

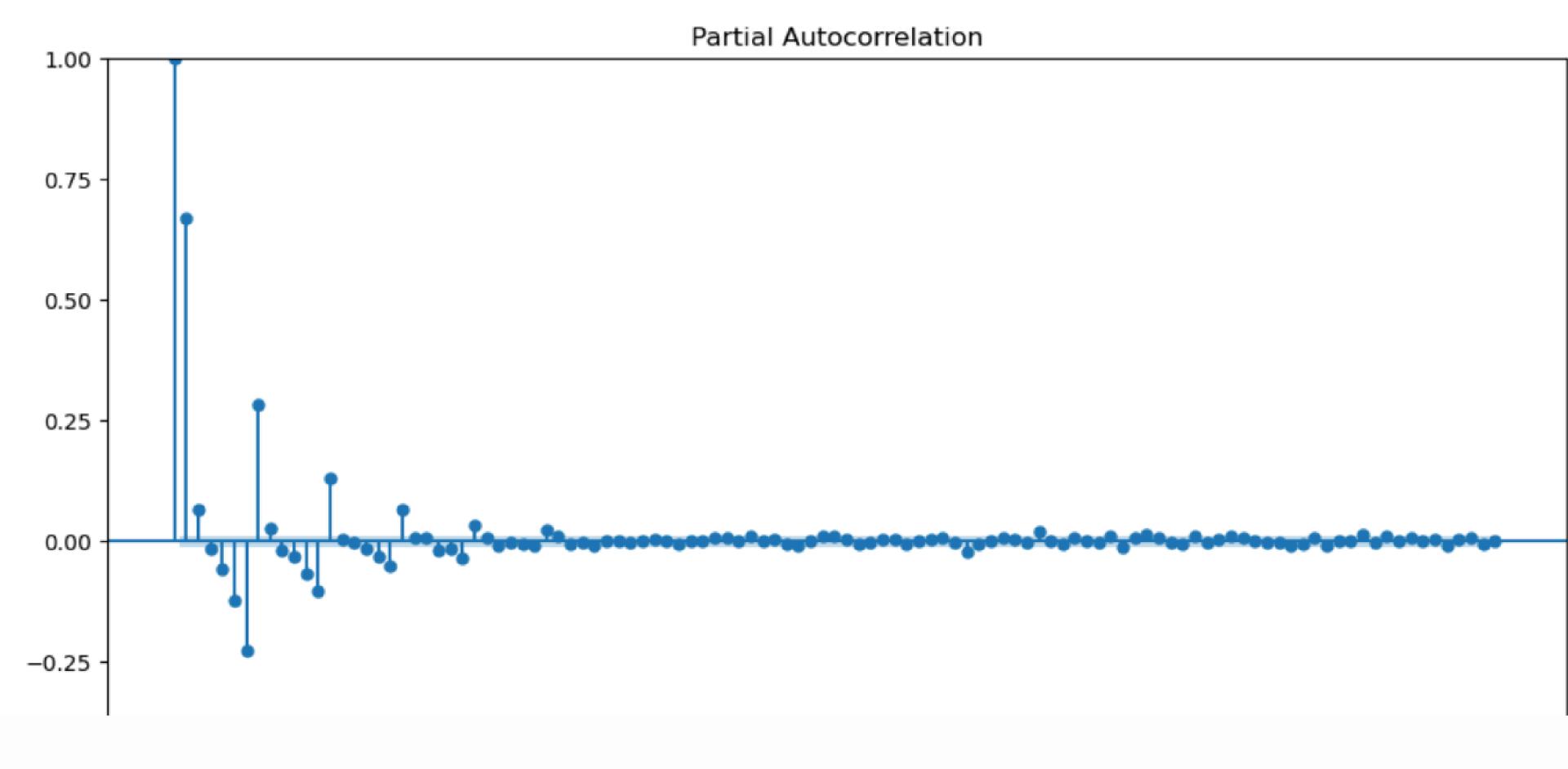
OPTIVER-TRADING AT THE CLOSE

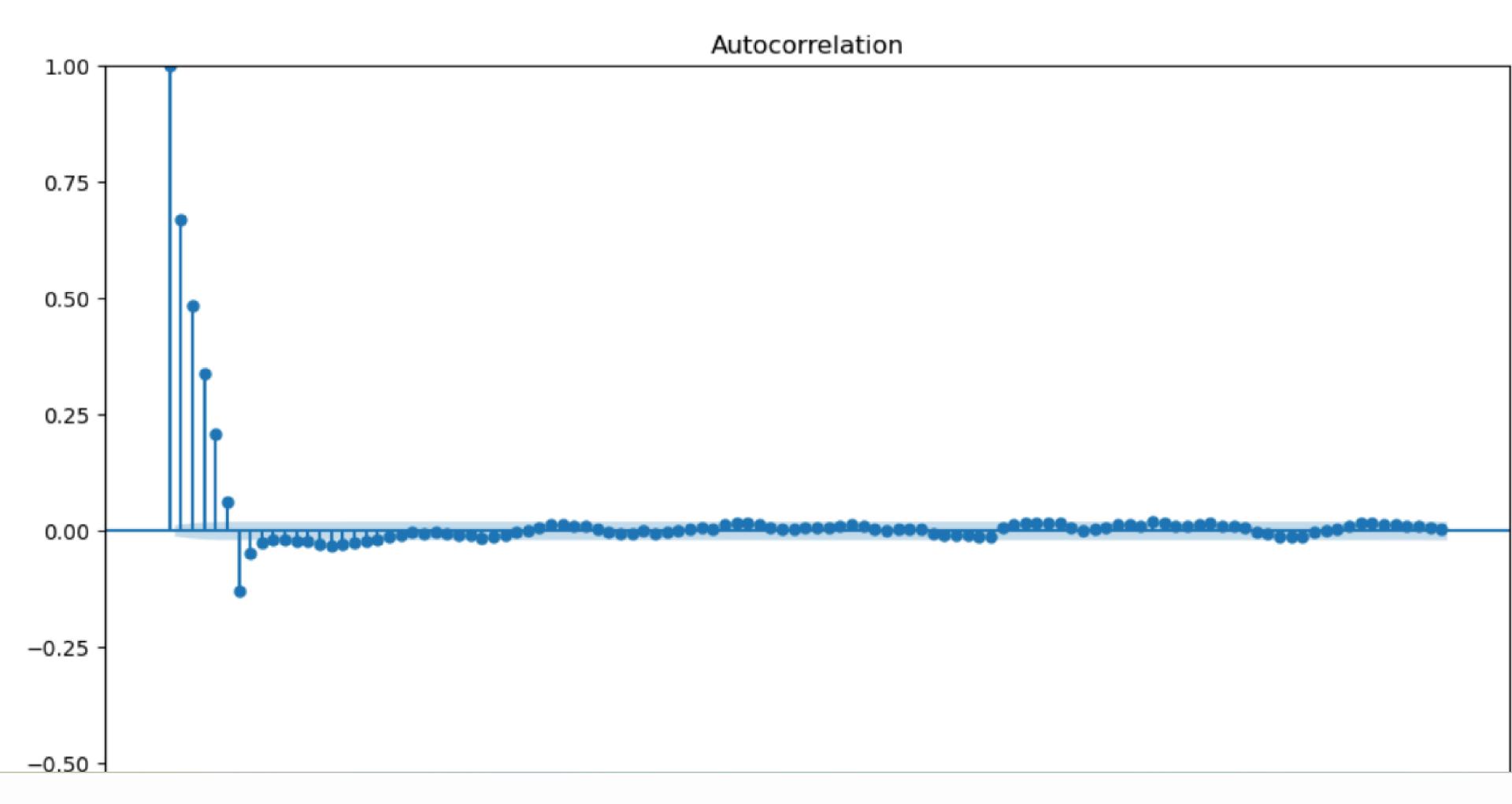
DIVYAMAN SINGH RAWAT

Stock ID	Input	sequence for target	Target prediction at time K		
0	Target at (K - N)	Target at $(K - N + 1)$	 Target at (K - 2)	Target at (K - 1)	Target prediction for stock ID 0 at K
1	Target at (K - N)	Target at $(K - N + 1)$	 Target at (K - 2)	Target at (K - 1)	Target prediction for stock ID 1 at K
199	Target at (K - N)	Target at $(K - N + 1)$	 Target at (K - 2)	Target at (K - 1)	Target prediction for stock ID 199 at K

