Wages of Wins

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19/09/2016, MLSA '16 @ECML/PKDD

Motivation

- Previous work : predicting match outcomes
- Decent accuracies but what for ?
- MLSA '13 recommendation : beat the bookie !
- Could one actually make money ?

Setting

- Predictions : NCAAB « march madness », NBA, NFL – all 2015/2016
 - Outcomes, not scores Haven't managed yet to train good score predictors
 - Basketball as in earlier work, Football described later
- Data: www.kenpom.com, www.basketball-reference.com, www.pro-football-reference.com
- Money Lines : www.vegasinsider.com

Money Lines

- Goal for sports book : make money !
 - Partition betting volume s.t.
 bettors' wins (+profit) covered by losses
 - Adjust « money lines »

Money Lines (2)

	Match-Up	Favorite	Underdog	Fav-Line	DogLine
1	Detroit at Atlanta	Atlanta	Detroit	300	240
2	Utah at Detroit	Detroit	Utah	110	-110

- Match 1 :
 - Correct bet Detroit, 100\$ bet = 240\$ gain

Atlanta clear favorite : motivate people to bet on Detroit

- Correct bet Atlanta, 300\$ needed to win 100\$ → 100\$ bet = 33.33\$ gain
- Match 2 : « Pick 'em »
 - Correct bet of 100\$ = 90.90\$ gain

Remember : wins + PROFIT

Initially based on model, bettor behavior shifts lines

Correct Dog-Bet > Correct Pick 'em-Bet > Correct Fav-Bet

Simulated betting

- Bet all matches
 - 100\$ per match
 - Separate regular/post-season for NBA/NFL
- At money-line w/smallest spread
 - E.g. 170/150 instead of 200/170

Conservative but NOT as conservative as I thought !

- Tally winnings per day
 - And over full period

« Vegas accuracy »

- « Predicts » only favorites
- Pick 'ems → Coin flips →
 50 % correct expected
 - Best case all correct

Can make big difference financially

- Worst case all wrong

NCAAB (67 matches)

5 Pick 'ems

<u>Vegas</u> w/ Pick 'ems w/o Pick 'ems Pay-out Accuracy Pay-out Exp. Acc Best Acc. Pay-out Worst Pay-out 0.6865 0.6612 30.26 0.6417 7.51 484.76 0.5821 -469.7341 Favs **Predictors Naïve Bayes Multilayer Perceptron Simplified KenPom** Classifier Accuracy 0.6865 0.6417 0.7014 Pay-out -30.83-834.81 -231.34

Distribution of predictions

Classifier	Favs	Dogs	PEs
NB	42	2	2
MLP	40	0	3
KP	43	0	4

NCAAB (67 matches)

