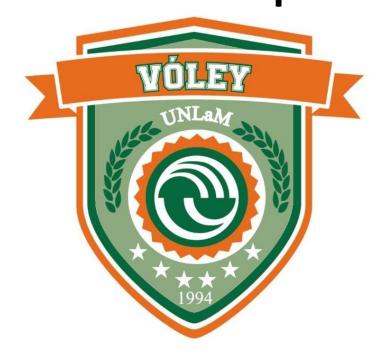


Index

- Project 1: Volleyball anual report National University of La Matanza
 - Tools: Excell, Powerpoint, Jamovi (R)
 - Dataset: Real volleyball data
- Project 2: Edith Cowan University Master of Exercise Science
 - Tools: Excel, Visual Basic
 - Dataset: ECU Monitoring unit assignment
- Project 3: Big Data Data Science course
 - Tools: Google Sheets, Google Looker Studio, Python, Deep Note
 - Datasets: "Services", "Internet", "Exams"

Project 1: University of La Matanza 2023 Anual Report



Volleyball D1 – Female

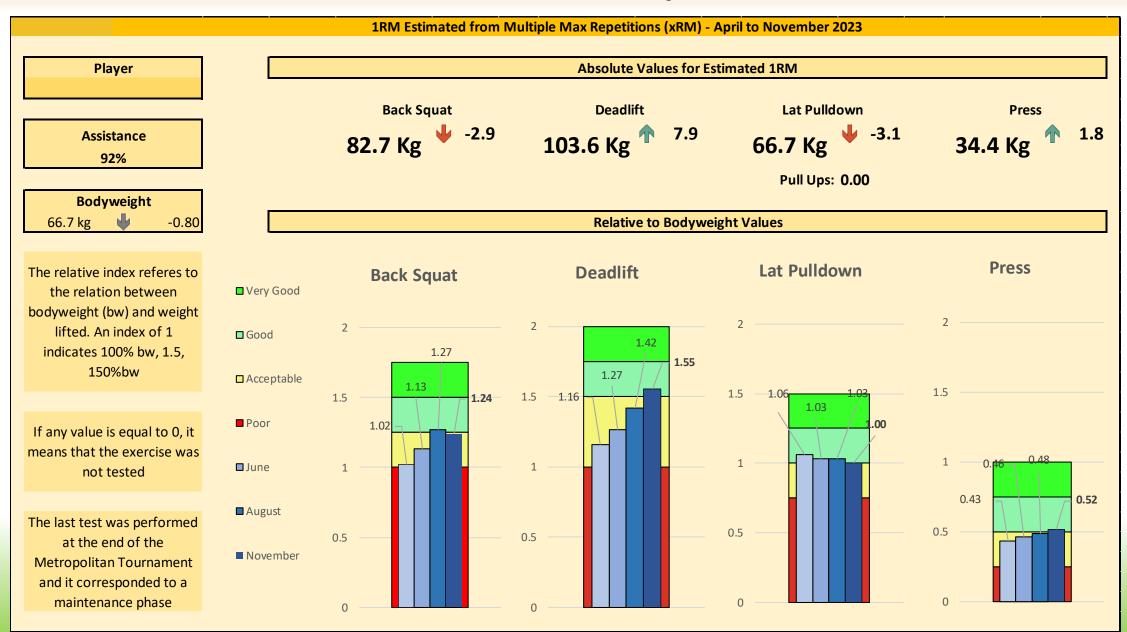


Personal Report





Personal Report



Backroom

Noviembre 2023

_					
-	ı	ı	ρ	rz	3

Despegue

Press

Tiron

Sent

												_	
	Atleta	~	PC kg ▼	S 1RI\ ▼	S rel ▼	TP 1RI	TP rel ▼	D 1RI ▼	D rel ▼	P 1RN ▼	P rel ▼	Dominada ▼	Asistencia▼
C			60.2				0.00				0.00	no	77%
C			66.7	82.7	1.24	66.7	1.00	103.6	1.55	34.4	0.52	si	92%
D			69.9	88.0	1.26	76	1.09	111.6	1.60	36.4	0.52	si	99%
G			72	91.9	1.28	55.6	0.77	90	1.25	34.0	0.47	no	76%
L			61.3		0.00		0.00		0.00		0.00		78%
L			64.4	81.5	1.27	74.9	1.16	87.9	1.36	36.1	0.56	si	91%
N			59.1	71.0	1.20	59.8	1.01	88	1.49	31.3	0.53	no	98%
R			61.3	83.3	1.36	64.5	1.05	92.4	1.51	30.1	0.49	si	86%
S			61.2	100.6	1.64	79.8	1.30	98.3	1.61	40.0	0.65	si	97%
S			89	103.5	1.16	73.7	0.83	151.8	1.71	39.2	0.44	no	88%
V			79.5	104.4	1.31	81.4	1.02	132.7	1.67	40.0	0.50	no	94%
F			72.5	98.9	1.36	84.3	1.16	111.1	1.53	40.0	0.55	si	87%
V			63.9	88.4	1.38	63.9	1.00	83.3	1.30	34.9	0.55	si	
Λ			79.5	103.3	1.30	77.8	0.98	105.3	1.32	33.3	0.42	no	90%

