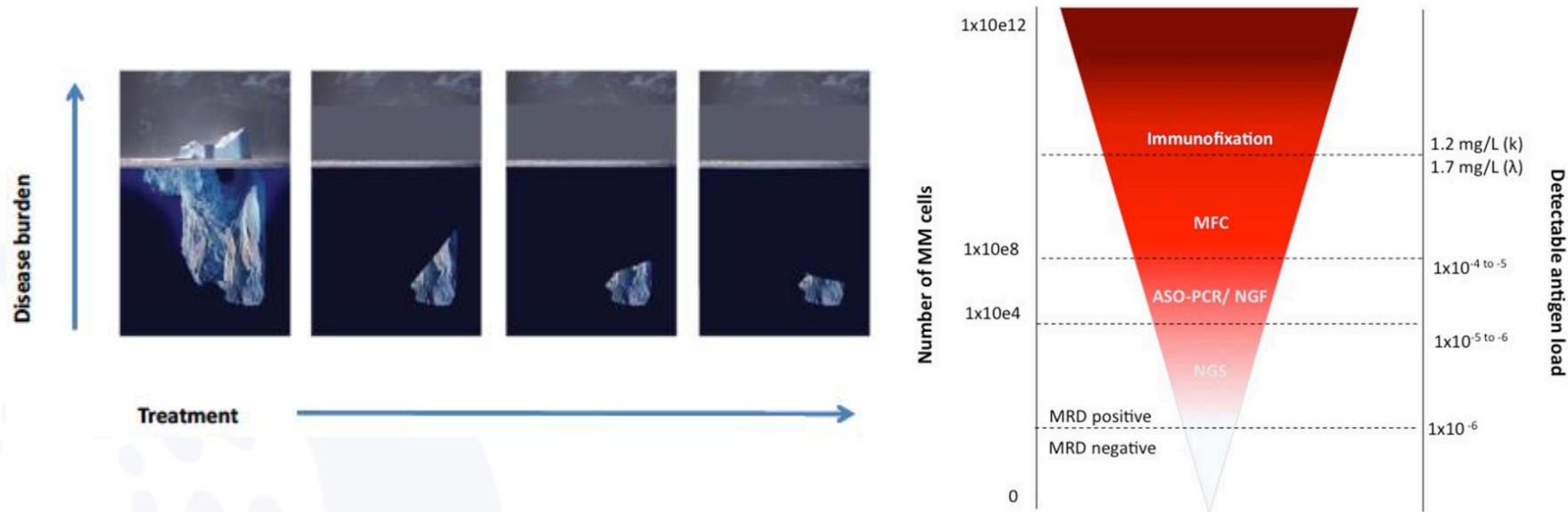
Myeloma follows a pattern of alternance of remission and relapse phases



AIM: to reach REMISSION phase and maintain this status

Multiple Myeloma CLINICS

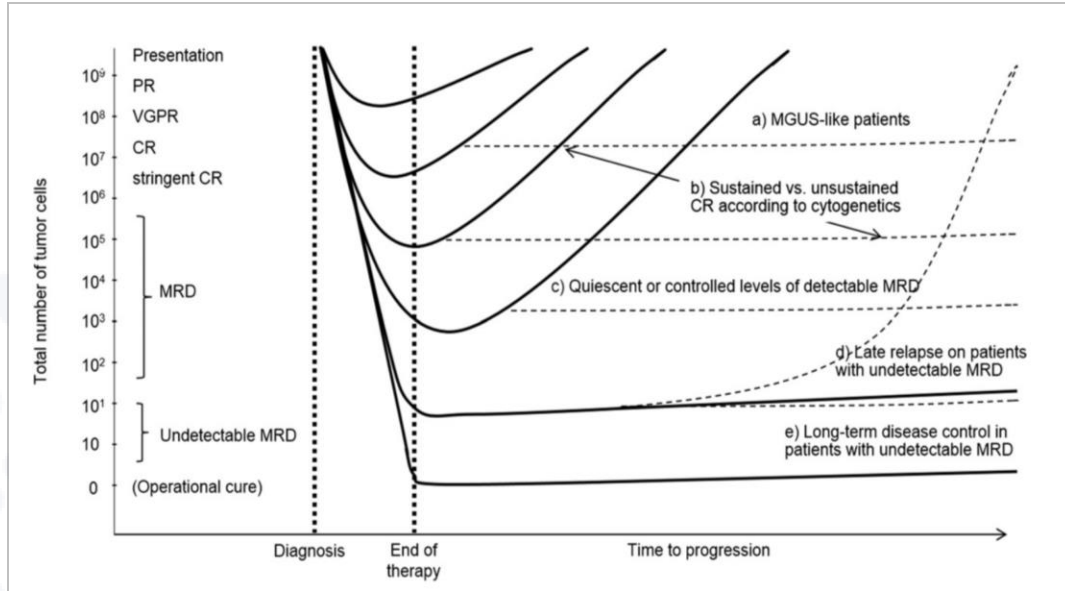
REMISSION concept evolution over last years until UNDETECTABLE MRD



