

IA-2022



Grupo 4

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Análisis Académico-FIUNA

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 221822 entries, 0 to 221821  
Data columns (total 23 columns):  
#   Column              Non-Null Count  Dtype  
---  ---  
0   Unnamed: 0          221822 non-null int64  
1   danho               221822 non-null int64  
2   ciclo               221822 non-null int64  
3   Cod.Assign          221822 non-null int64  
4   Asignatura          221822 non-null object  
5   Cod.Car.Sec         221822 non-null object  
6   Cod.Curso           221822 non-null int64  
7   Convocatoria        221822 non-null int64  
8   Anho                221822 non-null int64  
9   Semestre            221822 non-null int64  
10  Aprobado             221822 non-null object  
11  Anho.Firma           221822 non-null int64  
12  Primer.Par           221822 non-null int64  
13  Segundo.Par          221822 non-null int64  
14  AOT                  221822 non-null int64  
15  Primer.Rec           221822 non-null int64  
16  Segundo.Rec          221822 non-null int64  
17  Nota.Final           137062 non-null object  
18  Tercer.Par           221822 non-null float64  
19  Asis                 221822 non-null float64  
20  Cuarto.Par           28037 non-null float64  
21  4P_60                28037 non-null float64  
22  id_anony             221822 non-null object  
dtypes: float64(4), int64(14), object(5)  
memory usage: 38.9+ MB
```


Regresión Lineal

```
array(['CGF-175-09', 'CGF-PLS09', 'CGF-PLS13', 'CIV-175-09',  
      'CIV-PLS09', 'CIV-PLS13', 'ECA-175-09', 'ECA-PLS09',  
      'ECA-PLS13', 'ECA9-OPT', 'ELE-175-09', 'ELE-PLS09',  
      'ELE-PLS13', 'IND-175-09', 'IND-PLS09', 'IND-PLS13',  
      'INT9-PROYT', 'INT9CONSTR', 'INT9ELECTR', 'INT9G-ECO',  
      'INT9MECANI', 'INT9ORTERR', 'INT9RNYMA', 'INT9SANEHI',  
      'INT9SDIGYT', 'INT9TRANSP', 'MCT-175-09', 'MCT-PLS09',  
      'MCT-PLS13', 'MCT9-OPT', 'MEC-175-09', 'MEC-PLS09',  
      'MEC-PLS13', 'MEC9-OPT', 'TIVAS'], dtype=object)
```

CIV-175-09
CIV-PLS09
CIV-PLS13

```
: np.unique(eca["Cod.Car.Sec"])  
: array(['ELE-175-09', 'ELE-PLS09', 'ELE-PLS13'], dtype=object)
```


CALCULO 4
 DINAMICA
 ESTRUCTURAS 1
 FISICA 4
 MECANICA DE MATERIALES 1
 Cantidad de elementos (8848, 23)

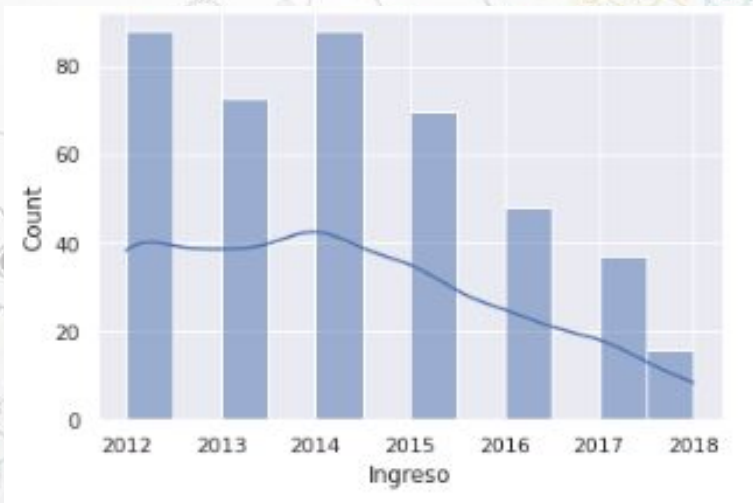
CALCULO 4
 CALCULO 5
 DINAMICA
 FISICA 4
 MECANICA DE MATERIALES 1
 Cantidad de elementos (6140, 23)

	Alumnos	Ingreso	Duracion
0	es_1164	2012	4
1	es_1185	2012	3
2	es_1204	2012	8
3	es_1244	2012	5
4	es_1253	2012	4
...
415	es_4087	2018	2
416	es_4104	2018	2
417	es_4134	2018	2
418	es_4156	2018	2
419	es_4189	2018	2

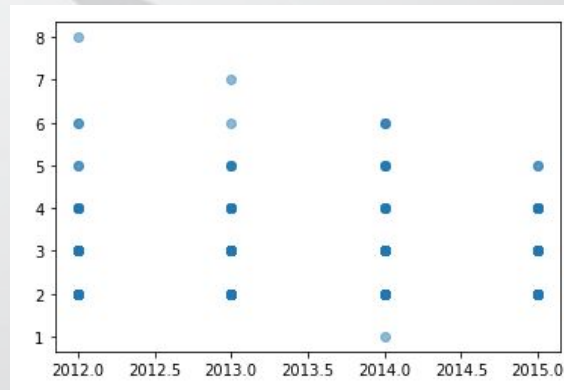
420 rows x 3 columns

