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Integrating Robust Machine Learning Classifiers to Automate Fluorescence Imaging-based *Ex Vivo* Drug Sensitivity Testing of Acute Lymphoblastic Leukemia

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Affiliations & Disclosures St. Jude Children's Research Hospital No Disclosures

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Brady S, et al. Nat Genetics 2022 & SEER 22 Data (2017-21)

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Population

Pediatric Acute Lymphoblastic Leukemia (ALL) is the most common type of childhood cancer representing ~25% of all new pediatric cancer diagnoses within the US.

Leukemia

Two main Types; B-ALL and T-ALL depending on which lymphocyte the disease originates from.

- 23 Subtypes for B-ALL
- 12 Subtypes for T-ALL

Pharmacotyping

Pharmacogenomics is the study of how genetic attributes affect drug response; <u>pharmacotyping</u> is defining a patient's leukemia blast phenotype *in vitro*.

Our pharmacotyping assay is reliant on fluorescent imaging and a Mesenchymal Stromal Cell (MSC) co-culture to evaluate a sample's morphology & sensitivity.



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Created with BioRender.com



Objective: Discerning leukemia cells from MSCs using nucleus area and intensity.











Same Well, Field of View, and Sample

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Input Image:











Finding cures. Saving children.

Re

Machine Learning Training

DATA SET INFO:

- Current model trained on 10 cases (DS0) (All cases passed QC)
- Objects selected from control wells [random], high concentration wells [random], low concentration wells [random], middle concentration wells [random].

Priority cell populations for model (n=100):

- Dead/Artifact
- MSCs
- Live ALL 1
- Live ALL 2

Viable ALL = Live ALL1 + Live ALL2 (Brighter ALL) (Dimmer ALL)







Differences in manual vs machine learning pipelines

Manual Thresholding

- Due to heterogeneity of samples, human input is required to determine the threshold based on intensity & area.
 - Potentially introduces bias.
 - Less consistent.
 - Classifies only 2 populations (MSC vs leukemia cell) therefore cannot label a cell as dead, etc.
 - Faster analysis time, due to less data.

Pre-trained Machine Learning

- Requires no human input for thresholding.
 - Binning of desired population to account for the variability among cases.
 - Uses Find Spots as its object identification block (as opposed to Find Nuclei/Cells).
 - Uses 200+ features to classify an object.
 - More populations for classifications reduces false positives.
 - Uses 3 different types of data collection blocks to make the classification.
 - Longer analysis time, due to more data.





Validation of Machine Learning Pipeline

Comparison of LC50s

- Manual vs. Machine Learning
- Red = Machine Learning
- Green = Manual



Correlation Plots

- Comparison of the imaging based LC50's between manual and automatic pipelines.
- Manual vs. Machine Learning







Validation of **Machine Learning Pipeline**

Comparison Plots

- R² = Correlation
- Black line $R^2 = 100\%$ • correlation
- Manual vs. Machine • Learning



Bland-Altman Plots

- Final validation comparing • the imaging based absolute percent viability differences.
- Manual (<0) vs. Machine • Learning (>0)





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Key Take Aways

- Use of machine learning classifiers shows capability of discerning leukemia cells from MSCs using 200+ features.
- Machine learning classifiers provide comparable results to the manual method while enhancing consistency across heterogeneous samples.
- Use of machine learning classifiers decreases the chances of ightarrowpotential bias by removing human dependency (for discerning populations).

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Pharmacotyping Team



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Things I didn't go over that people might be interested in

- What is the magnification of your images? Why did you decide that to be the magnification/FOV# for your. assay?
 - Our images are read at 20x, 9-FOVs, & Confocal. Prior testing had been performed before my arrival and these imaging settings were deemed acceptable.
- What Dyes are used for your analysis?
 - AOPI: Control wells
 - CQG: All wells minus AOPI
- When you mention texture, what features are extracted?
 - Mainly SER features such as: Hole, Edge, Ridge, etc.
- Do you use Brightfield channel for any of your data collection blocks?
 - No, we do use this method in other pipelines, and although brightfield can provide a lot of insight on the morphology and texture of a desired cell culture, the MSC co-culture set-up can cause challenges with data collection.
- What type of samples were used for training?

- Would thresholding texture properties make the manual pipeline more accurate?
 - Potentially yes. On the other hand, you are adding a level of complexity for the user to find that goldilocks zone of classification. In return you are increasing data size, increasing analysis time, increasing dependency on thresholding.
- You mentioned your machine learning classifier was a SVM, what is that?
 - A support vector machine (SVM) is a supervised machine learning algorithm that classifies data by finding an optimal line or hyperplane that maximizes the distance between each class in an N-dimensional space.
- How much time is required for a full report on a patient?
 - Generally, 7 days, 4 for incubation/imaging and 3 for quality control and analysis.
- What does SER and STAR mean in your data collection blocks?
 - Spots, Edges, Ridges + Symmetry Threshold, **Axial**, Radial



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• PDX, Patient, Pediatric samples