The main goal of current therapies, is to minimize the disease present at diagnosis to MRD, a time when the patient can be defined as being in COMPLETE REMISSION. The deeper and longer this status, the greater the survival

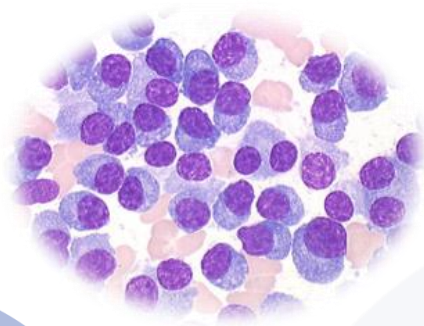
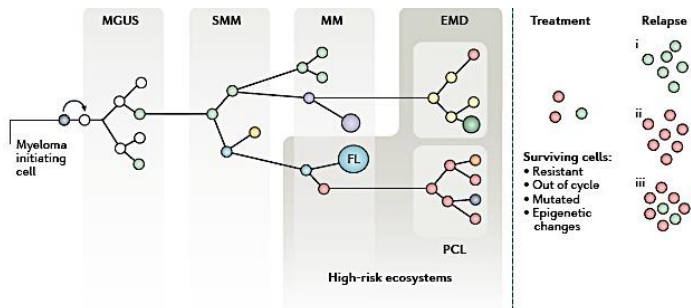
Multiple Myeloma CLINICS?

... also BIOLOGY matters!!

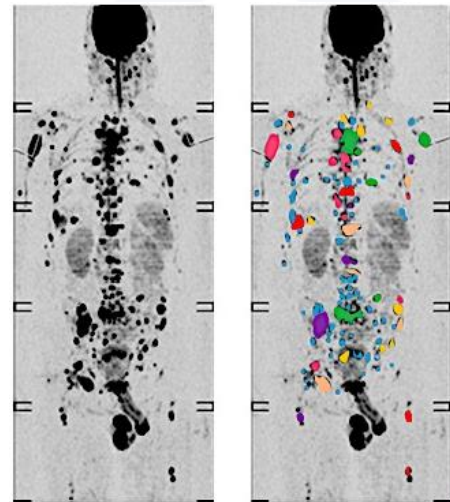
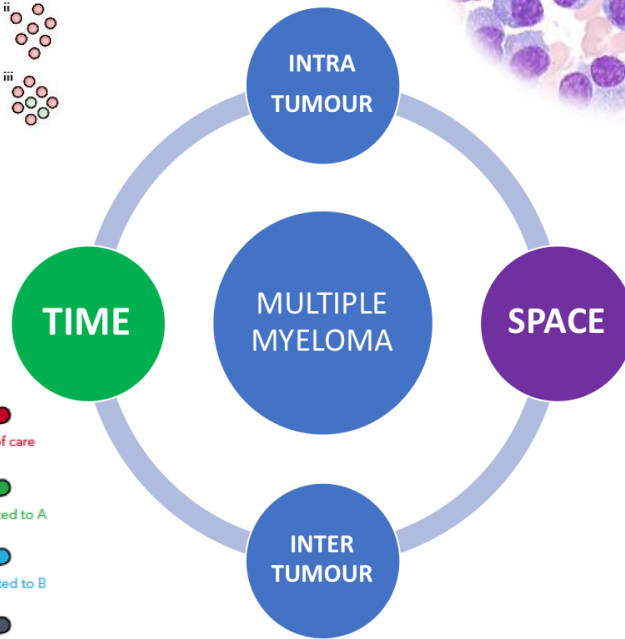
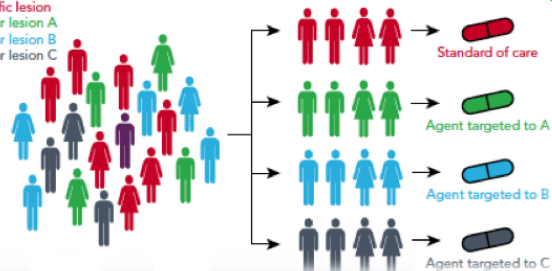


Multiple Myeloma BIOLOGY!

