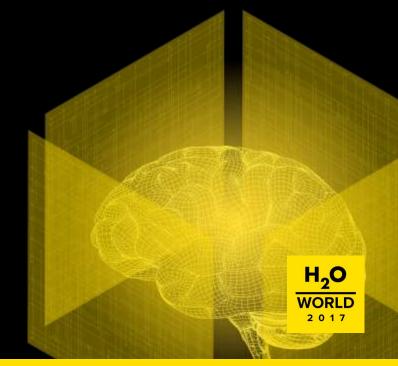




Harnessing AI to Create a Trillion Dollar Asset Class

John Mercer Chief Product Officer Ledger Investing

@john_d_mercer john@ledgerinvesting.com



Outline

- 1. My background and what led to using H2O at Ledger Investing
- 2. Ledger History a tech company from the valley
- 3. Our Mission
 - 1. Mental model to walk away with
- 4. Ledger Solutions for the Insurance Value Chain
- 5. H2O Powered Stochastic Loss Reserving
 - a. Primer on SLR
 - b. Loss Development EDA
 - c. ML approaches
 - d. Performance
 - e. How these results are used



My backgound and success with H2O

Broad Institute of M

Using ML on function networks

Patents

Method and system for a networks

WO 2016118513 A1

twork diagram shown here represents a map ு நகிein-protein interactions in a yeast (Saccharomyces cerevisiae) cell (Image by

https://www.google.com/patents/WO2016118513A1?cl=en

Hawoong Jeong, KAIST, Korea).

