### Complexity of ballooned hepatocyte feature recognition: Defining a training atlas for artificial intelligence-based imaging in NAFLD

Liver Forum, April 22, 2022

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## Complexity of ballooned hepatocyte feature recognition: Defining a training atlas for artificial intelligence-based imaging in NAFLD

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Advisory Board: Alnylam/Regeneron, Pfizer.

Consultancy agreements: Arrowhead, Cymabay, Histoindex, Intercept, Medpace, NGM, Perspectum Diagnostics.

## Special thanks to

- The amazing pathologists who did this work
- Quentin Anstee: stats and intellectual (hepatology) input
- Dean Tai: stats and intellectual (engineer) input
  - Both for sharing some of these slides
- Clinical hepatology colleagues

### **Current Concept**

Necessary for dx of **NASH**  NOT NASH NASH Resolution

### Ballooning 1-2

**Ballooning 0** 



Steatosis, Inflammation, +/- Fibrosis

Steatosis, Inflammation, +/- Fibrosis

# Background

- Increase in clinical trials in NASH puestions:
  - Reproducibility of pathologists' interpretations on "routine stains"
    - Ballooning
    - Diagnosis: NASH
    - Features for NASH Resolution
  - Role of Machine learning/AI for "consistency"
    - ? Replace pathologists
    - ? Guide pathologists

9 International Expert Hepatopathologists Independent Annotation of All Ballooned Cells in 10 Digitised Liver Biopsy Images



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Establish a 'Concordance Atlas' of Annotated Ballooned Cell Images Training of SHG/TPE AI Algorithm





## M/M

- ROI of digitized slides chosen (by EMB) to
  - Equalize amount of liver tissue/slide for each bx
  - Represent the <u>spectrum of NAFLD</u>; steatosis (**B0**) active steatohepatitis (**B1-B2**)
  - Represent the spectrum of <u>slide quality</u> as seen in clinical trials
- Agreement statistics: 3 binary conditions
  - Presence of ANY ballooned hepatocytes
  - Presence of <a>> 5 ballooned hepatocytes</a>
  - Non-NASH v NASH diagnosis

#### Inter-Observer Concordance between Pathologists for Number of Ballooned Cells Identified.

Range of ballooned hepatocytes observed after rotation of slides by pathologists: **32%-91%** 

"NO BALLOONING" -2 pathologists agreed B0: #4, #7 -5 pathologists agreed B0: #8, #9 -6 pathologists agreed B0: #6

-At least one pathologist recorded B0 for all slides except #3, #5

-All pathologists except two (H, I) recorded B0 at least once



**B0** = no ballooning

Intraclass correlation coefficient for consistency: 0.718 (0.511-0.900): moderate for <u>ballooning burden</u>

#### Inter-Observer Concordance between Pathologists for Ballooned Cells Identified.



Mean: 173

			Comparator Pathologist									
		Sum	А	В	С	D	E	F	G	Н	I.	
Reference Pathologist	А	138	-	89 (64%)	81 (59%)	63 (46%)	73 (53%)	97 (70%)	45 (33%)	21 (15%)	52 (38%)	
	В	322	89 (28%)	-	171 (53%)	109 (34%)	142 (44%)	196 (61%)	82 (25%)	56 (17%)	65 (20%)	
	С	535	81 (15%)	171 (32%)	-	126 (24%)	169 (32%)	197 (37%)	84 (16%)	55 (10%)	85 (16%)	
	D	173	63 (36%)	109 (63%)	126 (73%)	-	130 (75%)	106 (61%)	68 (39%)	38 (22%)	52 (30%)	
	E	287	73 (25%)	142 (49%)	169 (59%)	130 (45%)	-	151 (53%)	87 (30%)	55 (19%)	59 (21%)	
	F	596	97 (16%)	196 (33%)	197 (33%)	106 (18%)	151 (25%)	-	79 (13%)	47 (08%)	71 (12%)	
	G	119	45 (38%)	82 (69%)	84 (71%)	68 (57%)	87 (73%)	79 (66%)	-	35 (29%)	37 (31%)	
	Н	122	21 (17%)	56 (46%)	55 (45%)	38 (31%)	55 (45%)	47 (39%)	35 (29%)	-	25 (20%)	
	I	160	52 (33%)	65 (41%)	85 (53%)	52 (33%)	59 (37%)	71 (44%)	37 (23%)	25 (16%)	-	

Heatmap: green is high agreement of comparator to reference pathologist for # of ballooned cells; red is low

Variation: 8%-75%; <u>kappa: 0.197</u>; <u>0.395 for >5 cells</u>

> 8,000 cells evaluated (by counting nuclei); 1188 ballooned cells circled.