How to face with this multilevel complexity

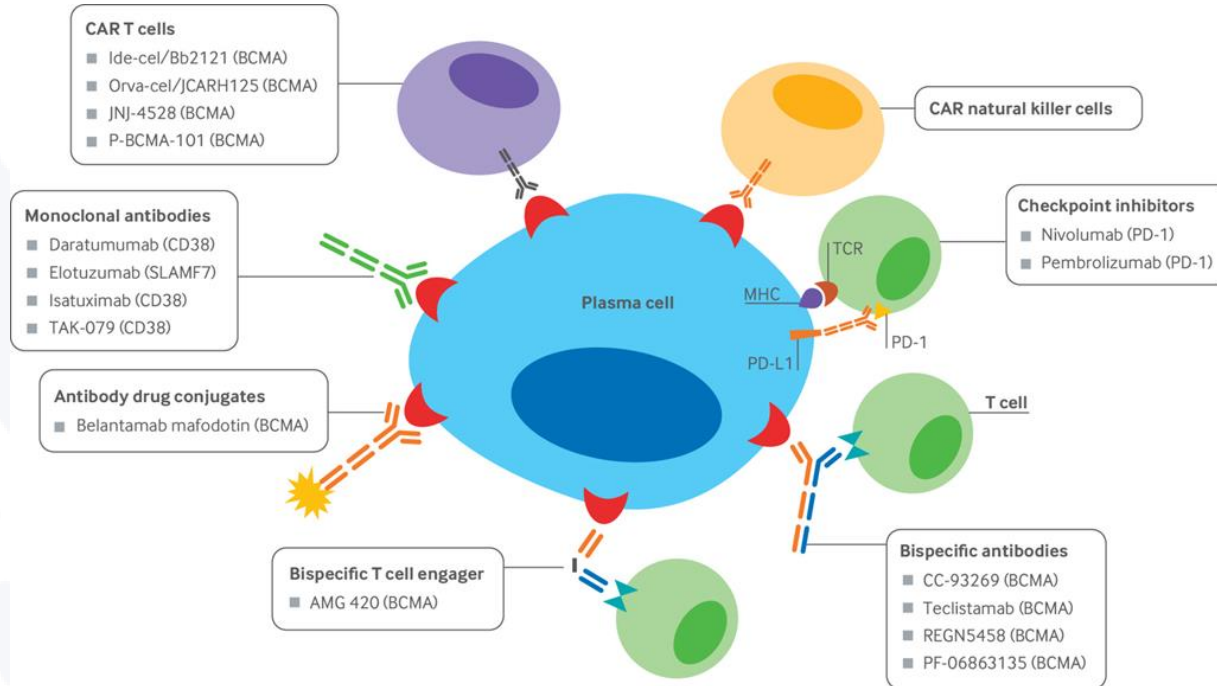


Myeloma patients:
 No specific lesion
 Molecular lesion A
 Molecular lesion B
 Molecular lesion C



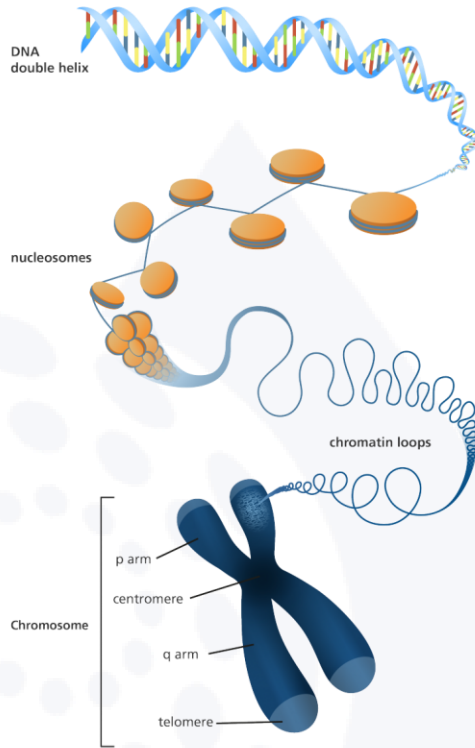
Multiple Myeloma BIOLOGY! ...outside the cells

It is essential to know BIOLOGY of the disease, including CLONE and MICROENVIRONMENT



Multiple Myeloma BIOLOGY!... Inside the cells

Genomic abnormalities



Primary abnormalities

Trisomies (~45%)

Odd-numbered chromosomes: 3, 5, 7, 9, 11, 15, 19, and 21

IgH translocations (~55%)

Translocations involving the IgH gene locus at 14q32

Translocation;locus;gene

t(4;14);4p16;FGFR3-MMSET
t(14;16);16q23;MAF
t(14;20);20q12;MAFB
t(8;14);8q24;MAFA
t(11;14);11q13;CCND1
t(6;14);6p21;CCND3