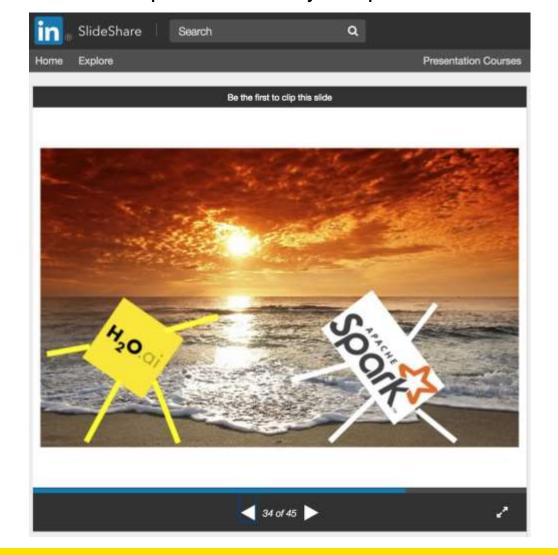


My backgound and success with H2O



https://www.slideshare.net/SparkSummit/spark-summit-eu-talk-by-johnathan-mercer

H2O & Spark Love Story at Spark Summit





Continuing this success at Ledger Investing



ledger Investing



Ledger Investing





Ledger Investing – Founders



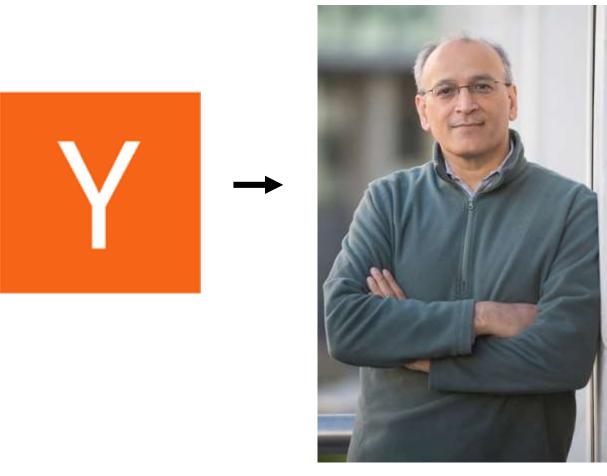
Julien Brissonneau



Aymeric Rabot



Ledger Investing – Meeting Samir Shah



Samir Shah, CEO









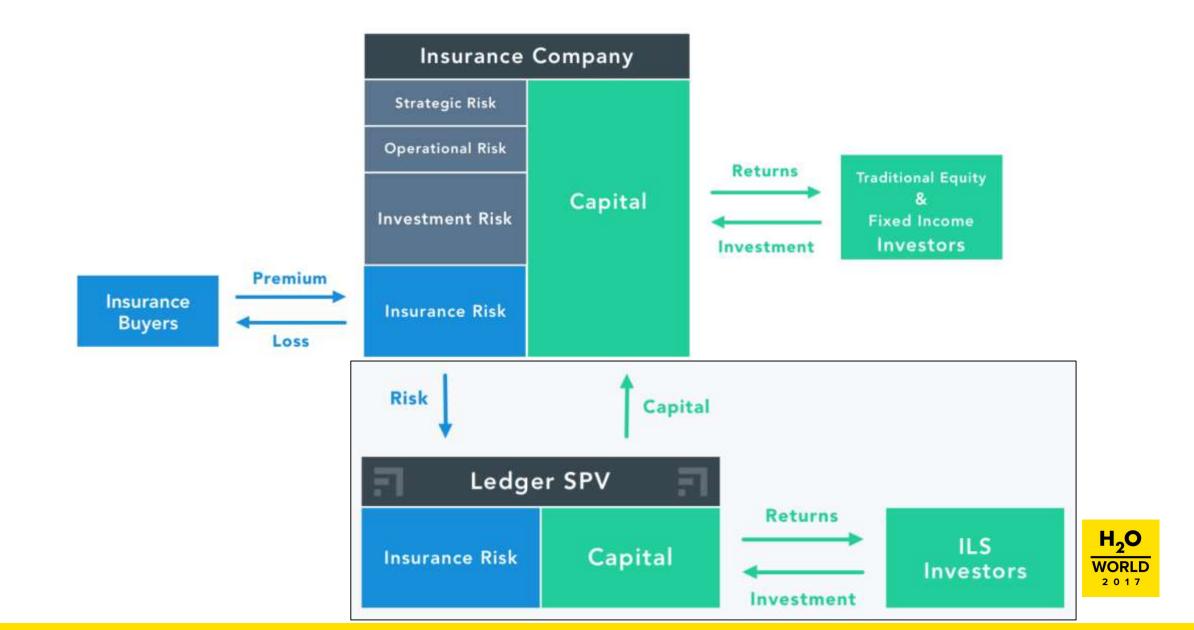




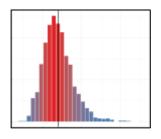




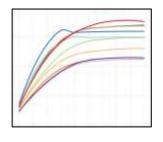
Insurance Value Chain Transformation



Ledger Solutions

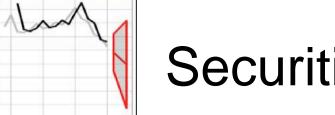


Risk Modeling H₂O



Stochastic Loss Reserving







H2O Powered Stochastic Loss Reserving (SLR)

- a. Primer on SLR
- b. Industry Loss Development EDA
- c. ML approaches
- d. Performance
- e. How these results are used



Premiums and Losses for Insurer ABC 1989-1998

Insurer ABC - Filing Year 1998 (Commerical Auto)										
Accident	AY	Net	Premiums	Losses	Ultimate					
Year	Index		(M)	(M)	Loss Ratio					
1989	9	\$	34,652	\$34,201	99%					
1990	8	\$	34,694	\$ 28,358	82%					
1991	7	\$	30,197	\$ 21,129	70%					
1992	6	\$	25,121	\$ 16,846	67%					
1993	5	\$	26,912	\$ 14,744	55%					
1994	4	\$	26,729	\$ 16,914	63%					
1995	3	\$	24,561	\$ 18,141	74%					
1996	2	\$	28,475	\$ 16,376	58%					
1997	1	\$	23,687	\$ 17,304	73%					
1998	0	\$	22,443	\$ 15,938	71%					



Premiums and Losses for Insurer ABC 1989-1998

Accident	AY	Net Premiums	Losses	Ultimate	
Year	Index	(M)	(M)	Loss Ratio	
1989	9	\$ 34,652	\$34,201	99%	