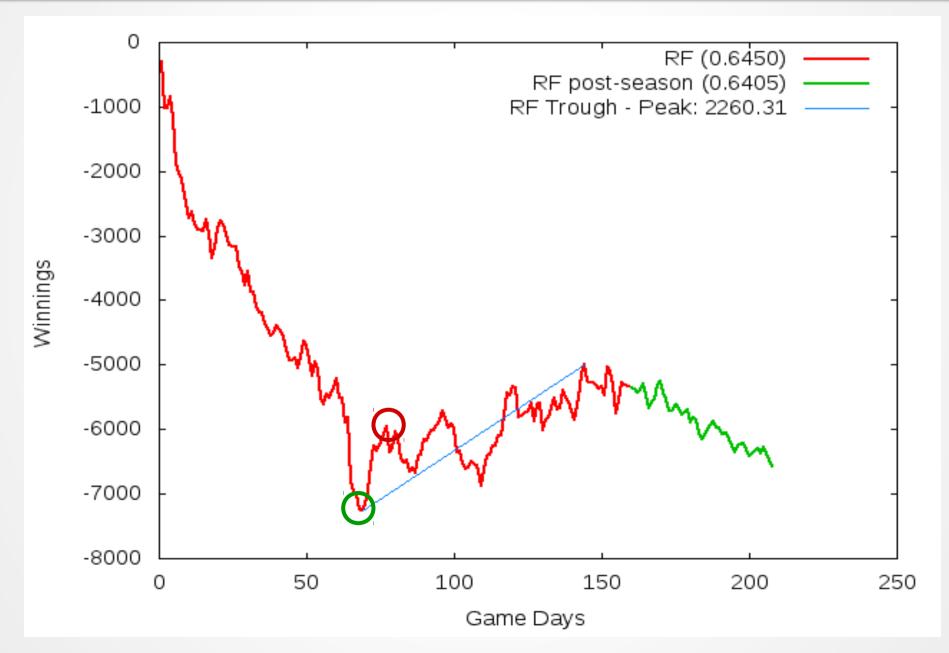


NBA (1288, 115 PEs)

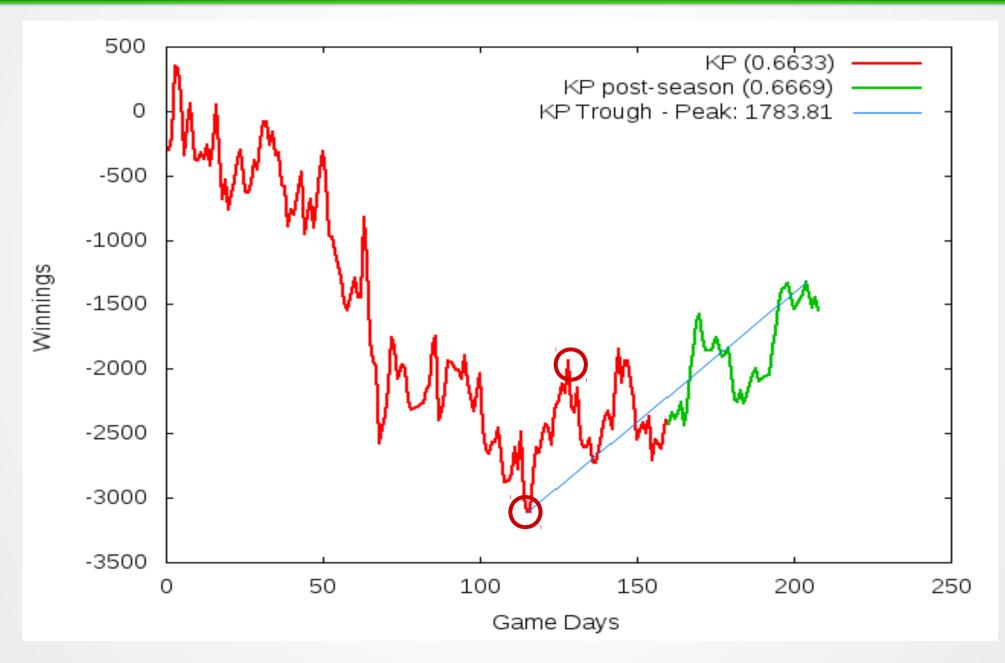
<u>Vegas</u>

w/	o Pick 'em	S		wl	Pick 'ems			
Accura			Pay-out			It Wor	st. Pav	-out
0.712		•	-1857.3			4 0.64		828.81
835 Favs			<u>Predictors</u>					
	Classifi	er NE	3	MLP	Random I	Forest	KP	
	Accuracy	0.6607		0.6615	0.6405		0.6669	
	Pay-out	-2882.21		-2005.56	-6580.88		-1543.05	
		<u>Distr</u>	ibutior	<u>n of pre</u>	dictions		gular + st-season	
		Classifier	Favs	Dog	s PE	S		
		NB	691	57	48			
		MLP	707	60	22			
		RF	685	61	28			
		KP	725	59	12			