MYC dysregulation

Cyclin dysregulation

Secondary abnormalities

Monosomies

Chromosome 13
Chromosome 17
Chromosome 14

Deletions

Chromosome 17p
Chromosome 1p

Amplification

Chromosome 1q
gain or amplification

Other genomic alterations
miRNA

Recurrent mutations

KRAS

NRAS

TP53

DIS3

FAM46C

BRAF

TRAF3

ROBO1

CYLD

EGR1

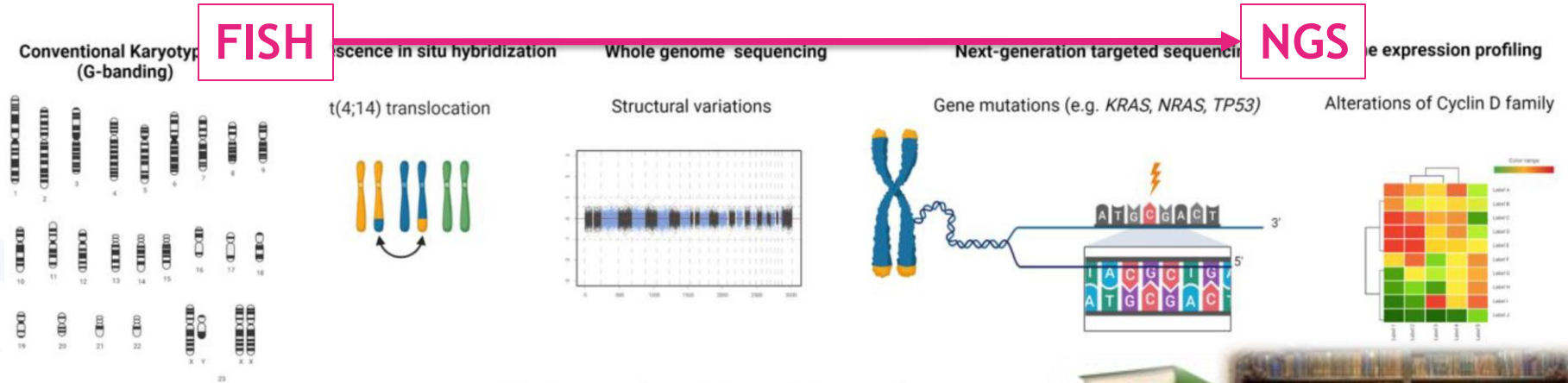
SP140

FAT3

CCND1

Multiple Myeloma BIOLOGY

Techniques used for studying the genomics of multiple myeloma



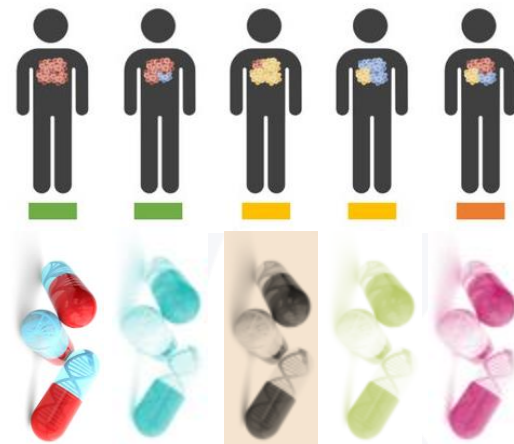
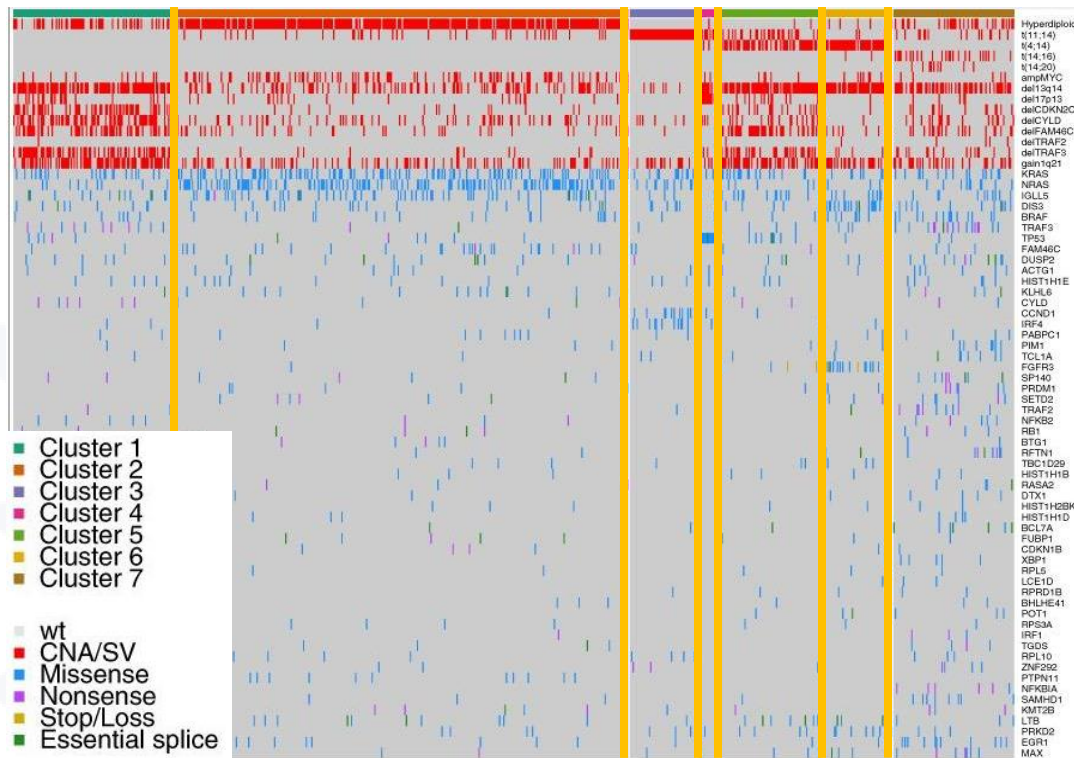
Techniques for studying MM genetics