Insurer ABC - Filing Year 1998 (Commerical Auto)										
	AY	Net Premiums					Losses	Ultimate		
Accident Year	Index	(M)	Lag1	Lag2	Lag3	•••	(M)	Loss Ratio		
1989	9	\$ 34,652	\$ 5,871	\$ 14,666	\$ 22,827	••••	\$ 34,201	99%		



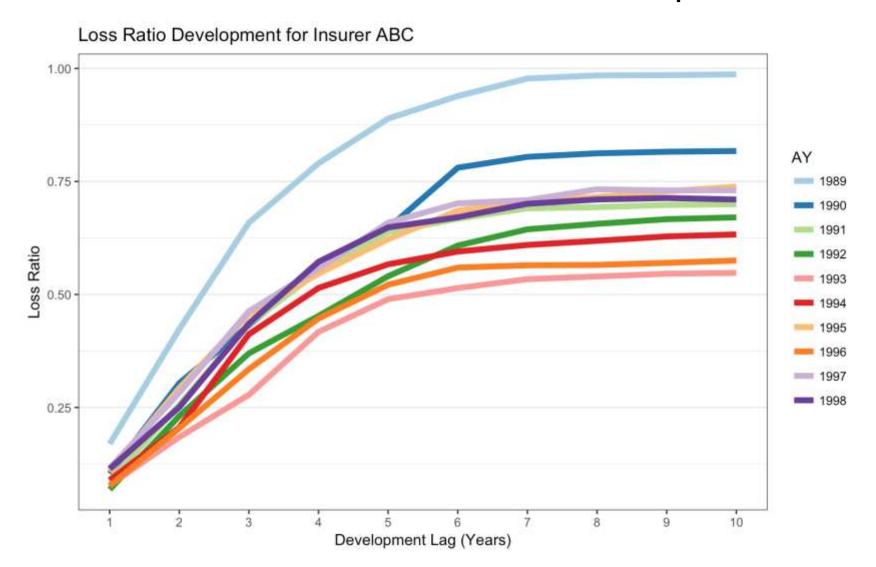
Insurer ABC 1989-1998 Loss Development for all AY's

Insurer ABC - Filing Year 1998 (Commerical Auto)														
Accident	AY	Net Premiums											Losses	Ultimate
Year	Index	(M)	Lag1	Lag2	Lag3	Lag4	Lag5	Lag6	Lag7	Lag8	Lag9	Lag10	(M)	Loss Ratio
1989	9	\$ 34,652	\$ 5,871	\$14,666	\$22,827	\$27,381	\$30,811	\$32,541	\$33,884	\$34,122	\$34,146	\$34,201	\$34,201	99%
1990	8	\$ 34,694	\$ 3,582	\$10,542	\$14,980	\$19,416	\$22,436	\$27,078	\$27,910	\$28,182	\$28,303	\$28,358	\$28,358	82%
1991	7	\$ 30,197	\$ 2,662	\$ 7,717	\$13,110	\$16,649	\$19,274	\$20,158	\$20,856	\$20,931	\$21,077	\$21,129	\$21,129	70%
1992	6	\$ 25,121	\$ 1,719	\$ 5,802	\$ 9,288	\$11,368	\$13,575	\$15,279	\$16,182	\$16,486	\$16,745	\$16,846	\$16,846	67%
1993	5	\$ 26,912	\$ 2,109	\$ 4,964	\$ 7,482	\$11,231	\$13,175	\$13,836	\$14,368	\$14,528	\$14,691	\$14,744	\$14,744	55%
1994	4	\$ 26,729	\$ 2,382	\$ 5,504	\$11,012	\$13,747	\$15,149	\$15,904	\$16,293	\$16,537	\$16,797	\$16,914	\$16,914	63%
1995	3	\$ 24,561	\$ 2,644	\$ 7,192	\$11,208	\$13,418	\$15,279	\$16,838	\$17,344	\$17,576	\$17,884	\$18,141	\$18,141	74%
1996	2	\$ 28,475	\$ 2,174	\$ 5,813	\$ 9,531	\$12,724	\$14,852	\$15,932	\$16,075	\$16,100	\$16,230	\$16,376	\$16,376	58%
1997	1	\$ 23,687	\$ 2,729	\$ 6,695	\$10,966	\$13,208	\$15,605	\$16,621	\$16,787	\$17,358	\$17,294	\$17,304	\$17,304	73%
1998	0	\$ 22,443	\$ 2,559	\$ 5,622	\$ 9,778	\$12,844	\$14,552	\$15,056	\$15,726	\$15,939	\$16,011	\$15,938	\$15,938	71%

