# Hotel Rooms and Revenue Forecaster

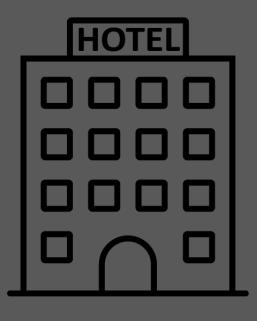
Author: Artem Lukinov



# Hotel Rooms and Revenue Forecaster

Author: Artem Lukinov





Room Nights Sold forecast: used for scheduling labor and controlling variable expenses to optimize business's bottom line daily

Revenue forecast: used for renovation schedules, large investments in operations, expansions.

#### **Forecast Process**

Story as old as time: "The mean king and the manual labor"



60%

Accuracy

### How Can We Help?

What can be done to optimize this process?

### Machine Learning!

Reliance on data
Better analysis of historical data
Hidden insights
Computational power

#### We want to

Determine whether it is possible to improve the accuracy of the baseline forecast

Use Machine Learning to forecast future data based on historical data

Stretch: wrap our model into software

#### Data Source

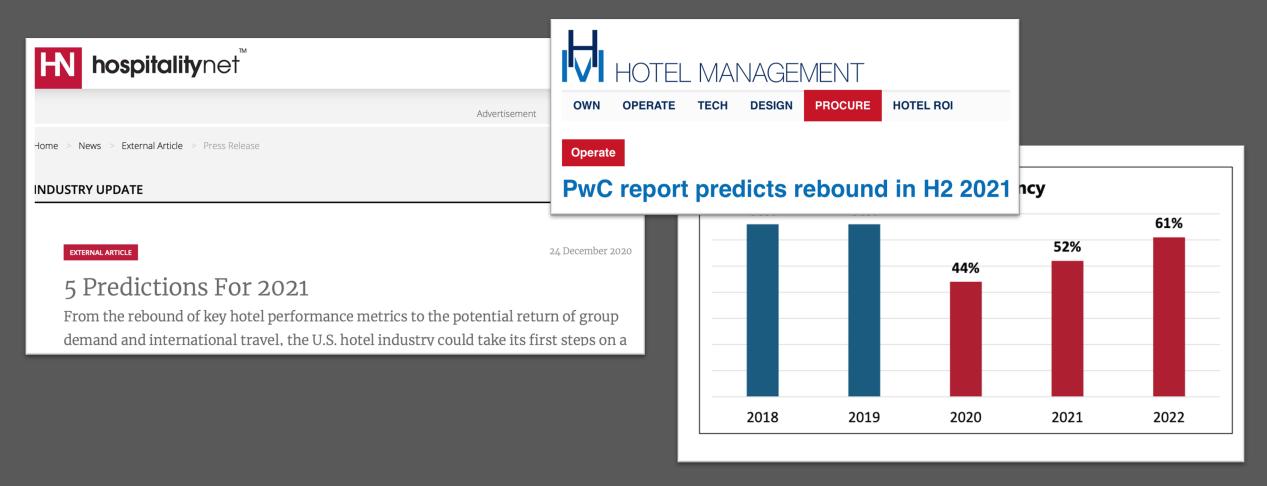
Anonymous hotel Easy cleaning

Mutual benefit Domain knowledge

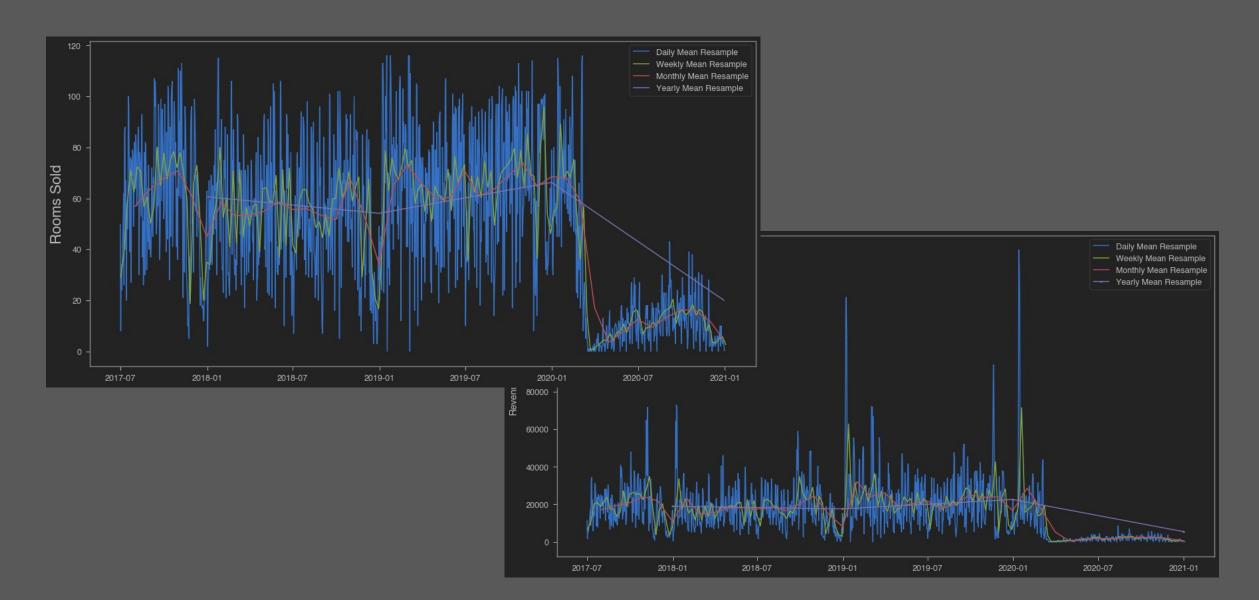
Popular format \*Multiple levels

Large sample Covid

### Covid data



### First Look



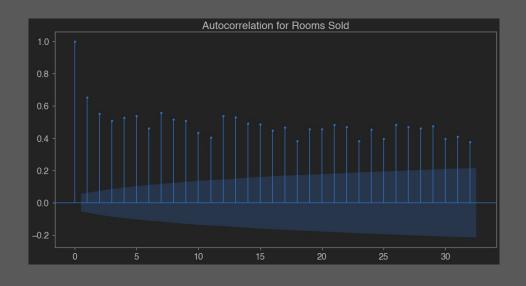
### Stationarity Check

#### Augmented Dickey-Fuller Test

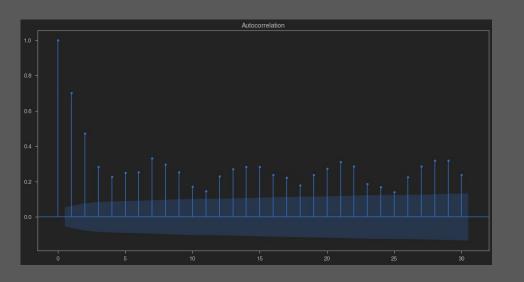
Metric	T-stat	Critical value at 1%	P-Value
Rooms	(-6.8264844063432655)	(-3.4310597571975685)	1.9419934100737604e-09
Revenue	(-9.727564233908703)	(-3.4310598342409824)	9.187072356670721e-17