Magnific Chaos



Multiple Myeloma BIOLOGY

Different patients = different myeloma = different subtypes



Multiple Myeloma CLINICS&BIOLOGY

Current MM risk stratification scores are suboptimal

Stage	ISS[3]	R-ISS[4]	IMWG[5]
I Low Risk	β 2-microglobulin <3.5 mg/L Albumin \geq 3.5 g/dL	ISS Stage I + Standard risk FISH + Normal LDH	ISS I or II Age < 55 No t(4;14), del 17p, 1q gain
II Standard Risk	Not stage I or III	Not stage I or III	Neither low or high risk
III High Risk	β 2-microglobulin \geq 5.5 mg/L	ISS stage III + t(4;14), t(14;16), or del 17p or LDH > upper limit	ISS 2 or 3 + t(4;14) or del 17p
Median OS	Stage I: 62 months Stage II: 44 months Stage III: 29 months	Stage I: 66 months Stage II: 42 months Stage III: 29 months	Low Risk: >120 months Standard Risk: 84 months High Risk: 24 months

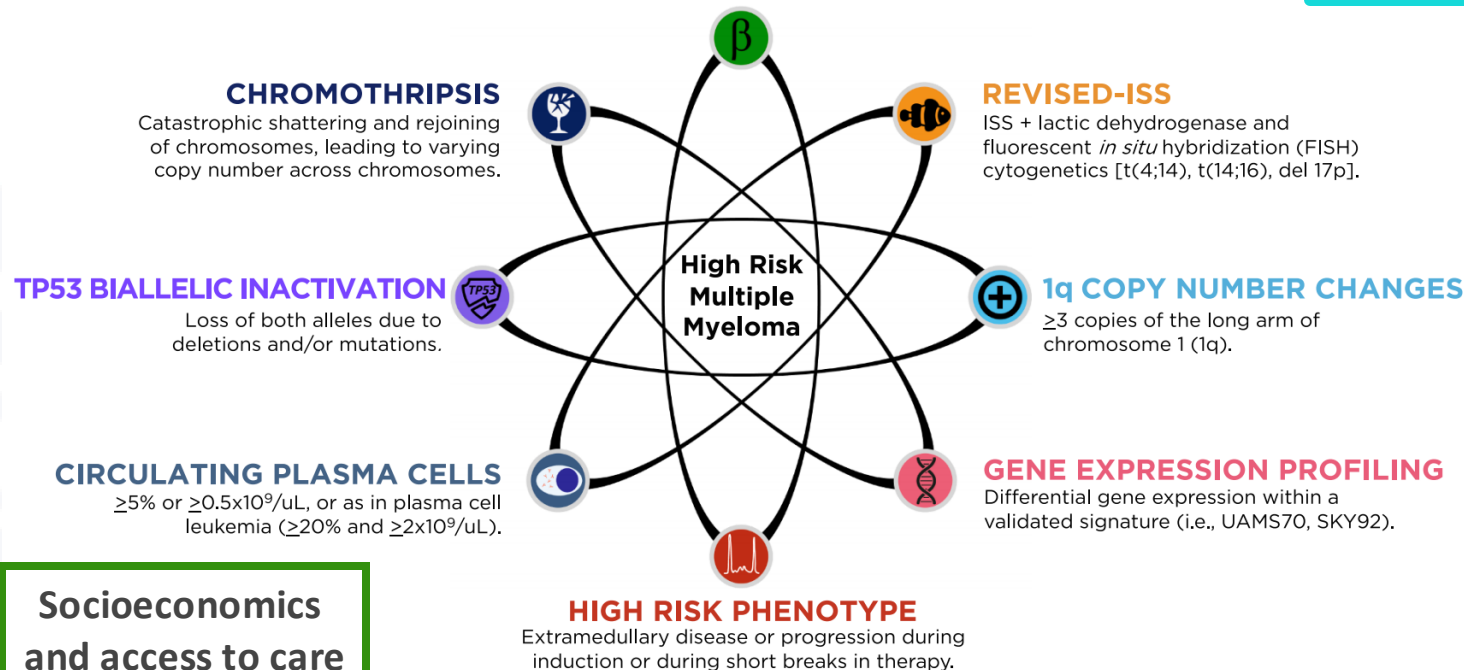
Multiple Myeloma CLINICS&BIOLOGY

Refining the definition of HIGH RISK patients

INTERNATIONAL STAGING SYSTEM (ISS)

Based on serum beta2-microglobulin and albumin.