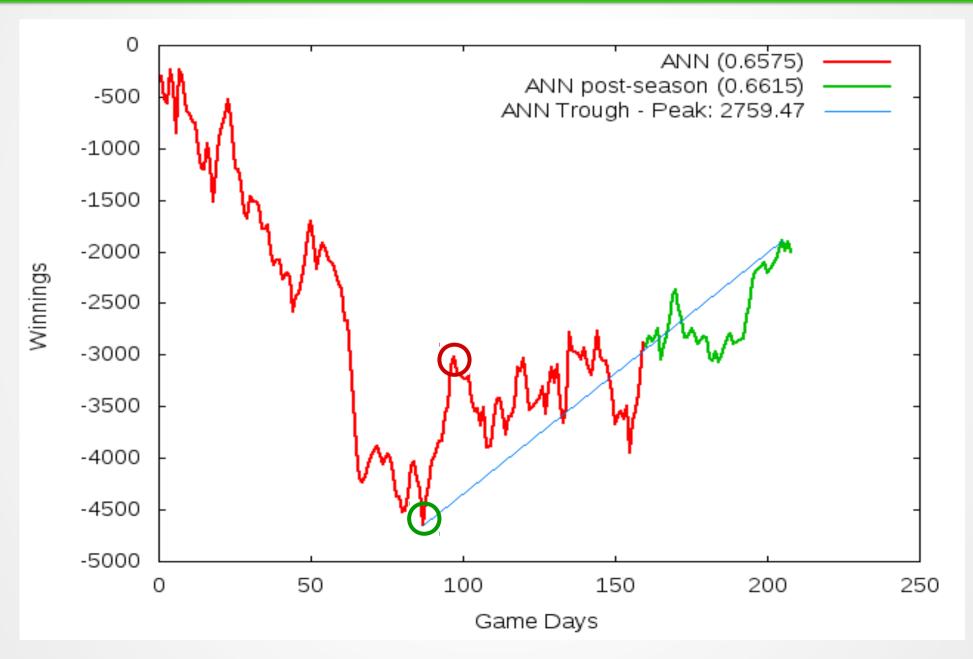
RF winnings curve



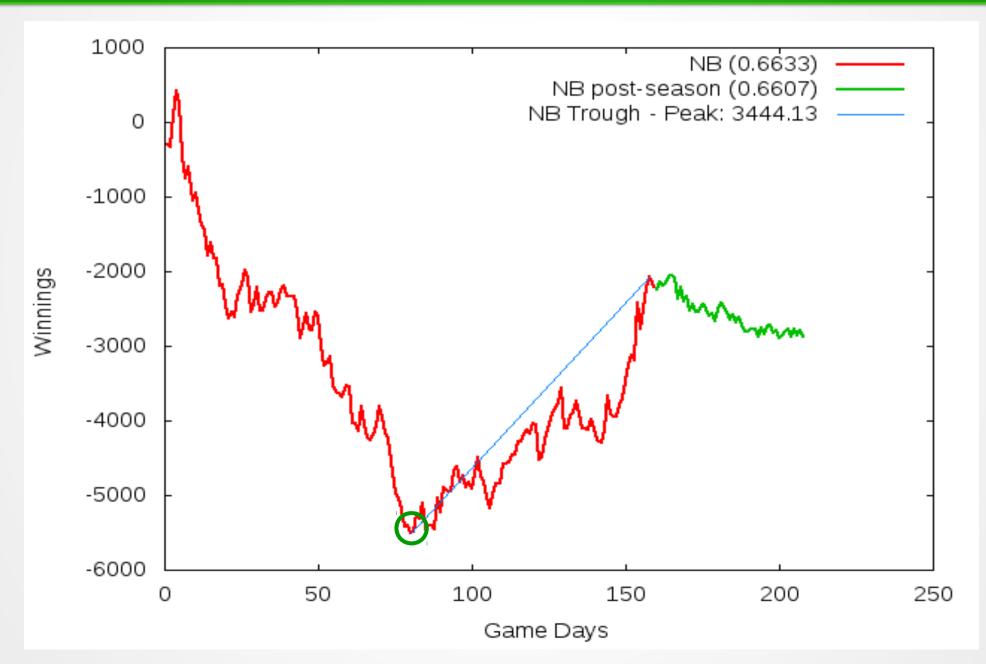
KP winning curve



MLP winning curve



NB winning curve



NFL prediction

- Different game :
 - Different outcomes for possessions
 - Not many scoring events
 - Two (three) distinct sub-teams
 - Fewer matches
- State of the art : « Defense-adjusted Yards above Replacement » (DYAR)
 - Needs fine-grained data, expensive calculation
- My « solutions » :
 - Averaged (adjusted) statistics
 - Opponents' statistics
 - Strength of schedule

NFL (251, 29 PEs)

<u>Vegas</u>

	w/o Pic	:k 'ems		w/ Pick 'ems						
	Accuracy	Pay-out	Exp. Acc	Pay-out	Best Acc.	Pay-out	Worst	Pay-out		
	0.6441	-1215.69	0.6294	-1251.92	0.6852	1420.68	0.5697	-4115.42		
143 Favs Predictors										
	Classifi	er	NB	MLP	RF	Sim	ple Rating	g System		
Accuracy			0.6335	0.5896	0.5737	0.5896	0.5896			
Pa	Pay-out		1777.98	729.92	-1591.68	-1255.71				