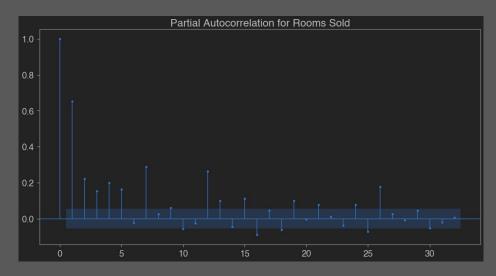
# Revenue

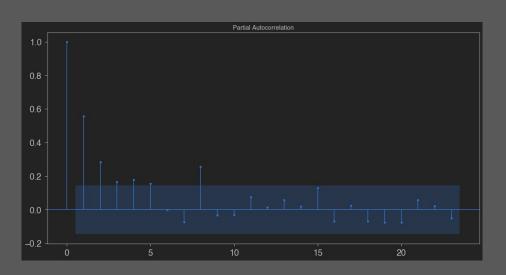
### Stationarity Check



Rooms







### Model choice

#### Time Series:

- ARIMA
- SARIMA

#### RNN:

- LSTM
- GRU

#### Additional:

- Prophet
- NeuralProphet







### Prophet and NeuralProphet

Logistic regressions



Seasonality



Special events

Linear and non-linear regressions



AR-Net
An Autoregressive
Neural Network

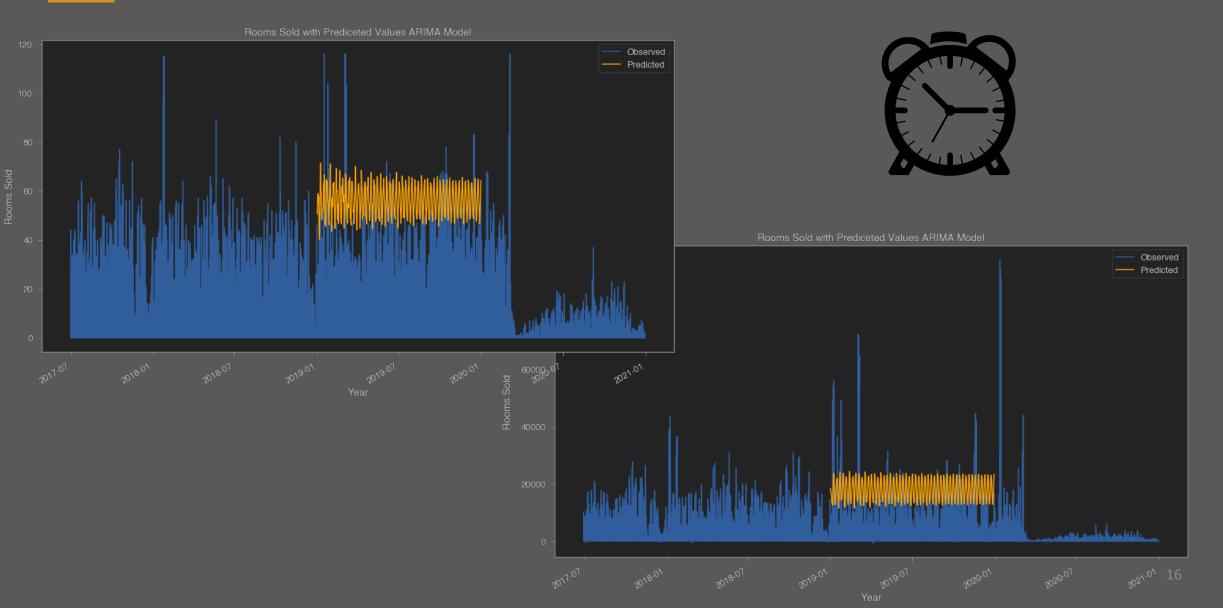
### Time Series Models

Model	AIC	RMSE	Improvement over Baseline
ARIMA (Baseline)	4995.92	27.83	-
ARIMA Grid Search	4888.26	27.11	3%
SARIMA	5039.08	48.36	-74%

#### Revenue

Model	AIC	RMSE	Improvement over Baseline
ARIMA (Baseline)	11615.94	17412.2	-
ARIMA Grid Search	11523.83	16481.31	5%
SARIMA	11690.08	24786.68	-42%

### Best Time Series Model



### Recurring Neural Networks

#### Why LTSM and GRU? Memory!

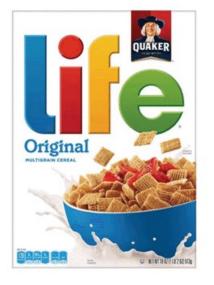
#### Customers Review 2,491



**Thanos** 

September 2018
Verified Purchase

Amazing! This box of cereal gave me a perfectly balanced breakfast, as all things should be. I only ate half of it but will definitely be buying again!



