A Deep Learning Approach to Analysis of MRCP Images Predicts Clinical Events and Progression to Cirrhosis in Patients With Primary Sclerosing Cholangitis

⁹University of Alberta, Edmonton, Canada; ¹⁰University of California Davis; ¹⁴Duke Clinical Research Institute, Durham, North Carolina, USA; ¹⁵Brigham and Women's Hospital, Harvard Medical School, Boston



Aaditya Prakash,¹ Hunter Elliot,¹ Michael C. Montalto,¹ Andrew H. Beck,¹ Murray Resnick,¹ Ilan Wapinski,¹ Oscar M. Carrasco-Zevallos,¹ Xiaomin Lu,² Xiangyu Liu,² Chuhan Chung,² Robert P. Myers,² Michael Manns,³ Stephen H. Caldwell,⁴ Raj Vuppalanchi,⁵ K. Rajender Reddy,⁶ Lisa Forman,⊓ Mitchell L. Shiffman,³ Aldo Montano-Loza,⁰ Christopher L. Bowlus,¹⁰ Cynthia Levy,¹¹ Kris V. Kowdley,¹² Michael Trauner,¹³ Andrew J. Muir,¹⁴ Clare Tempany-Afdhal¹⁵

¹PathAl. Inc., Boston, Massachusetts, USA; ²Gilead Sciences, Inc., Foster City, California, USA; ³Medizinische Hochschule Hannover, Germany; ⁴University of Virginia; Charlottesville, USA; ⁵University of Pennsylvania, Philadelphia, USA; ¹University of Colorado Anschutz Medical Campus, Aurora, USA; ³Bon Secours Mercy Health, Richmond, Virginia;

Gilead Sciences, In 333 Lakeside Driv oster City, California, USA 9440 800-445-323

Introduction

- Primary sclerosing cholangitis (PSC) is a chronic heterogeneous cholangiopathy and prognostic scores have modest ability to predict clinical outcomes^{1,2}
- Magnetic resonance cholangiopancreatography (MRCP) is the primary method of PSC diagnosis, but its prognostic utility remains unclear^{3,4}