Stock ID	Input	Input sequence for target prediction at time ID K			Target prediction at time K	
0	Target at (K - N)	Target at (K - N + 1)		Target at (K - 2)	Target at (K - 1)	Target prediction for stock ID 0 at K
1	Target at (K - N)	Target at (K - N + 1)		Target at (K - 2)	Target at (K - 1)	Target prediction for stock ID 1 at K
199	Target at (K - N)	Target at (K - N + 1)		Target at (K - 2)	Target at (K - 1)	Target prediction for stock ID 199 at K
					<u></u>	
					1	
Stock ID	Input se	equence for target pr	edi	ction at time ID (K+1)	Target prediction at time (K + 1)
0	Target at (K - N + 1)	Target at (K - N + 2)		Target at (K - 1)	Target at K	Target prediction for stock ID 0 at (K+1)
1	Target at (K - N + 1)	Target at (K - N + 2)		Target at (K - 1)	Target at K	Target prediction for stock ID 1 at (K+1)
199	Target at (K - N + 1)	Target at (K - N + 2)		Target at (K - 1)	Target at K	Target prediction for stock ID 199 at (K+1)

Stock ID	Input	sequence for target	pre	diction at time ID	K	Target prediction at time K
0	Target at (K - N)	Target at $(K - N + 1)$		Target at (K - 2)	Target at (K - 1)	Target prediction for stock ID 0 at K
1	Target at (K - N)	Target at $(K - N + 1)$		Target at (K - 2)	Target at (K - 1)	Target prediction for stock ID 1 at K
199	Target at (K - N)	Target at $(K - N + 1)$		Target at (K - 2)	Target at (K - 1)	Target prediction for stock ID 199 at K
					1	
Stock ID	Input se	equence for target pr	edi	ction at time ID (K+1)	Target prediction at time (K + 1)
0	Target at $(K - N + 1)$	Target at $(K - N + 2)$		Target at (K - 1)	Target at K	Target prediction for stock ID 0 at (K+1)
1	Target at (K - N + 1)	Target at (K - N + 2)		Target at (K - 1)	Target at K	Target prediction for stock ID 1 at (K+1)
					•••	
199	Target at (K - N + 1)	Target at $(K - N + 2)$		Target at (K - 1)	Target at K	Target prediction for stock ID 199 at (K+1)
					—	
					1	
Stock ID	Input s	equence for target pr	redi	ction at time ID (K+2)	Target prediction at time (K + 2)
0	Target at $(K - N + 2)$	Target at (K - N + 1)		Target at K	Target at (K+1)	Target prediction for stock ID 0 at (K+2)
1	Target at (K - N + 2)	Target at (K - N + 1)		Target at K	Target at (K+1)	Target prediction for stock ID 1 at (K+2)
199	Target at (K - N + 2)	Target at (K - N + 1)		Target at K	Target at (K+1)	Target prediction for stock ID 199 at (K+2)
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• Condition 1: -The mean must be constant and not vary with time.

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- Condition 2: The variance must be constant and not vary with time.

- Condition 1: -The mean must be constant and not vary with time.
- Condition 2: The variance must be constant and not vary with time.
- Condition 3: -There must be no periodicity in the data.

