

Text-based Crude Oil Price Forecasting

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□ Introduction:

Forecasting Methods: Time Series, Econometrics

$$\hat{Y} = f(X) + T$$

T: Sentiment Score, Topics, **Word Embedding, Sentence Embedding**

Forecasting Model: Long-Short Term Memory(LSTM)

□ Variables:

Target : y_t — WTI spot price

1. y_t : WTI spot price $\rightarrow F(y_t)$;
2. T_t : Text Variable $\rightarrow F(T_t)$;
3. M_t : Marketing Variables, including futures, DIJA, USDA and google $\rightarrow F(M_t)$;
4. S_t : Sentence Embedding Variables
 S_{1t} —SIF Embedding Variables ,
 S_{2t} —Power Mean Embedding Variables ,

□ Model:

$$\mathbf{y}_t = F(\mathbf{y}_t) + F(\mathbf{T}_t) + F(\mathbf{M}_t) + \mathbf{S}_t$$

1. $\mathbf{y}_t = F(\mathbf{y}_t)$ --ARIMA, LSTM
2. $\mathbf{y}_t = F(\mathbf{y}_t) + F(\mathbf{M}_t)$ --Market
3. $\mathbf{y}_t = F(\mathbf{y}_t) + F(\mathbf{T}_t)$ --Text
4. $\mathbf{y}_t = F(\mathbf{y}_t) + F(\mathbf{T}_t) + F(\mathbf{M}_t)$ --cross impact analysis
5. $\mathbf{y}_t = F(\mathbf{y}_t) + F(\mathbf{T}_t) + F(\mathbf{M}_t) + \mathbf{S}_t$ --Sentence Embedding

□ Result:

Table 1: Performance of different models(RMSE)

Proportion of Testing Datasets		20%	30%	40%
Time Series Model	ARIMA	0.0144	0.0128	0.0121
	Auto Arima	0.0141	0.0131	0.0124
	ETS	0.0137	0.0127	0.0121
LSTM	WTI	0.0143	0.013	0.0123
	WTI + M	0.0084	0.0082	0.0087
	WTI + T	0.0153	0.0144	0.0138
	WTI + M + T	0.009	0.0094	0.0094
	WTI + M + T + S_1	0.0079	0.0079	0.0083
	WTI + M + T + S_2	0.0072	0.0077	0.0076

□ Update:

Table 2: items updated

Target: y_t	Before	Update	Notes
Numeric Variables	$M = \{M_t, M_{t-1}, \dots, M_{t-p}\}$	$M = \{M_{t-1}, \dots, M_{t-p}\}$	more practical
Corpus	yearUnited', 'U.K.', 'kkkkkkkkk'	token	cost a long time
Topic	LDA	to ensure converge	
Sentiment Score	TextBlob	TextBlob(to be comparable)	few methods available
Word Embedding	trained by our full text	Google (d = 300)	larger corpus, more precise
Sentence Embedding	$S = \{S_t, d = 80\}$	$S = \{S_{t-1}, d = 300\}$	more practical
lag	VAR	Granger Causal Text	may lag more

□ Appendix:

1. 20%: ARIMA(5,2,0)

	drift	s.e.
Coefficients:	-7e-04	5e-04
sigma^2 estimated as 0.0002284		
log likelihood=2352.3		
AIC=-4700.61 AICc=-4700.59 BIC=-4691.12		

2. 30%: ARIMA(2,1,1)

	ar1	ar2	ar3	ar4	ar5
Coefficients:	-0.8791	-0.7066	-0.5423	-0.3168	-0.1201
s.e.	0.0296	0.0384	0.0407	0.0384	0.0296
sigma^2 estimated as 0.000228: log likelihood=3136.73					
AIC=-6261.46 AICc=-6261.38 BIC=-6231.28					

3. 40%: ARIMA(0,1,0)

	ar1	ar2	ma1
Coefficients:	0.9159	0.0570	-0.9598
s.e.	0.0428	0.0321	0.0293
sigma^2 estimated as 0.0002092: log likelihood=2787.74			
AIC=-5567.48 AICc=-5567.44 BIC=-5547.89			