Each Accident Year has had 10 years of development



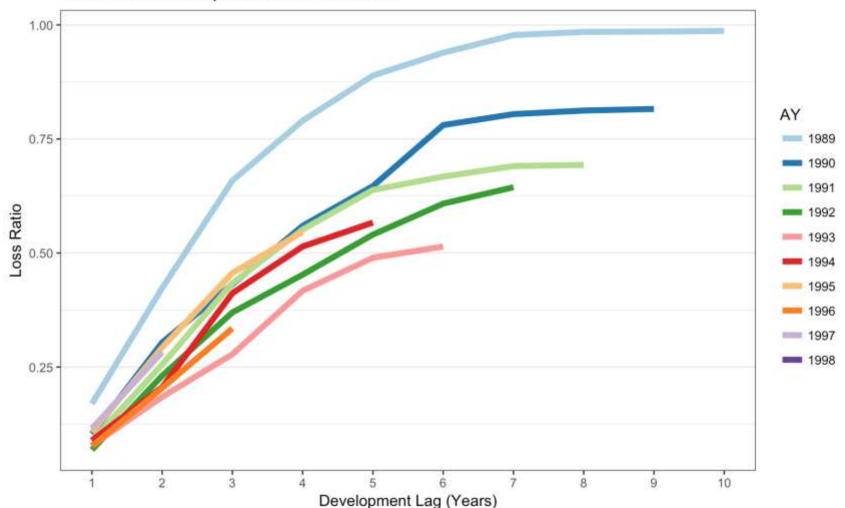
Insurer ABC 1989-1998 Visual Loss Development