A Box of Cereal \$3.99

### Recurring Neural Networks

#### LSTM

#### GRU

#### Rooms neurons in a hidden layer epochs trained **RMSE** improvement over baseline 8 100 30.26 32 100 29.98 1% 64 50 38.56 -27%

#### Revenue

neurons in a hidden layer	epochs trained	RMSE	improvement over baseline
8	100	18918.36	
32	100	20001.84	-6%
64	50	21398.04	-13%

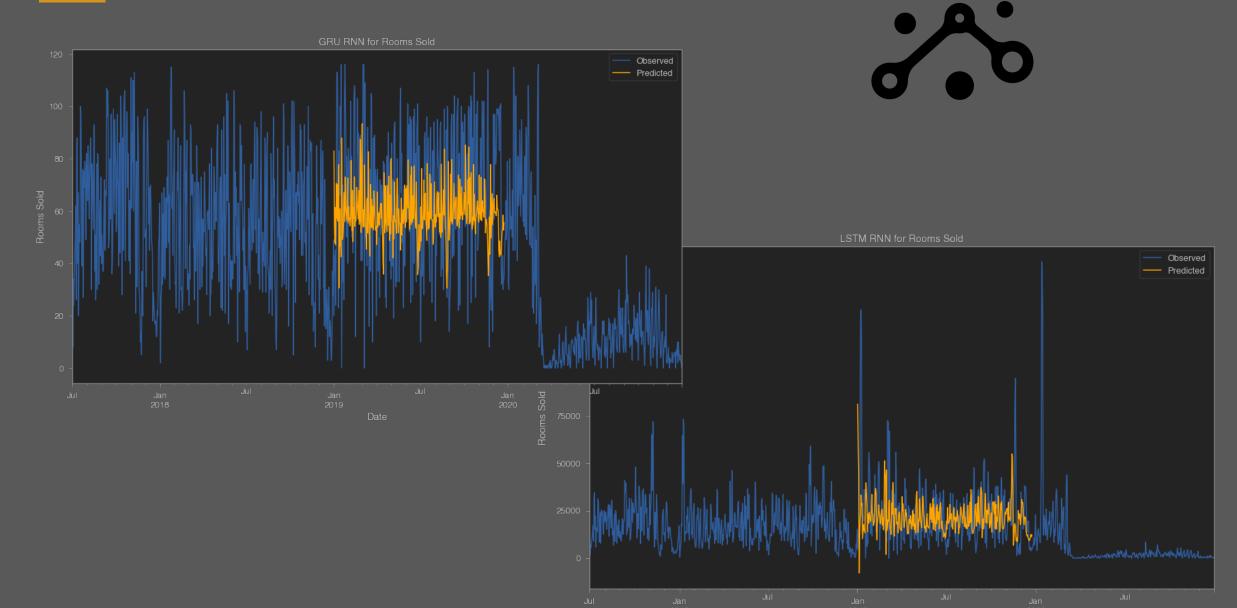
#### Rooms

neurons in a hidden layer	epochs trained	RMSE	improvement over baseline
8	100	29.87	
32	100	31.82	-7%
64	50	30.27	-1%

#### Revenue

neurons in a hidden layer	epochs trained	RMSE	improvement over baseline
8	100	19246.39	
32	100	22128.54	-7%
64	50	20892.71	-1%

### Best RNNs



### Prophet vs. NeuralProphet vs. the rest

#### Rooms

Model	RMSE	Improvement over base
Prophet (baseline)	24.43	-
Prophet Grid Search	25.61	-5%
NeuralProphet	22.59	8%

#### Revenue

Model	RMSE	Improvement over base
Prophet (baseline)	41479.04	-
Prophet Grid Search	11015.68	73%
NeuralProphet	9554.99	77%

#### **Rooms**

Rank	Model	RMSE
1	NeuralProphet	22.59
2	Prophet (baseline)	24.43
3	ARIMA Grid Search	27.11

#### Revenue

Rank	Model	RMSE
1	NeuralProphet	9554.99
2	Prophet Grid Search	11015.68
3	ARIMA Grid Search	16481

### NeuralProphet



### Key Takeaways

NeuralProphet was the best performer

RNNs have potential

Machine Learning can help in yet another industry

Exploration leads to efficiency

### How can we use this?

Explore NeuralProphet and others

Include all market segments

Use as core algorithm for a product

Thank you

**Q&A**