

Redes neuronales

Análisis de partículas PM 2.5 mediante redes LSTM

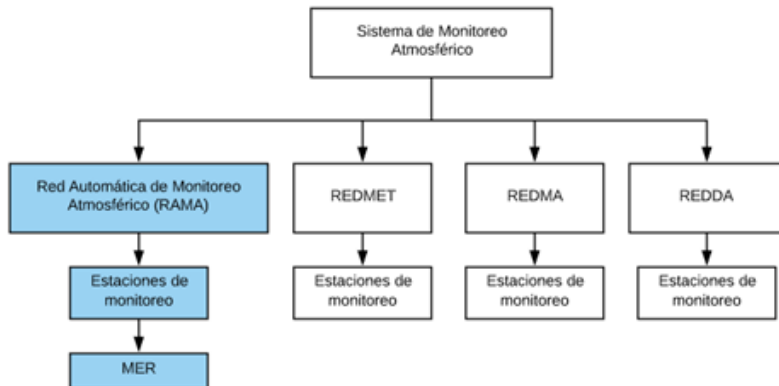
Zúñiga González Daniel Iván

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Introducción



Introducción



Objetivos

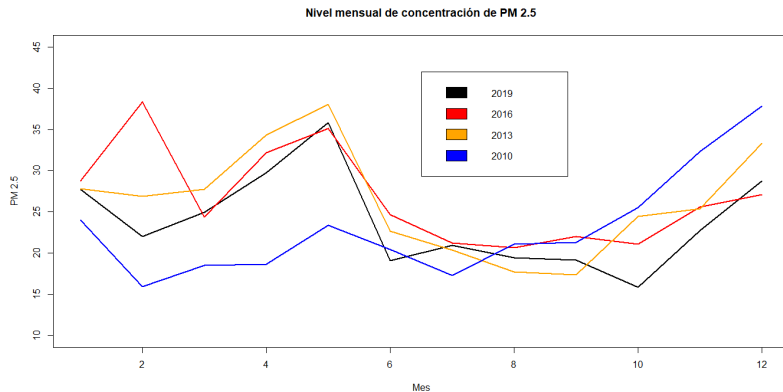
- Uso de redes neuronales artificiales (LSTM) como alternativa a los modelos estadísticos clásicos;
- Predicción año 2020

Datos (2010-2019)

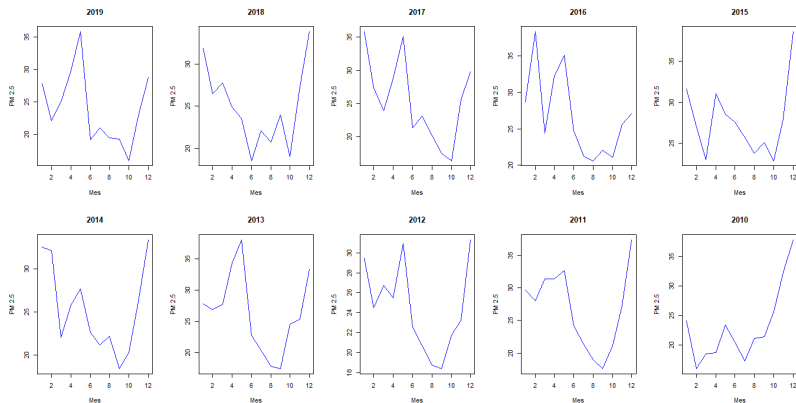
	A	B	C	D	E	F	G	H	I	J
1	FECHA	HORA	AJM	AJU	BJU	CAM	CCA	COY	GAM	HGM
2	01/01/2019	1	19	35	62	90	66	-99	-99	51
3	01/01/2019	2	17	24	88	104	84	-99	-99	61
4	01/01/2019	3	14	20	107	140	95	-99	-99	74
5	01/01/2019	4	6	15	101	162	97	-99	-99	91
6	01/01/2019	5	4	8	121	133	88	-99	-99	91
7	01/01/2019	6	7	7	93	106	77	-99	-99	101
8	01/01/2019	7	12	8	84	98	51	-99	-99	111
9	01/01/2019	8	15	7	101	82	39	-99	-99	91
10	01/01/2019	9	24	3	89	54	26	-99	-99	91
11	01/01/2019	10	24	-99	88	76	26	-99	-99	91
12	01/01/2019	11	49	-99	102	77	66	-99	-99	91

Fuente: Red automática de Monitoreo Atmosférico del Gobierno de la CDMX (<http://www.aire.cdmx.gob.mx/default.php?opc=%27aKBh%27>)