But what if we were in 1998? Complete the trajectories

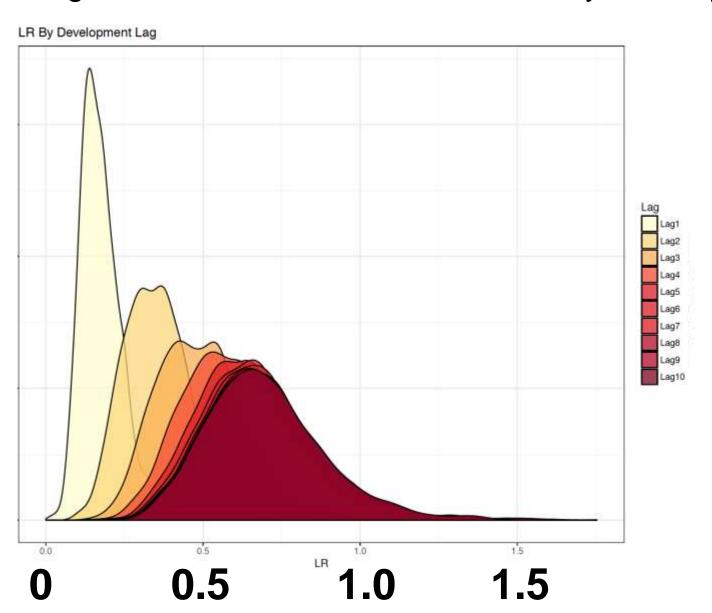
Loss Ratio Development for Insurer ABC





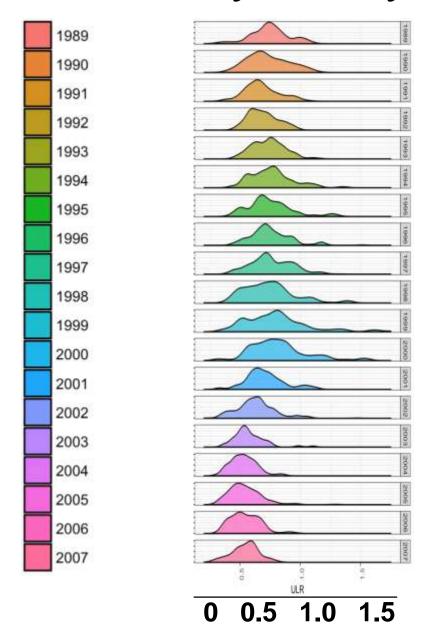
Loss Development EDA

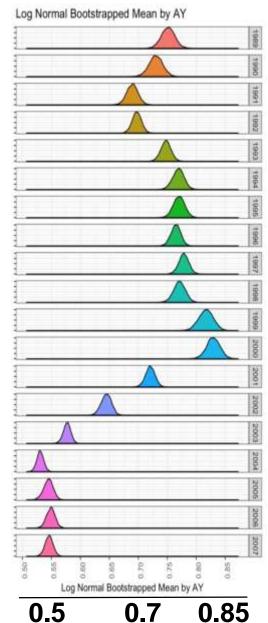
The Convergence of LR Predictive Distribution by Development Year