Frailty



Multiple Myeloma CLINICS&BIOLOGY

The Genomed4All project: Multiple Myeloma use case



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EMILIA-ROMAGNA
Azienda Ospedaliero - Universitaria di Bologna

IRCCS Istituto di Ricovero e Cura a Carattere Scientifico

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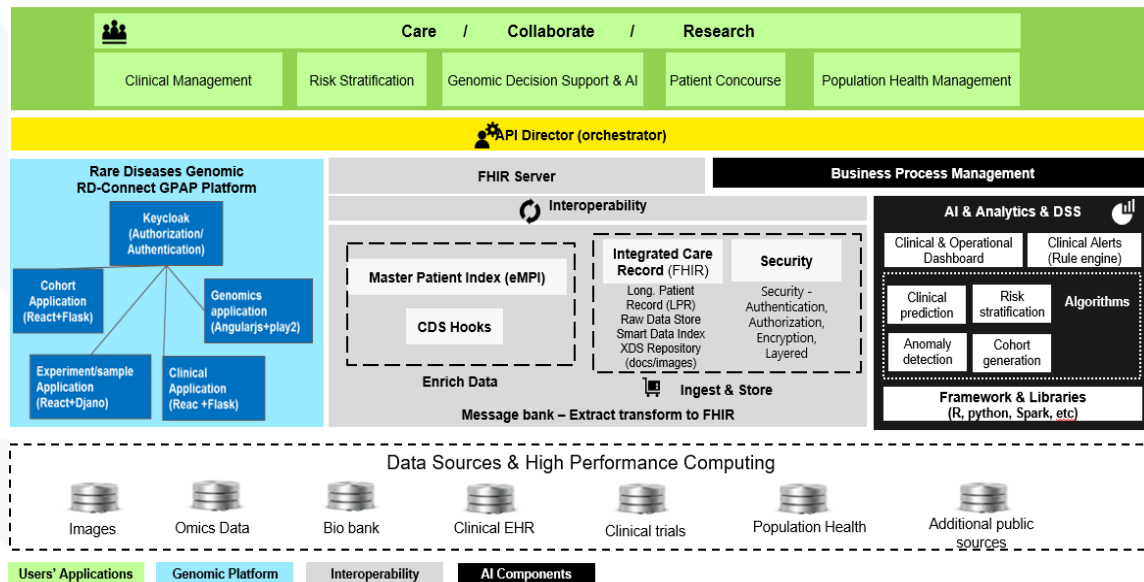
- 300 newly diagnosed MM patients enrolled by the Institute of Hematology «L. e A. Seràgnoli» - Bologna, Italy
- 995 newly diagnosed MM patients from MMRF CoMMpass Public repository

FINAL AIM: to refine the definition of HIGH RISK patients by possibly identifying predictors of EARLY relapse patients (within 18 months) based on the integration of baseline genomic, clinical & imaging data using artificial intelligence

Genomed4All project: AI for MM

Main components of the data sharing platform, data flows and connection to repositories