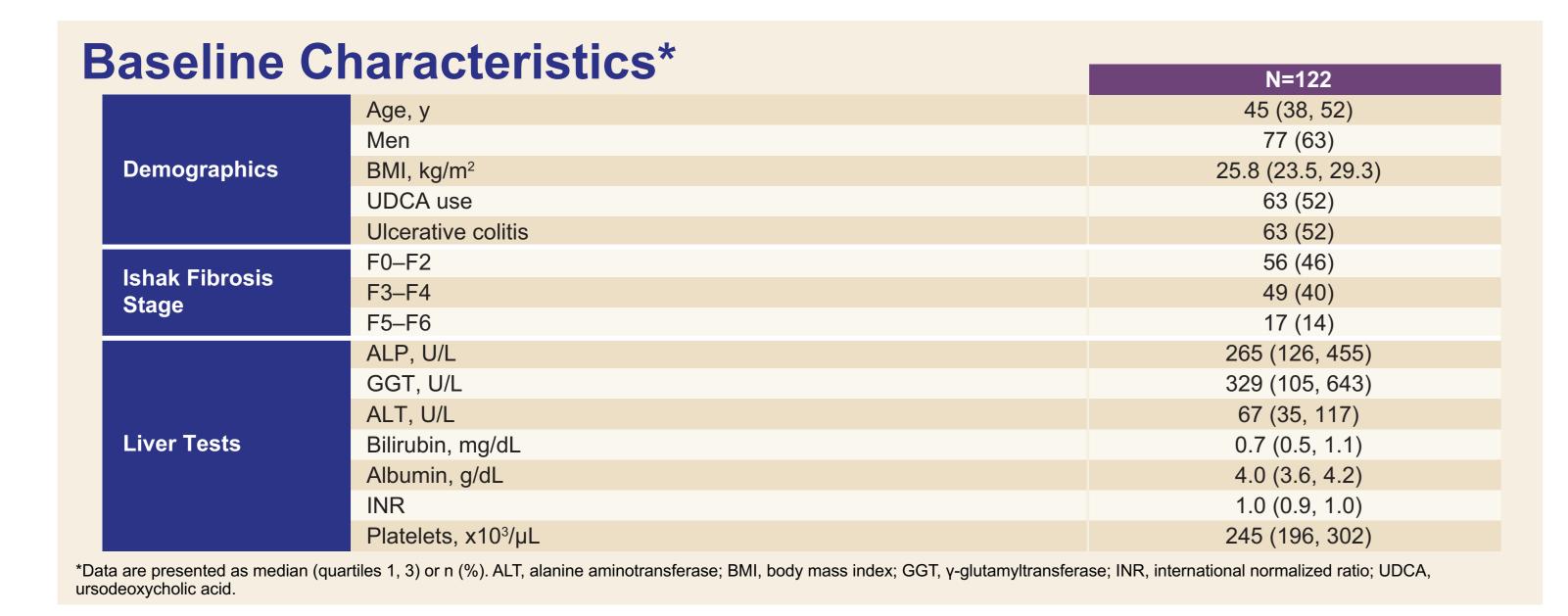
Objectives

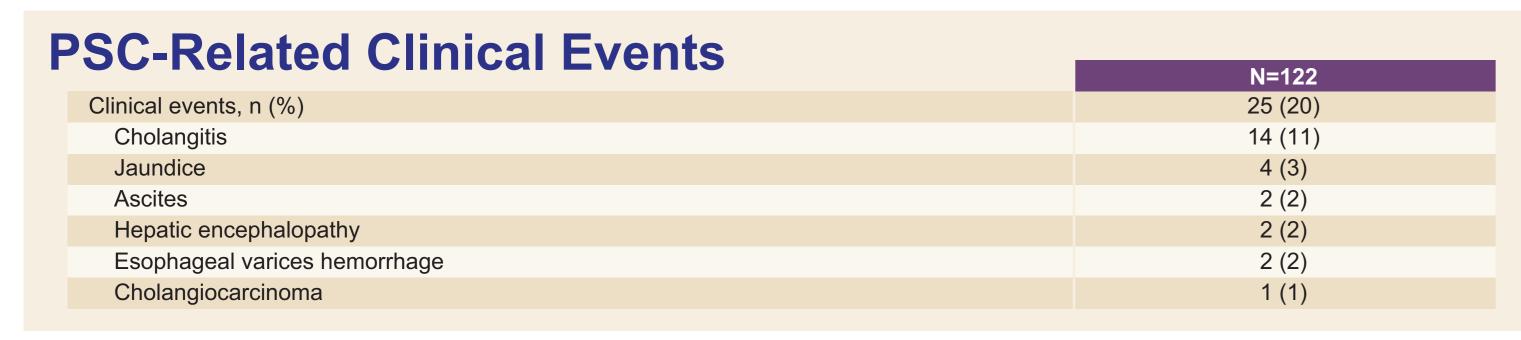
◆ To develop a machine learning (ML) algorithm based on MRCP images and evaluate its ability to predict clinical outcomes in patients with PSC enrolled in a clinical trial

Methods

- ◆ Baseline MRCP and liver biopsy images were available from patients (N=122) with compensated PSC enrolled in a 96-wk, Phase 2b clinical trial of simtuzumab (ClinicalTrials.gov NCT01672853)⁵
- ◆ A state-of-the-art, multilayer, convolution neural network (CNN) pretrained on ImageNet images was applied to the MRCP volumes⁶
- Activation maps generated from MRCP slices to a CNN were trained and crossvalidated on a binary outcome of PSC-related clinical events (eg, hepatic decompensation, ascending cholangitis, and cholangiocarcinoma)
- The activation maps were input into a random forest classifier to generate ML MRCP scores; these scores were used for predicting progression to cirrhosis or development of PSC-related clinical events
- Discrimination of the resultant ML MRCP scores at baseline (range 0–1) for PSC-related clinical events and histologic progression to cirrhosis was determined using areas under receiver operating characteristic curves (AUROCs) and compared with other clinical parameters, including a semiquantitative MRCP risk score,⁷ Ishak fibrosis stage, collagen proportionate area (CPA), serum alkaline phosphatase (ALP), Enhanced Liver Fibrosis (ELF™; Siemens Healthcare GmbH, Erlangen, Germany) score, Mayo risk score, PSC risk estimate tool (PREsTo) algorithm, and ML-based histologic features (PathAI, Inc., Boston, Massachusetts, USA)⁸