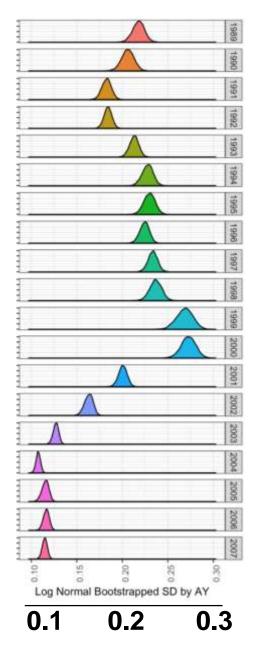




Industry LR by Accident Year – 1989-2007







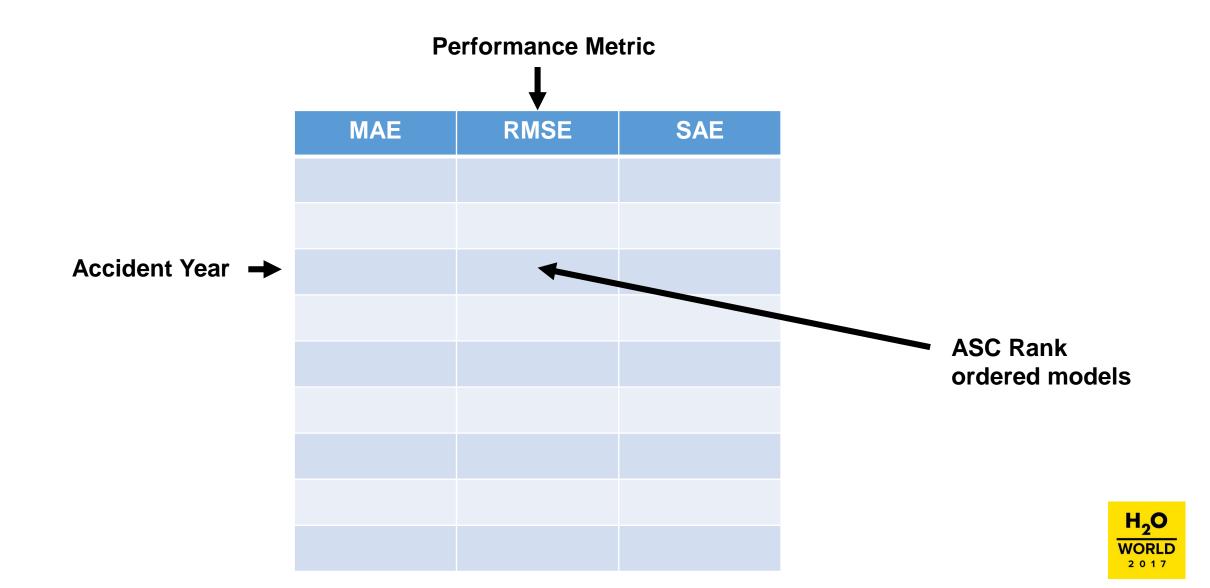


ML Approaches

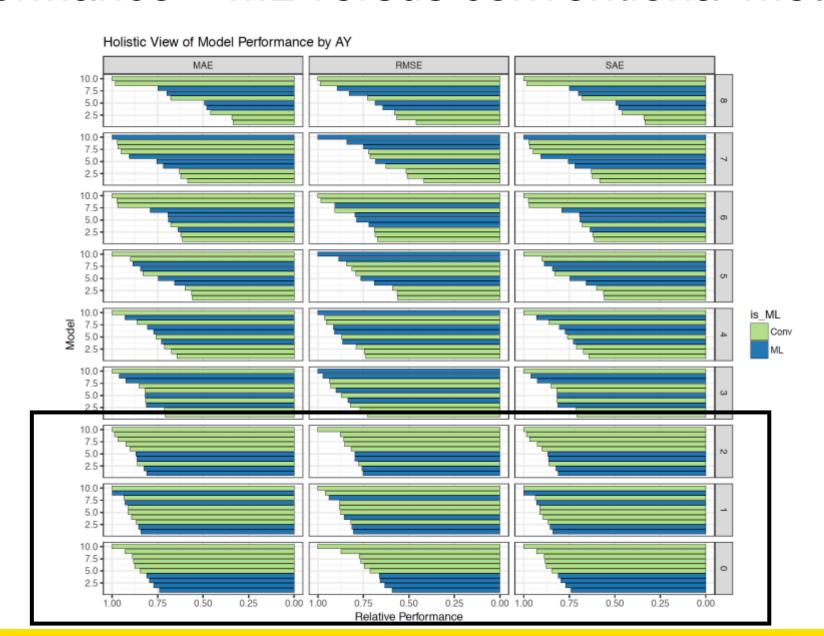
- 1. GLMs
- 2. Random Forests
- 3. Gradient Boosted Machines
- 4. XGBoost
- 5. Artificial Neural Networks



Performance – ML versus conventional methods



Performance – ML versus conventional methods





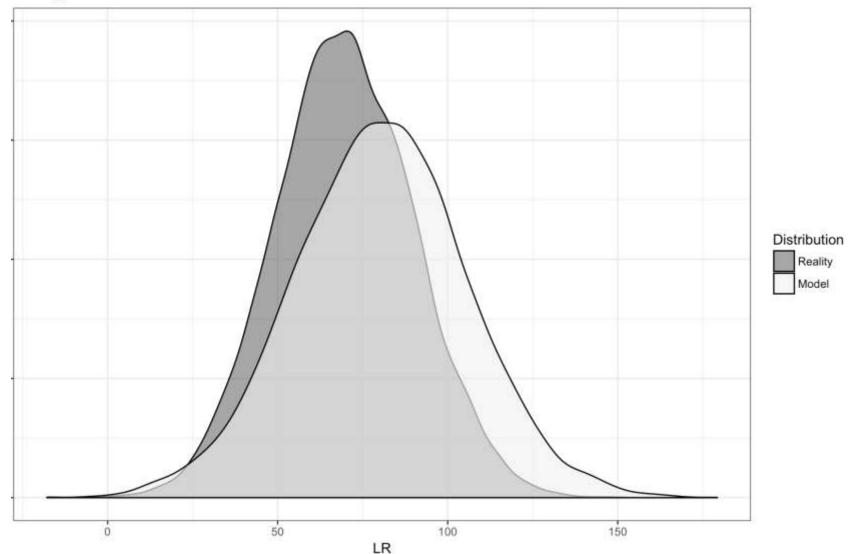
Performance – ML approaches dominate in most important areas