```
estudiantes=np.unique(df['id_anony'])
materias=np.unique(df['Asignatura'])
print(len(estudiantes))
```

968



	Alumnos	Ingreso	Duracion
0	es_1120	2012	6
1	es_1243	2012	4
2	es_1261	2012	4
3	es_1298	2013	3
4	es_1301	2012	6
...
246	es_4040	2018	2
247	es_4084	2018	2
248	es_4101	2018	2
249	es_4179	2018	2
250	es_770	2012	4




```
Train Data shape: (378, 1)
test Data shape: (42, 1)
Train labels shape: (378,)
test labels shape: (42,)
```

```
Coeficiente de determinacion 0.18393233172103607
b0: 663.0921736215854
b1: [-0.32755153]
```

```
Error cuadratico medio: 1.6431632056465548
```

```
prediccion = lm.predict(X_test)
prediccion
```

```
array([3.73093682, 2.74828222, 4.05848835, 3.73093682, 3.73093682,
       3.40338529, 4.05848835, 3.07583375, 3.40338529, 3.07583375,
       4.05848835, 3.73093682, 3.73093682, 3.40338529, 2.74828222,
       3.40338529, 3.73093682, 2.09317915, 2.74828222, 3.07583375,
       4.05848835, 4.05848835, 3.07583375, 4.05848835, 2.74828222,
       2.42073069, 3.07583375, 3.73093682, 3.73093682, 3.40338529,
       2.42073069, 4.05848835, 3.07583375, 3.07583375, 3.73093682,
       2.74828222, 4.05848835, 3.73093682, 3.73093682, 2.42073069,
       4.05848835, 3.07583375])
```

Cuanto tarda en promedio un alumno en aprobar cursos básicos?

```
3.38778759366996
```


Regresión Lineal (Electromecánica)

```
Datos=pd.DataFrame({'Alumnos':Alumnos,'Ingreso':Inicio,'Duracion':Duracion})
Datos
```

	Alumnos	Ingreso	Duracion
0	es_1120	2012	6
1	es_1243	2012	4
2	es_1261	2012	4
3	es_1298	2013	3
4	es_1301	2012	6
...
246	es_4040	2018	2
247	es_4084	2018	2
248	es_4101	2018	2
249	es_4179	2018	2
250	es_770	2012	4

251 rows x 3 columns

```
DatosIngreso=Datos[ (Datos["Ingreso"]>=2012) & (Datos["Ingreso"]<=2015)]
DatosIngreso.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 204 entries, 0 to 250
Data columns (total 3 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Alumnos     204 non-null   object
1   Ingreso     204 non-null   int64
2   Duracion    204 non-null   int64
dtypes: int64(2), object(1)
memory usage: 6.4+ KB
```

```
Train Data shape: (183, 1)
test Data shape: (21, 1)
Train labels shape: (183,)
test labels shape: (21,)
```

Con 10% de test

```
Coeficiente de determinacion 6.0217176670929184e-05
b0: 18.699918115697617
b1: [-0.00775493]
```

3.0829760929948784

Regresión Logística (CIVIL)