- Data harmonization
- Data sharing



Multiple Myeloma CLINICS&BIOLOGY

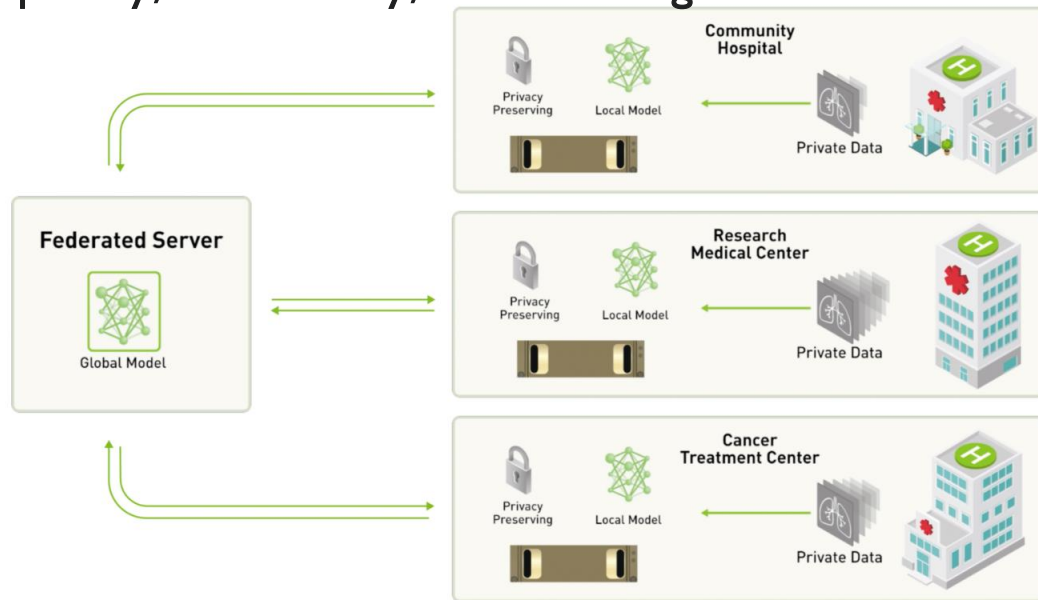
The Genomed4All projects' phases

1. FEDERATED LEARNING – UPM
2. DATA MODELLING and ML TRAINING ALGORITHMS FOR RISK PREDICTION – UNIBO, UNITO
3. RADIOMICS - UNIBO

Multiple Myeloma CLINICS&BIOLOGY

Federated Learning (also known as *collaborative learning*)

DEF: machine learning technique that trains an algorithm via multiple independent sessions, each using its own dataset without sharing data thus addressing critical issues such as data privacy, data security, data access rights and access to heterogeneous data



**DATASET DISCUSSION
FEATURES SELECTION**

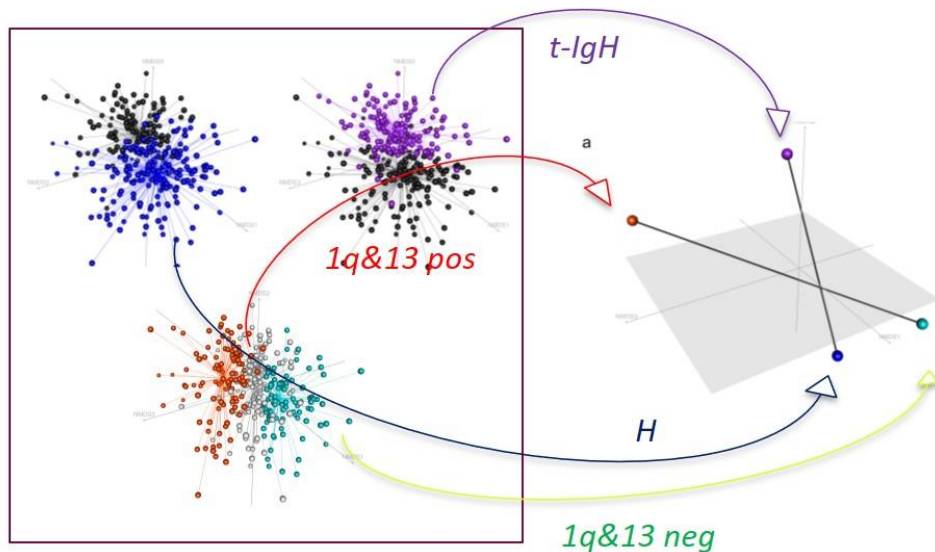
SYNTHETIC DATA GENERATION

SYNTHETIC DATA VALIDATION

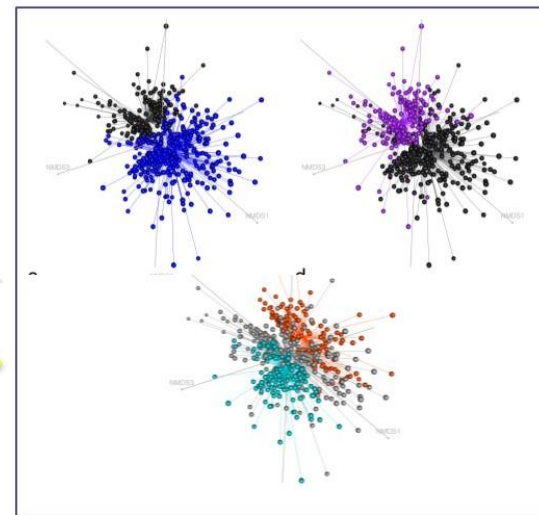
Multiple Myeloma CLINICS&BIOLOGY

Data modelling and Machine Learning algorithms

the organized chaos of Multiple Myeloma => the major component of MM heterogeneity