Performance – we need the predictive distribution, not just the expected value

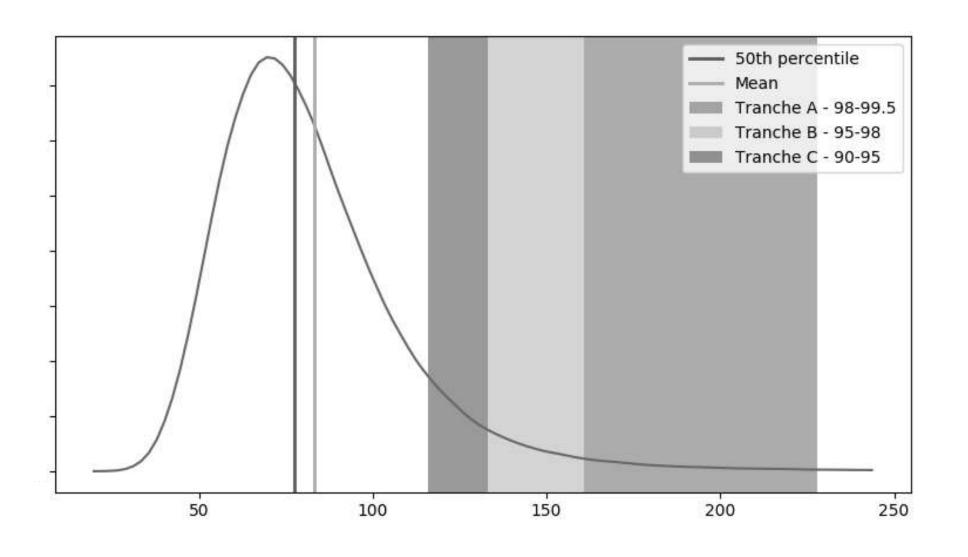






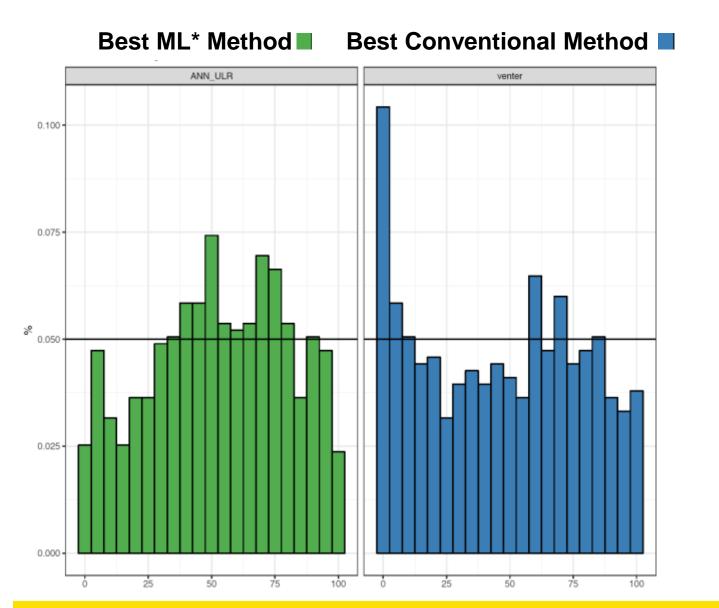
How investors use these results

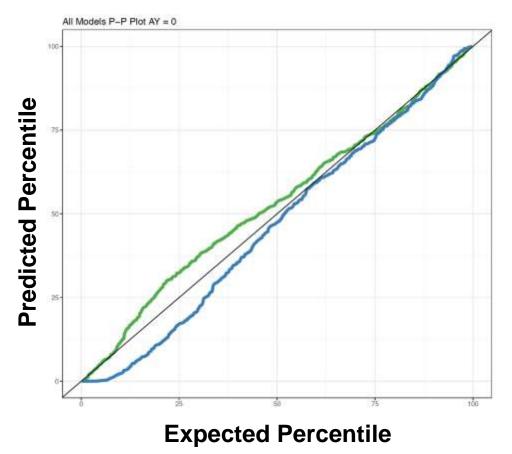
Risk-return tradeoffs based on probabilities





Performance – are the percentiles of our predictive distribution uniformly distributed?







Why we chose H2O

What makes this group so unique