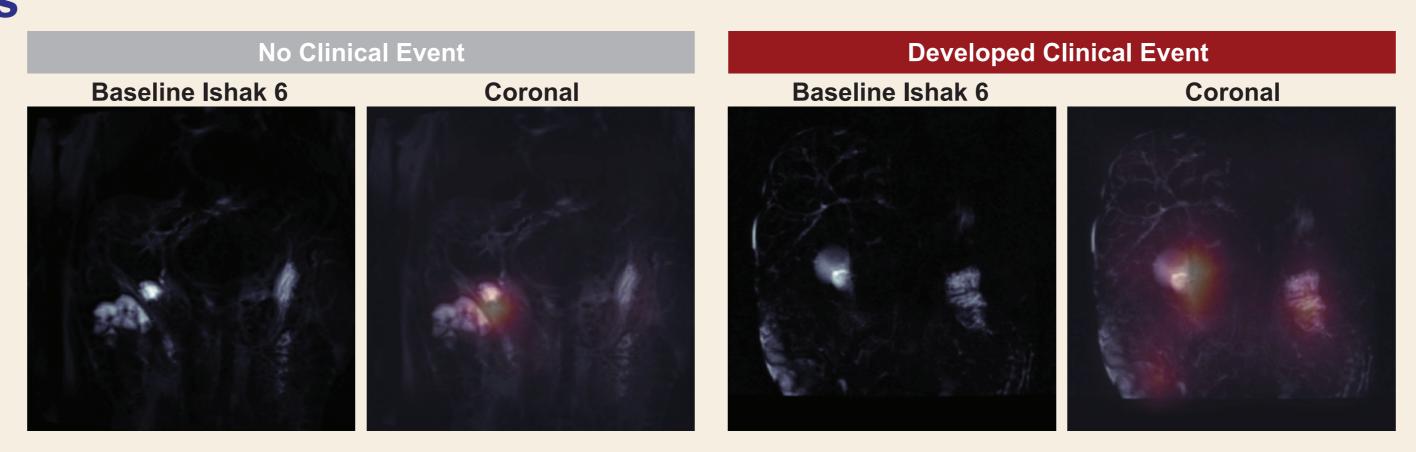
Results



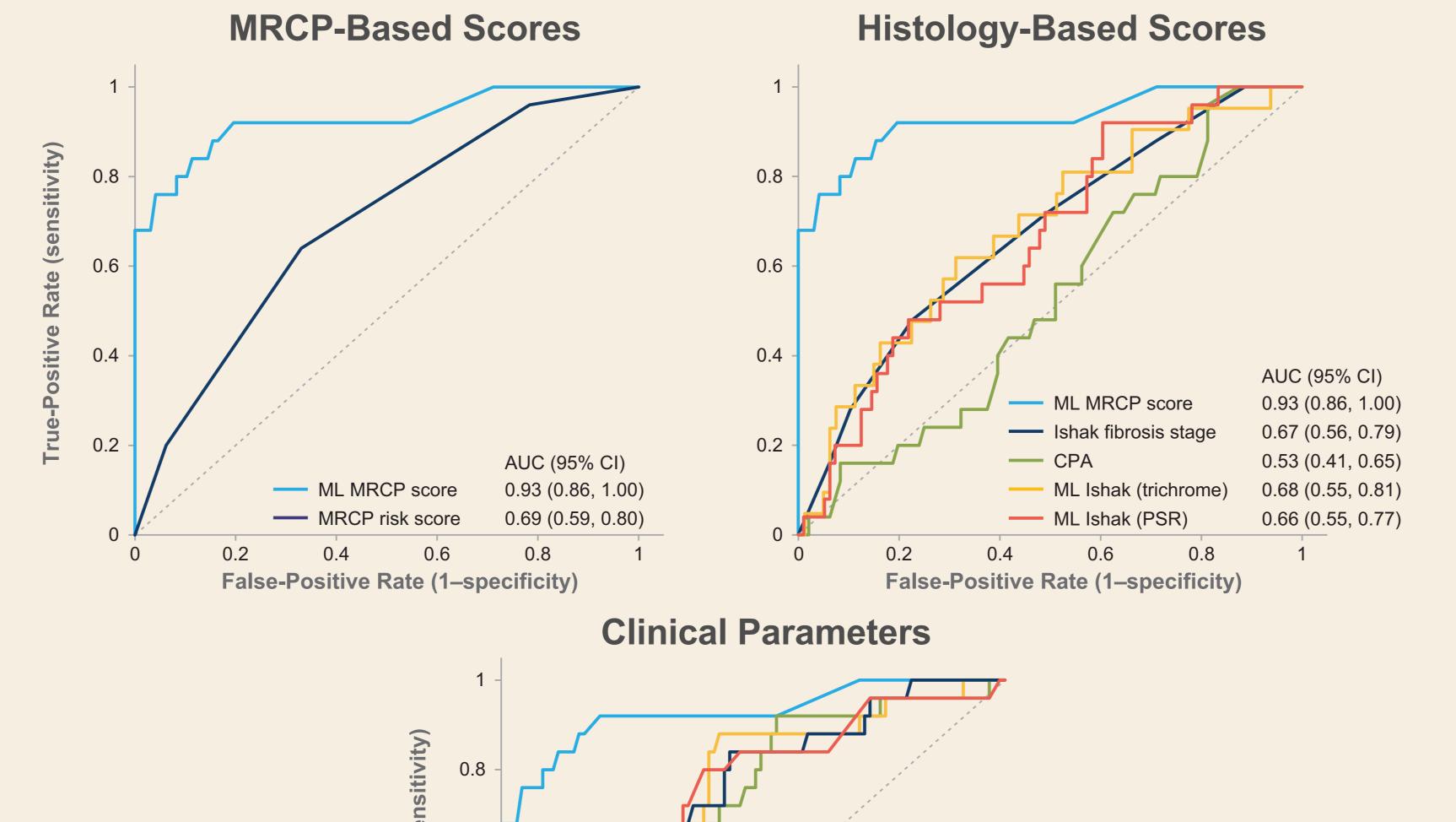


- During a median follow-up of 23.0 months (range 0.5–24.7), 25/122 patients (20%) had a PSC-related clinical event
- ♦ In all, 18/105 patients (17%) progressed to cirrhosis

CNN-Generated "Activation Maps" of MRCP Images Trained on Clinical Events*