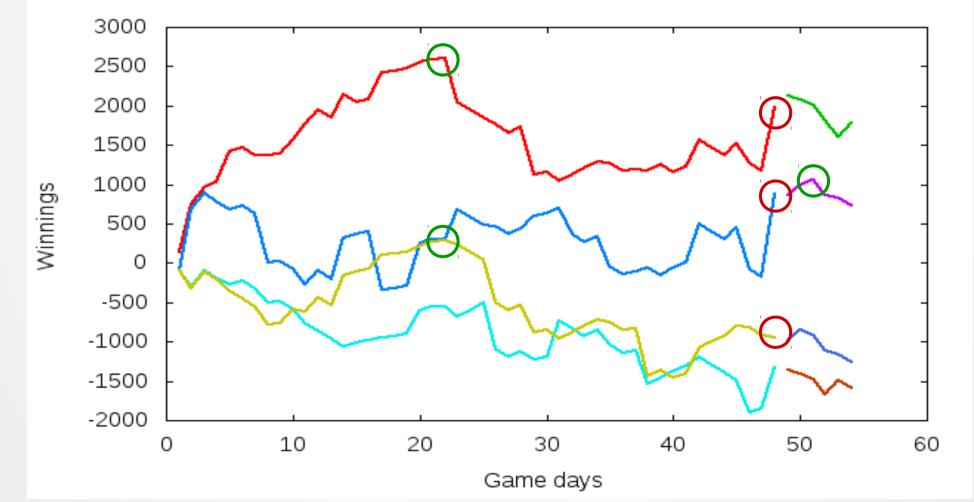
Distribution of predictions

Classifier	Favs	Dogs	PEs
NB	119	26	14
MLP	103	29	16
RF	111	17	16
SRS	115	18	15

NFL winning curves

- RF (0.5792) -----
- RF post-season (0.5737) ----
 - SRS (0.5958) -----
- SRS post-season (0.5896)

- NB (0.6417)
- NB post-season (0.6335) -
 - ANN (0.5917) -----
- ANN post-season (0.5896) -



Can an amateur make money ?

- « Jein »
- March madness : too few matches, too unpredictable
- NBA : know when to get in
- NFL : know when to get out

Use Naïve Bayes !

Future work

 Characterize distribution by confidence

Not all Favs/Dogs are equal !

- Optimize gains instead of accuracy
- Learn strategies for which matches to bet on

I assume no liability for betting losses !

Work-in-progress, unpublished things (like NFL methodology) :

http://scientificdm.wordpress.com

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