Classification of Human Leukemias

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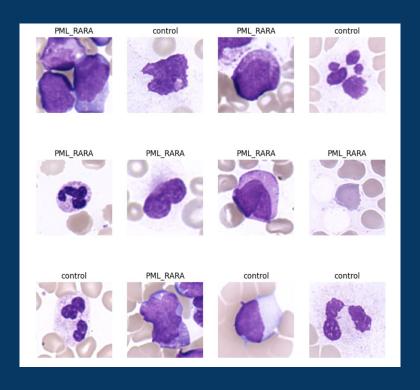
Cancer Image Archive

Blood Cell Images from Normal and Leukemia patients

4 Classes of Leukemia:

- 1. **Normal** (20,305 images)
- 2. Leukemia with PML::RARA fusion (11,584 images)
- 3. Leukemia with NPM1 mutation (17,710 images)
- 4. Leukemia with CBFB::MYH11 fusion (17,212 images)
- 5. Leukemia with RUNX1::RUNX1T1 fusion (14,403 images)

Normal (control) vs PML::RARA Leukemia



Normal (control) vs PML::RARA Leukemia

Normal (control) vs Leukemia with RUNX1 fusion

Input (64)

Conv2D (128)
Batch Normalization (128)
Max Pooling (128)

Conv2D (192) Batch Normalization (192) Max Pooling (192)

Learning Rate: 0.00001

Learning Rate: 0.0001

Input (64

Conv2D (256) Batch Normalization (256) Max Pooling (256)

Conv2D (192)
Batch Normalization (192)
Max Pooling (192)

Learning Rate: 0.00001

Flatten

Dense layer (128)

Dense layer (1)

Output

Flatten

Dense layer (256)

Dense layer (1)

Output

Normal (control) vs Leukemia with CBFB-MYH11 fusion

Input (32)

Conv2D (128)
Batch Normalization (128)
Max Pooling (128)

Conv2D (256) Batch Normalization (256) Max Pooling (256)

Conv2D (256) Batch Normalization (256) Max Pooling (256)

Flatten

Dense layer (192)

Dense layer (1)

Output

Normal (control) vs Leukemia with NPM1 Mutation

Input (64)

Conv2D (256) Batch Normalization (256) Max Pooling (256)

Conv2D (192)
Batch Normalization (192)
Max Pooling (192)

Learning Rate: 0.0001

Conv2D (192) Batch Normalization (192) Max Pooling (192)

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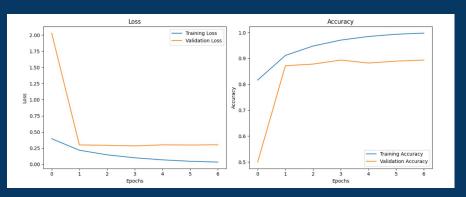
Flatten

Dense layer (256)

Dense layer (1)

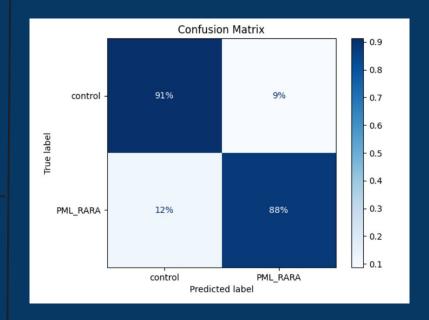
Output

Normal (control) vs PML::RARA Leukemia



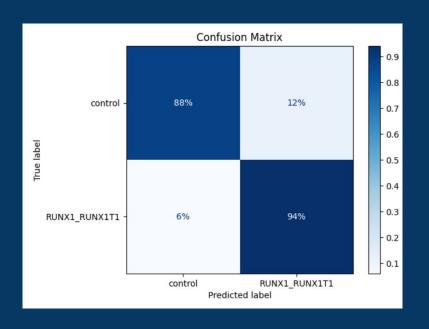
Training Loss (Post hyperparameter tuning)

Predictions	Precision	Recall	f1-Score	Support	
Control	0.88	0.91	0.9	800	
PML_RARA	0.91	0.88	0.9	800	
Accuracy			0.9	1600	
Macro Avg	0.9	0.9	0.9	1600	
Weighted Avg	0.9	0.9	0.9	1600	



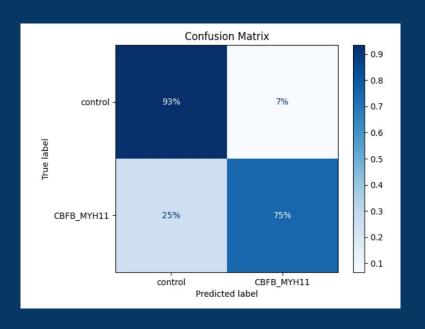
Normal (control) vs Leukemia with RUNX1 fusion