ML MRCP Score Was Superior to Clinical Parameters and Other Prognostic Models in Predicting PSC-Related Clinical Events



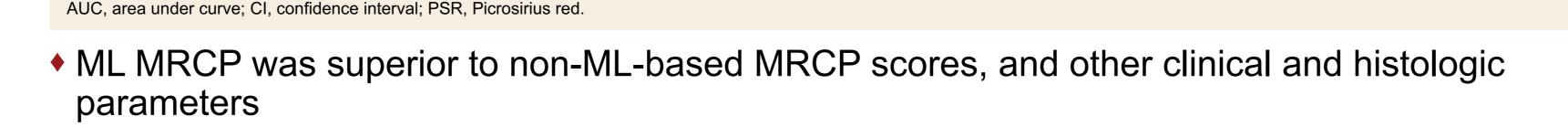
AUC (95% CI)

0.68 (0.57, 0.79)

0.69 (0.59, 0.80)

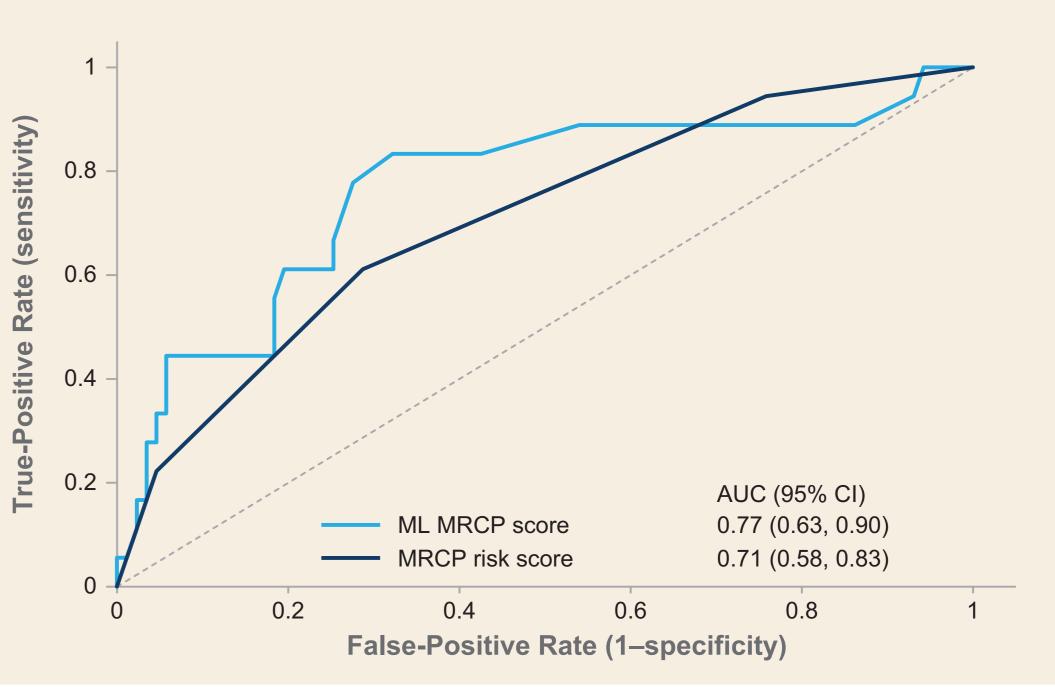
0.69 (0.58, 0.80)