The **ONLY** ballooned 9 agreed

#### Inter-Observer Concordance between Pathologists for Number of Ballooned Cells Identified: Trends Identified

Item Statistics							
	Mean	Std. Deviation	N				
Α	13.80	17.184	10				
В	32.20	51.873	10				
С	53.50	66.185	10				
D	17.30	36.237	10				
Е	28.70	68.668	10				
F	59.60	67.160	10				
G	11.90	27.534	10				
Н	12.20	26.452	10				
	16.00	17.969	10				

Propensity to "see" ballooning:

More	Fewer
F > C	G <i>,</i> H

#### Inter-Observer Concordance between Pathologists for Number of Ballooned Cells Identified: Trends Identified



Log2 transformed graph showing Median-IQR Robust Scaled Cell Count. Longer lines imply greater deviation by given pathologist from group median for each slide. Dotted lines 1 SD.

**Item Statistics** Mean Std. Deviation N 13.80 17.184 Α 10 51.873 10 32.20 B C 53.50 66.185 10 17.30 36.237 10 D 28.70 68.668 10 E 59.60 67.160 10 F 11.90 27.534 10 G 12.20 26.452 10 Н 16.00 17.969 10

Propensity to "see" ballooning:MoreFewerF > CG, H

#### Ballooned hepatocyte diameter by Pathologist: Is this the key to the differences?



### SQBS: semi-quantitative balloon score created

- 0-2 to align with NAS and SAF scores
- Transformed ballooned hepatocyte count/image/pathologist
  - 0 < 5
  - 1 = 5-75
  - 2 >75
- Cut-off b/w 1-2 derived from overall mean +1SD of ballooned cells reported per slide

Trend of Semi-Quantitative Ballooning Score (0-2) by Slide and Pathologist



Trend of Semi-Quantitative Ballooning Score (0-2) by Slide and Pathologist



Overall "fair" agreement kappa 0.29 (95% CI 0.210-0.371); pairwise observer kappa 0.231-1.000.

Trend of Semi-Quantitative Ballooning Score (0-2) by Slide and Pathologist





#### Comparison of 'non-NASH NAFL' vs. 'NASH' diagnostic call by Pathologist and Image: Correlation with ballooning.

Blue = no ballooned hepatocytes Red = many ballooned hepatocytes

			Digital Image #									
		Call	1	2	3	4	5	6	7	8	9	10
	А	1/10	NASH	NASH	NASH	NASH	NASH	NASH	NASH	Not NASH	NASH	NASH
	В	3/10	Not NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH
	С	2/10	NASH	NASH	NASH	NASH	NASH	Not NASH	NASH	Not NASH	Not NASH	Not NASH
<b>j</b> ist	D	2/10	Not NASH	NASH	NASH	NASH	NASH	Not NASH	NASH	Not NASH	Not NASH	NASH
tholog	E	1/10	NASH	Not NASH	NASH	NASH	NASH	NASH	NASH	Not NASH	Not NASH	NASH
Ра	F	1/10	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH	Not NASH	NASH
	G	2/10	NASH	NASH	NASH	NASH	NASH	Not NASH	Not NASH	Not NASH	Not NASH	NASH
	Н	7/10	Not NASH	Not NASH	Not NASH	Not NASH	NASH	Not NASH				
	I	2/10	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH
	Modal '	Consensus'	6/9	7/9	8/9	7/9	9/9	5/9	7/9	6/9	6/9	7/9
		NASH	NASH	NASH	NASH	NASH	NASH	NASH	Not NASH	Not NASH	NASH	

Comparison of 'non-NASH NAFL' vs. 'NASH' diagnostic call by Pathologist and Image: Correlation with ballooning.

			Minority Call	1	2	3	4	Digital 5	Image # 6	7	8	9	10
Pathologist		А	1/10	NASH	NASH	NASH	NASH	NASH	NASH	NASH	Not NASH	NASH	NASH
		В	3/10	Not NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH
		С	2/10	NASH	NASH	NASH	NASH	NASH	Not NASH	NASH	Not NASH	Not NASH	Not NASH
	jist	D	2/10	Not NASH	NASH	NASH	NASH	NASH	Not NASH	NASH	Not NASH	Not NASH	NASH
	tholog	E	1/10	NASH	Not NASH	NASH	NASH	NASH	NASH	NASH	Not NASH	Not NASH	NASH
	Ра	F	1/10	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH	Not NASH	NASH
		G	2/10	NASH	NASH	NASH	NASH	NASH	Not NASH	Not NASH	Not NASH	Not NASH	NASH
		Н	7/10	Not NASH	Not NASH	Not NASH	Not NASH	NASH	Not NASH	Not NASH	Not NASH	Not NASH	Not NASH
		I	2/10	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH	NASH
		Modal '	Consensus'	6/9 NASH	7/9 NASH	8/9 NASH	7/9 NASH	9/9 NASH	5/9 NASH	7/9 NASH	6/9 Not NASH	6/9 Not NASH	7/9 NASH
			Iodal QBS Score:	0	1	1	1	2	<b>0</b> J Hepatol	<b>0</b> doi: 10.1016/	<b>0</b> .he.2022/01.(	<b>0</b> 011. PMID: 3	<b>1</b> 5090960.