Análisis exploratorio



Análisis exploratorio



Código

```
df = pd.read_csv(r'C:/')
```

```
train, test = df[:-12], df[-12:]
```

```
scaler = MinMaxScaler()  
scaler.fit(train)  
train = scaler.transform(train)  
test = scaler.transform(test)
```

```
n_input = 12  
n_features = 1
```

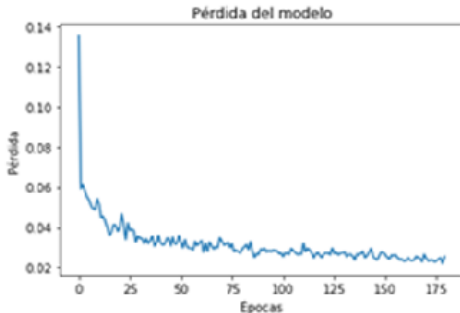
```
generator = TimeseriesGenerator(train,train,length = n_input,  
                                batch_size = 6)
```

```
model = Sequential()  
model.add(LSTM(200, activation='relu', input_shape= (n_input,n_features)))  
model.add(Dropout(0.15))  
model.add(Dense(1))  
model.compile(optimizer='Adam', loss='mse')
```

```
H=model.fit_generator(generator, epochs=100)
```


Código

```
Epoch 176/180  
16/16 [=====] - 1s 51ms/step - loss: 0.0232  
Epoch 177/180  
16/16 [=====] - 1s 52ms/step - loss: 0.0242  
Epoch 178/180  
16/16 [=====] - 1s 52ms/step - loss: 0.0249  
Epoch 179/180  
16/16 [=====] - 1s 52ms/step - loss: 0.0220  
Epoch 180/180  
16/16 [=====] - 1s 53ms/step - loss: 0.0254
```



Código

```
pred_list = []

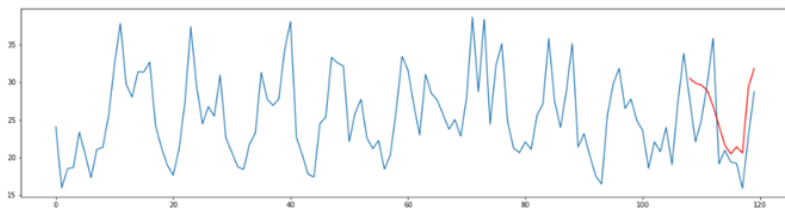
batch = train[-n_input:].reshape((1, n_input, n_features))

for i in range(n_input):
    pred_list.append(model.predict(batch)[0])
    batch = np.append(batch[:, 1:, :], [[pred_list[i]]], axis=1)

df_predict = pd.DataFrame(scaler.inverse_transform(pred_list),
                          index=df[-n_input:].index, columns=['Predicciones'])

df_test = pd.concat([df, df_predict], axis=1)
```

Código



Código

```
train = df

scaler.fit(train)
train = scaler.transform(train)

n_input = 12
n_features = 1

generator = TimeseriesGenerator(train, train, length = n_input,
                                batch_size = 6)

model.fit_generator(generator, epochs=180)

pred_list = []

batch = train[-n_input:].reshape((1, n_input, n_features))

for i in range(n_input):
    pred_list.append(model.predict(batch)[0])
    batch = np.append(batch[:, 1:,:], [[pred_list[i]]], axis=1)
```

Código

```
df_predict = pd.DataFrame(scaler.inverse_transform(pred_list),  
                           index=future_dates[-12:].index,  
                           columns=['Predicciones'])  
df_proj = pd.concat([df,df_predict],axis=1)
```

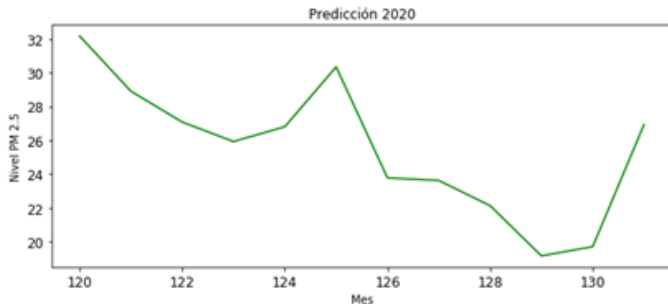
```
df_predict = pd.DataFrame(scaler.inverse_transform(pred_list),  
                           index=future_dates[-n_input:].index,  
                           columns=['Prediccion'])  
df_proj = pd.concat([df,df_predict], axis=1)
```

Predicción mensual 2020

Mes	Enero	Febrero	Marzo	Abril	Mayo	Junio
PM 2.5	32.1	28.9	27.0	25.9	26.7	30.3

Julio	Agosto	Septiembre	Octubre	Noviembre	Diciembre
23.7	23.6	22.1	19.1	19.6	26.9

Predicción mensual 2020



Predicción mensual 2020