Check for conditions 1 and 2 (Constructing Samples)

stock_id	time_id	target
0	0	3.54
0	1	2.91
0	2	6.6
0	3	0.45
0	4	-3.4
0	5	-5.4
0		
0		
0	26452	-6.5
0	26453	2
0	26454	4.32

CHECK FOR STATIONARITY - 5 Check for conditions 1 and 2 (Constructing Samples)

					Windows	ize=3		
-11-11					stock_id	time_id	target	
stock_id	time_id	target			0	0	3.54	
0	0	3.54			0	1	2.91	1 st Sample
0	1	2.91			-	1		
0	2	6.6			0	2	6.6	
0	3	0.45			0	3	3.54	
0			 Convert to samples	>	0	4	2.91	2 nd Sample
U	4	-3.4			0	5	6.6	1
0	5	-5.4			0			
0					0			1
0						00450	0.5	no colli
٥	26452	-6.5			0	26452	-6.5	8818 th
U					0	26453	2	Sample
0	26453	2			0	26454	4.32	
0	26454	4.32						

Check for conditions 1 and 2 (Constructing Samples)

						Windows	ize=3									
and the	Norway Lab					stock id	time_id	target								
stock_id	time_id	target	_				0	3.54								
0	0	3.54				0	U		1 st Sample							
0	1	2.91				0	1	2.91								
0	2	6.6				0	2	6.6								
0	2					0	3	3.54								
0	3	0.45		Convert to samples	>	0	Δ	2.91	2 nd Sample	 Run ANOV	A for the s	amales	>	Run Leven	e test for ti	he samules
0	4	-3.4		Convert to sumptes	,	-	-		2 Sample	 III.III.AII O	Tion the s	urripa c a		nun Leven	teat for th	re auripres
0	5	-5.4				0	5	6.6								
0	-	-0.4				0										
U	***					0										
0						0	26452	-6.5	8818 th							
0	26452	-6.5				0	26453	2								
0			-			_		_	Sample							
U	26453	2				0	26454	4.32								
0	26454	4.32														

LSTM MODEL - 1

Constructing input sequence

Lagged Target Values with High DirectCorrelation: -

- Lag 1		
- Lag 7		

LSTM MODEL - 2

Constructing input sequence

Lagged Target Values with High DirectCorrelation: -

- Lag 1		
- Lag 7		

Lagged Target Va	Lagged Target Values with High Indirect Correlation:								
- Lag 2									
- Lag 3									
- Lag 4									
- Lag 5									
- Lag 7									

LSTM MODEL - 3

Constructing input sequence

Lagged Target Values with High DirectCorrelation: -

- Lag 1		
- Lag 7		

Lagged Target Values with High Indirect Correlation:								
- Lag 2								
- Lag 3								
- Lag 4								
- Lag 5								
- Lag 7								

						Prediction for current Time ID	
						Λ	
	Sequence					1	
>	Lag 7 Target	Lag 5 Target	Lag 4 Target	Lag 3 Target	Lag 2 Target	Lag 1 Target	

LSTM MODEL - 4 Testing

	time_id					
26283	26285	26286	26287	26288	26289	26290

LSTM MODEL - 5 Testing

	time_id					
26283	26285	26286	26287	26288	26289	26290
26284	26286	26287	26288	26289	26290	26291
26285	26287	26288	26289	26290	26291	26292
26286	26288	26289	26290	26291	26292	26293
26287	26289	26290	26291	26292	26293	26294
26288	26290	26291	26292	26293	26294	26295
26289	26291	26292	26293	26294	26295	26296

LSTM MODEL - 6 Testing

	time_id					
26283	26285	26286	26287	26288	26289	26290
26284	26286	26287	26288	26289	26290	26291
26285	26287	26288	26289	26290	26291	26292
26286	26288	26289	26290	26291	26292	26293
26287	26289	26290	26291	26292	26293	26294
26288	26290	26291	26292	26293	26294	26295
26289	26291	26292	26293	26294	26295	26296
26290	26292	26293	26294	26295	26296	26297
26291	26293	26294	26295	26296	26297	26298

OBSERVATIONS & CONCLUSIONS - 1

- The MAD achieved on the test data is an improvement over the MAD achieved by the winning team on the Kaggle competition.
- Further experimentation needed:-
 - Alternative models
 - Additional features

OBSERVATIONS & CONCLUSIONS - 2

- The MAD achieved on the test data is an improvement over the MAD achieved by the winning team on the Kaggle competition.
- Further experimentation needed:-
 - Alternative models
 - Additional features

Version 2 of the project planned soon.



CONTACT ME

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Project Github: - https://github.com/DivNewBeg/Optiver-trading-at-close