#### Is there a "strong" correlation of presence of ballooning with diagnosis of NASH by pathologists?

Is there a way to improve/ standardize interpretation of ballooning in NASH?



Courtesy V Desmet

### SHG/TPEF: on Unstained Slides: Machine Learning...Al



#### Quantitative, objective and repeatable data with reliable continuous readouts

Algorithm: Out of 45 parameters, 7 were chosen: 1 is fibrosis related; 6 are ballooning related

Pathologists' annotation

qBallooning algorithm



qBallooning2 training-set cell-selection criteria	Number of ballooned cells Identified by Pathologists	Number of ballooned cells Identified by qBallooning2	Overlap between qBallooning2 and majority concordance of 25- Pathologists	Positive Predictive Value Proportion of ballooned cells called by qBallooning2 are 'True Positive' *	False Discovery Rate Proportion of ballooned cells called by qBallooning2 are 'False Positive' *	True Positive Rate (Sensitivity) Proportion of ballooned cells identified by qBallooning2	False Negative Rate Proportion of ballooned cells missed by qBallooning2
Agreement of any 1 pathologist	1188	346	54	54/346 (16%)	292/346 (84%)	54/133 (41%)	79/133 (59%)
Agreement of at least 2 pathologists	481	250	51	51/250 (20%)	199/250 (79.6%)	51/133 (38%)	82/133 (62%)
Agreement of at least 3 pathologists	284	170	37	37/170 (22%)	133/170 (78.2%)	37/133 (28%)	96/133 (72%)
Agreement of at least 4 pathologists	188	114	25	25/114 (22%)	89/114 (78%)	25/133 {19%}	108/133 (81%)
Agreement of at least 5 patholopists	133	88	22	22/88 (25%)	66/88 (75%)	22/133 (17%)	111/133 (83%)
Agreement of at least 6 pathologists	86	59	16	16/59 (27%)	43/59 (73%)	16/133 (12%)	117/133 (88%)
Agreement of at least 7 pathologists	59	40	15	15/40 (38%)	25/40 (62.5%)	15/133 (11%)	118/133 (89%)
Agreement of at least 8 pathologists	26	24	5	5/24 (21%)	19/24 (79%)	5/133 (4%)	128/133 (96%)

- qballooning2 was compared with every pathologist to fine tune the algorithm
- had pairwise overlap with individual pathologists ranging from 19% (with Pathologist F) to 42% (with Pathologist G), which was comparable to the level of inter-observer variation between pathologists of 8-75%
- For studies, a simple majority (>5) was chosen

Table comparing the performance of qBallooning2 in the development dataset. The algorithm was optimized to detect ballooned cells using data derived from each level of interobserver concordance and shows how the level of interobserver concordance stipulated affects the performance of the algorithm. \* Relative to majority concordance of ≥5-pathologists.

### 'Ground Truth' Atlas

qBallooning2 training-set cell-selection criteria	Number of ballooned cells Identified by Pathologist s
Agreement of any 1 pathologist	1188
Agreement of any 2 pathologists	481
Agreement of any 3 pathologists	284
Agreement of any 4 pathologists	188
Agreement of any 5 pathologists	133
Agreement of any 6 pathologists	86
Agreement of any 7 pathologists	59
Agreement of any 8+ pathologists	26



\* Relative to majority consensus of ≥5-pathologists. § Based on an estimated me

**Proof of Principle: Can the algorithm/machine detect change?** 

Quantification data showing the relative change in the number of ballooned hepatocytes and the

qBallooning index for patients with and without ballooning reduction.



### Conclusions -1-

• Hepatocyte ballooning is recognized as THE key to distinguishing NASH from other forms of NAFLD within the spectrum

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### Conclusions -1-

- Hepatocyte ballooning is recognized as THE key to distinguishing NASH from other forms of NAFLD within the spectrum
- Loss of ballooning (BO) has been identified as necessary, along with decrease in inflammation and no worsening of fibrosis, for "NASH resolution" in clinical trial assessment (FDA, EMA)
- Pathologists define ballooning similarly...but do we see it similarly in slides?

### Conclusions -2-

- Broad divergence in assessment of hepatocyte ballooning amongst expert pathologists
  - Not based on level of training or geographic location
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- Broad divergence in assessment of hepatocyte ballooning amongst expert pathologists
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  - Suggests there are <u>nuances</u> in ballooned hepatocytes differently appreciated by pathologists
- Implies that requiring ballooning score 0 for endpoint efficacy in trials is quite possibly unrealistic
  - It may be more realistic to look for trends in ballooning burden between pre and post intervention biopsies
- Training of AI/machine learning by concordance atlas is doable and can give appropriate PPV; use of such may bring a more standardized means of assessing efficacy in clinical trials