```
daf=civil[(civil["Aprobado"]=="S")|(civil["Aprobado"]=="N") ]
```

```
estudiantes=np.unique(daf['id_anony'])  
materias=np.unique(daf['Asignatura'])  
print(len(estudiantes))
```

2006



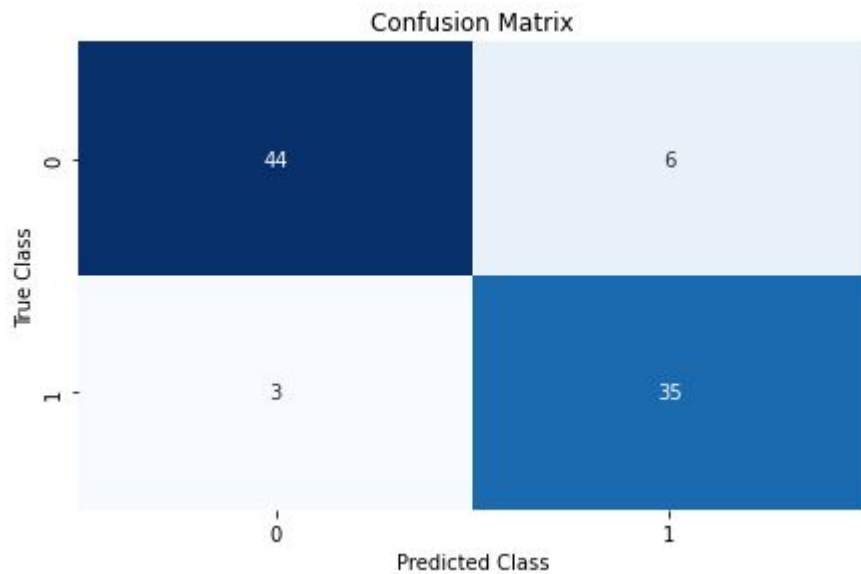
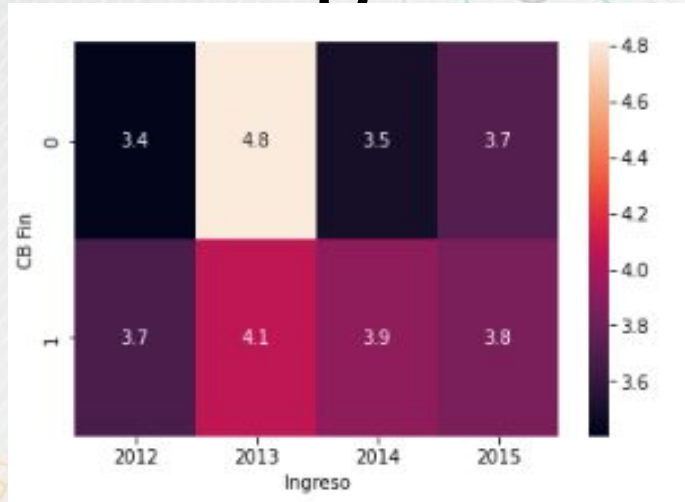
	Alumnos	Ingreso	Duracion	Materias	CB Fin
0	es_104	2012	9	14	0
1	es_1102	2012	1	11	0
2	es_1104	2012	4	22	1
3	es_1108	2012	7	11	0
4	es_1114	2012	4	20	1

```
Train Data shape: (1368, 3)  
test Data shape: (153, 3)  
Train labels shape: (1368,)  
test labels shape: (153,)
```

En un lapso de 8 años de 153 alumnos cuantos aprobaron Cursos Basicos

```
La cantidad de alumnos: (153,)  
Alumnos que no aprobaron: 105  
Alumnos que aprobaron: 48
```


Regresión Logística (Electromecánica)



```
X_train,X_test,Y_train,Y_test=train_test_split(X,Y,test_size=0.10)
print("Train Data shape: " + str(X_train.shape))
print("test Data shape: " + str(X_test.shape))
print("Train labels shape: " + str(Y_train.shape))
print("test labels shape: " + str(Y_test.shape))
```

Train Data shape: (789, 3)
test Data shape: (88, 3)
Train labels shape: (789,)
test labels shape: (88,)

```
] print("Precision del modelo:",metrics.accuracy_score(Y_test,Y_pred))
```

Precision del modelo: 0.8977272727272727