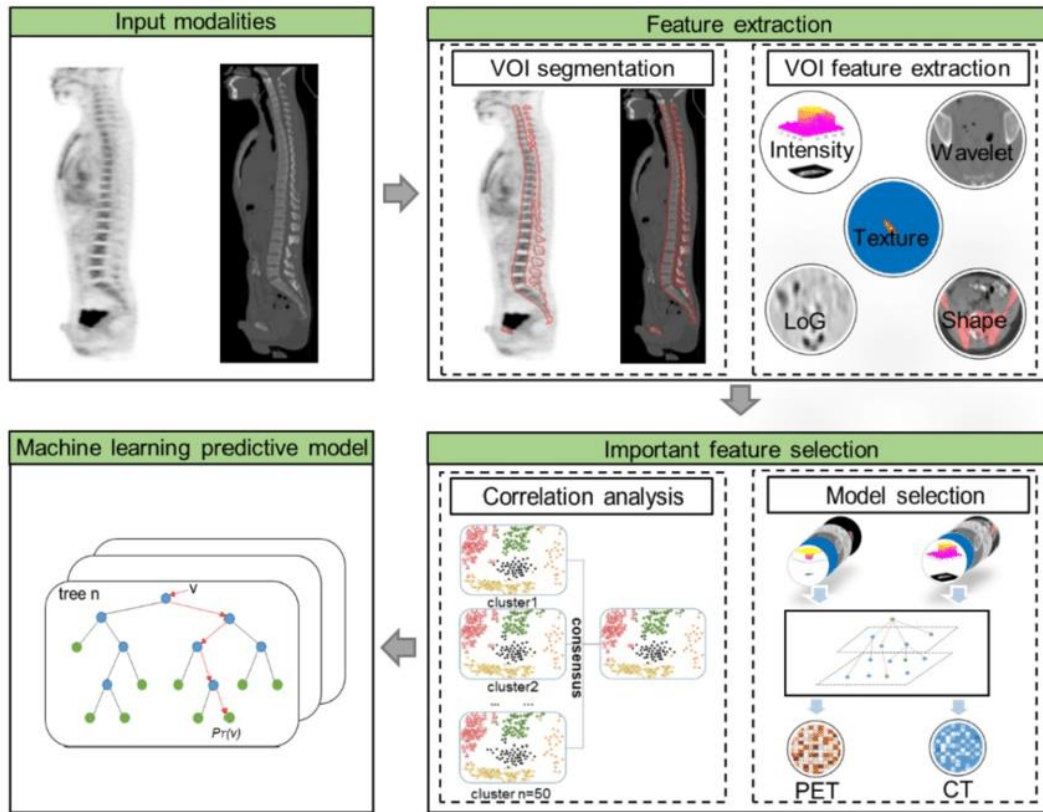
BO dataset



CoMMpass dataset

Multiple Myeloma CLINICS&BIOLOGY

Radiomics



Multiple Myeloma CLINICS&BIOLOGY

The Genomed4All project: Multiple Myeloma use case WRAP-UP

- ❑ Multiple Myeloma is a COMPLEX disease, but it needs a CURE
- ❑ Strong need to interconnect CLINICS & BIOLOGY to better predict PATIENTS' RISK
- ❑ A precise PREDICTION requires HUGE AMOUNT OF DATA to be statistically significant and clinically meaningful
- ❑ ARTIFICIAL INTELLIGENCE can help in data management, interconnection and prediction
- ❑ With our partners in the GENOMED4ALL project we aimed at the integration of both CLINICAL, GENOMIC and IMAGING of a large cohort of newly diagnosed MM patients in order to identify PREDICTORS OF EARLY RELAPSE that can refine the definition of high risk patients
- ❑ In the FUTURE, these results will contribute to an IMPROVED DESIGN of clinical trials and a BETTER PATIENTS' MANAGEMENT, towards a more PRECISE MEDICINE

Thanks!

Multiple Myeloma Research Unit

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Gastone Castellani

Santiago Zazo

Piero Fariselli

MOLECULAR BIOLOGY LAB

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Marina Martello

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Ilaria Vigliotta

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Ajsi Kanapari

CLINICAL RESEARCH UNIT

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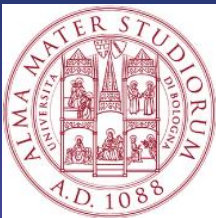
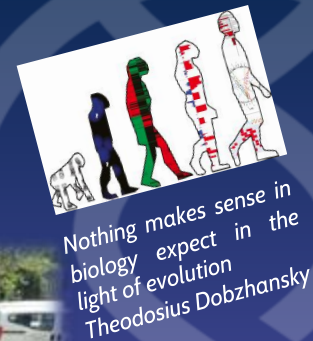
Paola Tacchetti

Lucia Pantani

Katia Mancuso

Ilaria Rizzello

Emanuele Favero



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Thanks!
Any questions?

GenoMed4All & ERN-EuroBloodNet

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Acknowledgements



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complex diseases



Network

Hematological
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