0.70 (0.58, 0.81)



False-Positive Rate (1–specificity)

Discrimination of ML MRCP Score for Prediction of Progression to Cirrhosis



Discrimination of MRCP-Based Scores, Histologic Features, and Clinical Parameters for Clinical Events and Progression to Cirrhosis

PSC-Related Clinical Events Progression to Cirrhosis

		N=122		n=105	
	Parameter	AUROC (95% CI)	p-Value vs ML MRCP Score	AUROC (95% CI)	p-Value vs ML MRCP Score
MRCP-Based Scores	ML MRCP score	0.93 (0.86, 1.00)	_	0.77 (0.63, 0.90)	_
	MRCP risk score	0.69 (0.59, 0.80)	<0.001	0.71 (0.58, 0.83)	0.54
Fibrosis	Ishak fibrosis stage	0.67 (0.56, 0.79)	<0.001	0.71 (0.59, 0.84)	0.61
	CPA	0.53 (0.41, 0.65)	<0.001	0.68 (0.55, 0.82)	0.47
	ML Ishak score (trichrome)	0.68 (0.55, 0.81)	<0.001	0.85 (0.74, 0.96)	0.28
	ML Ishak score (PSR)	0.66 (0.55, 0.77)	<0.001	0.75 (0.63, 0.86)	0.85
Clinical Parameters	ALP	0.68 (0.57, 0.79)	<0.001	0.75 (0.66, 0.85)	0.90
	ELF score	0.69 (0.59, 0.80)	<0.001	0.78 (0.67, 0.90)	0.84
	Mayo risk score	0.69 (0.58, 0.80)	<0.001	0.69 (0.53, 0.84)	0.48
	PREsTo	0.70 (0.58, 0.81)	<0.001	0.69 (0.53, 0.84)	0.50

ML MRCP Score Correlated Poorly to MRCP Risk Score, Clinical Parameters, and Other Prognostic Scores at Baseline

		ML MRCP Score (trained for clinical event) N=122		
	Parameter	Correlation Coefficient	p-Value	
MRCP-Based Scores	MRCP risk score	-0.2025	0.025	
Fibrosis	Ishak fibrosis stage	-0.1348	0.14	
	% collagen morphometry	0.0252	0.78	
	ML Ishak score (trichrome)	-0.1806	0.07	
	ML Ishak score (PSR)	-0.1673	0.07	
Clinical Parameters	ALP	-0.1524	0.09	
	ELF score	-0.1848	0.042	
	Mayo risk score	-0.1288	0.16	
	PREsTo	-0.1902	0.036	

Conclusions

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- A deep learning approach to MRCP image analysis predicted clinical events in patients with PSC, and its prognostic utility exceeded that of other radiographic, clinical, and histologic assessments
- Further validation of these findings may provide a quantitative, ML-based assessment of PSC-related prognosis based on routinely collected MRCP images

References: 1. Eaton JE, et al. Hepatology 2020;71:214-24; 2. Goode E, et al. Hepatology 2019;69:2120-35; 3. Lemoinne S, et al. Clin Gastroenterol Hepatol 2019;17:2785-92; 4. Ruiz A, et al. Hepatology 2014;59:242-50; 5. Muir A, et al. Hepatology 2019:69:684-98; 6. Hu J, et al. 2018 IEEE/CVF Conference on Computer Vision and Pattern Recognition 2018:7132-41; 7. Muir A, et al. AASLD 2017, poster 279; 8. Travis N, et al. EASL 2020, poster FRI173