Personal Report - Anthropometrics

Atlhete:

mucle-bone index

Ideally, we are looking for the green zone and as far as the red zone as possible. The cut-off point for female university volleyball players is 3.8 (arbitrary units)

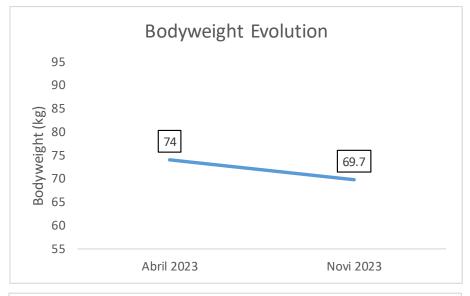
sum of 6 skinfolds

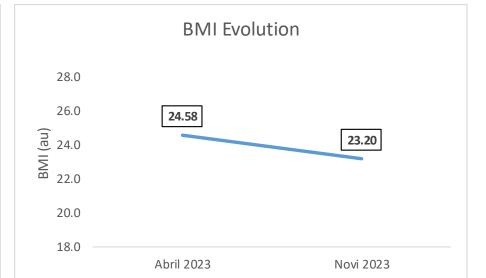
In this case, we aim to be below the 85 milimeters mark, indicated by the green zone

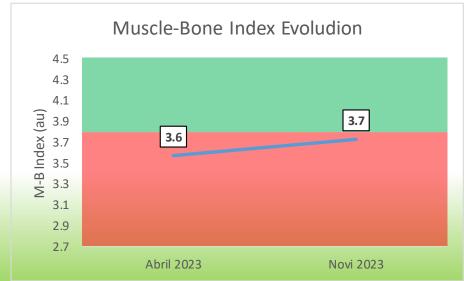
Important!!!

None of the markers indicated on the graphs is a predictor of performance

Anthropometric Data Evolution during 2023



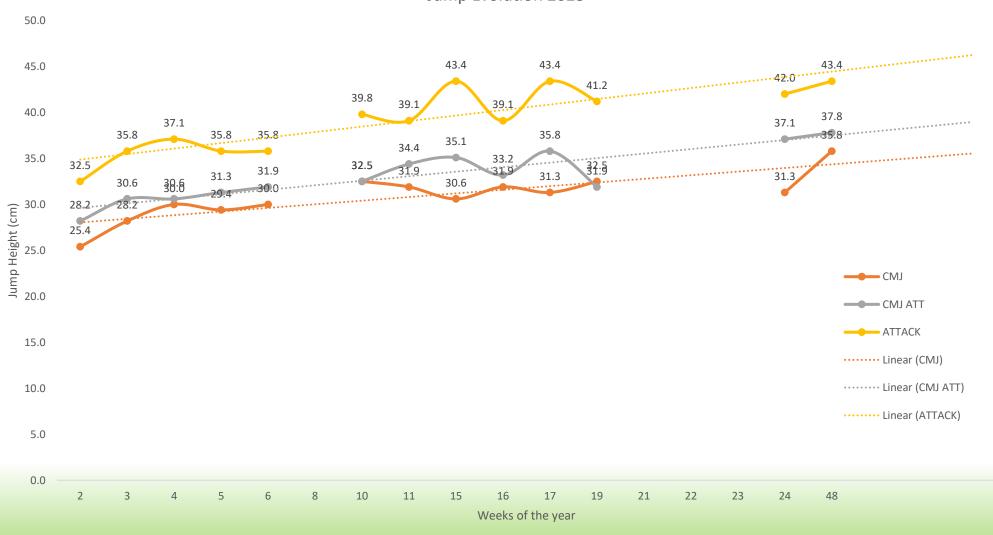






Personal Report - Jumps

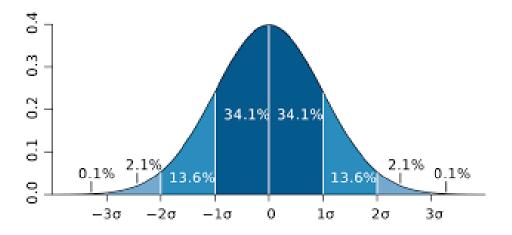
Jump Evolution 2023



Z-Scores

Z-score is a statistical measurement that describes a value's relationship to the mean of a group of values. Z-score is measured in terms of standard deviations from the mean. If a Z-score is 0, it indicates that the data point's score is identical to the mean score.