Predictions	Precision	Recall	f1-Score	Support
Control	0.94	0.88	0.91	800
RUNX1_RUNX1T1	0.89	0.94	0.91	800
Accuracy			0.91	1600
Macro Avg	0.91	0.91	0.91	1600
Weighted Avg	0.91	0.91	0.91	1600



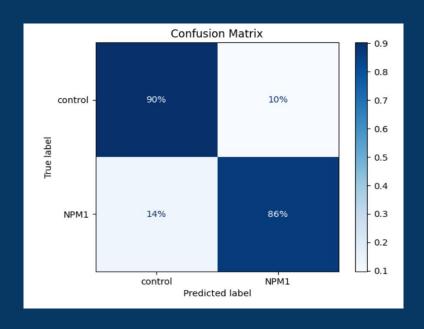
Normal (control) vs Leukemia with CBFB-MYH11 fusion

Predictions	Precision	Recall	f1-Score	Support
Control	0.79	0.93	0.86	800
CBFB-MYH11	0.92	0.75	0.83	800
Accuracy			0.84	1600
Macro Avg	0.85	0.84	0.84	1600
Weighted Avg	0.85	0.84	0.84	1600



Normal (control) vs Leukemia with NPM1 Mutation

Predictions	Precision	Recall	f1-Score	Support
Control	0.87	0.90	0.89	800
NPM1	0.90	0.86	0.88	800
Accuracy			0.88	1600
Macro Avg	0.88	0.88	0.88	1600
Weighted Avg	0.88	0.88	0.88	1600



Conclusion

Established 4 Image Classification models for for 4 different types of Human Leukemias.

F1 score of each model ~0.8 - 0.9.

Conclusion

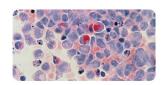
Each model was trained with ~5000 images from normal individual and from Leukemia patients.

Dataset and Notebooks are NOW available on Kaggle

https://www.kaggle.com/datasets/gchan357/human-aml-cytomorphology-dataset

Human Leukemia Cytomorphology Dataset

Dataset of Blood Cell Images from normal patients and patients with Leukemia



Data Card Code (5) Discussion (0) Suggestions (0)

About Dataset

This dataset comprises blood cell images from four prevalent Acute Myeloid Leukemia (AML) subtypes with defining genetic abnormalities and typical morphological features according to the WHO 2022 classification: (i) APL with PML::RARA fusion, (ii) AML with NPM1 mutation, (iii) AML with CBFB::MYH11 fusion (without NPM1 mutation), and (iv) AML with RUNX1::RUNX1T1 fusion, as well as a control group of healthy stem cell donors.

Each folder under the root directory represent one classs. These are:

- (1) normal patients (control)
- (2) APL with PML::RARA fusion (PML_RARA)
- (3) AML with NPM1 mutation (NPM1)
- (4) AML with CBFB::MYH11 fusion without NPM1 mutation (CBFB_MYH11)
- (5) AML with RUNX1::RUNX1T1 fusion (RUNX1_RUNX1T1).

Each sub-folder under the folder of each class represent images from a single patient with that Leukemia. Each patient's initial is the name of that folder.

Usability © 7.50

License

Other (specified in description)

Expected update frequency

Never Tags

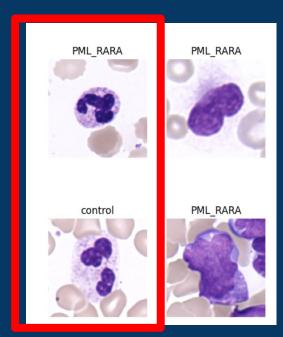
Cancer

Image Classification

Diseases

Leukemia patients sample may contain normal white blood cells.

- May in part explain misclassification?
- Datasets needs to be curated?



World Health Organisation (WHO) ²³⁶	Incidence	FAB type ^{237,238}	Incidenc
AML with recurrent genetic abnormalities t(8;21)(q22;q22) inv(16)(p13q22) or t(16;16)(p13;q22) t(15;17)(q22;q12), (PML/RARα) and variants 11q23 (MLL) abnormalities	30-40%	M2 M4eo M3	25-30% 5-15% 5-10%
2 AML with multilineage dysplasia Following MDS or MDS/MPD Without antecedent MDS or MPD, with dysplasia in at 50% of cells in 2 or more myeloid lineages	10-15% least		
3 AML and MDS, therapy related Alkylating agents/radiation-related Topoisomerase 2 inhibitor-related Others	5-10%		
AML, not otherwise categorised Minimally differentiated Without maturation With maturation Acute myelomonocytic leukemia Acute monoblastic/monocytic leukemia Acute erythroid leukemia Acute megakaryblastic leukemia Acute basophilic leukemia Acute panmyelosis with myelofibrosis Myeloid sarcoma	30-40%	M1 M0 M4 M5 M6 M7	10-15% 2-5% 5-15% 15-25% -3% 5-10%
MDS indicates myelodysplastic syndrome; MPD myeloproliferative	e syndrome		

Models presented here may be used as a base model to train classification of other Leukemias using Transfer Learning.