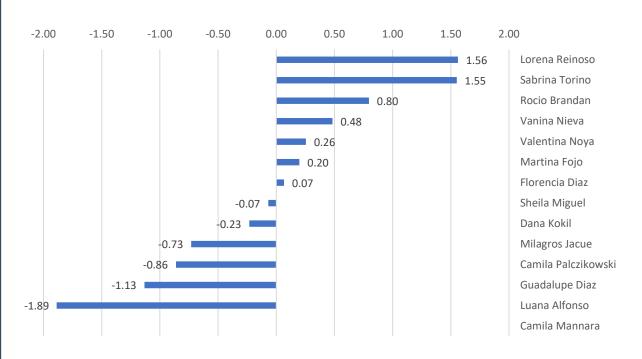
It is important for us so we can understand where our players are situated in relation to the group. On the following cases, the relative-to-bodyweight values were compared for the selected exercises.



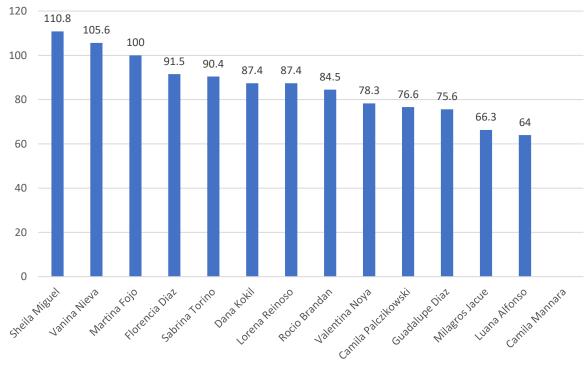


Squat

Relative Z-Score - Squat



Estimated 1RM - Squat





Backroom

Peso Corporal		Sentadillas				Tiron Palmar				Despegue			Press			
Testeo 💌	Atleta ■ PC ■	Testeo	Atleta	O UDM	S rel	Testeo	Atleta		TP rel	Testeo Atleta	1011	D rel	Testeo 💌	Atleta •	P IRI ■	P re ▼
1ro - Abril C	55.00	1ro - Abril	С			1ro - Abril C		53.30	1.06	1ro - Abril C			1ro - Abril	С	26.00	0.47
1ro - Abril C	69.20	1ro - Abril	C	70.70	1.23	1ro - Abril C		73.30	0.99	1ro - Abril C	80.30	0.00	1ro - Abril	С	29.80	0.43
1ro - Abril E	74.00	1ro - Abril		90.70	1.15	1ro - Abril E		73.30	0.76	1ro - Abril 🛛			1ro - Abril	D	30.50	0.41
1ro - Abril C	68.50	1ro - Abril	G	78.80	1.35	1ro - Abril C		52.30	1.07	1ro - Abril G				G	29.70	0.43
1ro - Abril L	59.40	1ro - Abril	L	80.00	1.02	1ro - Abril L		63.30	0.93	1ro - Abril L	108.00			Li	30.40	0.51
1ro - Abril L	64.20	1ro - Abril	L	65.20	0.00	1ro - Abril L		60.00	0.95	1ro - Abril L	88.00			Li	27.30	0.43
1ro - Abril N	56.90	1ro - Abril	M			1ro - Abril N	Y	54.20	0.85	1ro - Abril M	64.40			M	25.60	0.45
1ro - Abril F	61.40	1ro - Abril	B	69.80	1.42	1ro - Abril F		52.00	1.26	1ro - Abril R	84.40			B	25.30	0.41
1ro - Abril S	63.70	1ro - Abril	S	90.40	0.78	1ro - Abril S		80.00	0.65	1ro - Abril S	95.60	0.00		S	34.00	0.53
1ro - Abril S	100.00	1ro - Abril	S	77.90	1.21	1ro - Abril S		65.20	0.81	1ro - Abril S	445.00	400		S		
1ro - Abril S	88.20	1ro - Abril	S	106.60	1.09	1ro - Abril S		71.40	1.01	1ro - Abril S 1ro - Abril E	115.20			S	20.40	0.51
1ro - Abril F	71.50	1ro - Abril	F	77.90 78.30	1.29	1ro - Abril F	•	72.50	1.07	110 110111 1	85.90			FI	36.40 30.40	0.51
1ro - Abril \	60.90	1ro - Abril	V	91.80	0.00	1ro - Abril \ 1ro - Abril \	l	65.20 82.30	0.00	1ro - Abril V 1ro - Abril V	64.40 118.20			V	35.90	0.50 0.46
	77.80 76.20	1ro - Abril	M	31.00	0.00			02.30	0.00	1ro - Abril V 1ro - Abril M	110.20	0.00		V. M	39.30	0.46
1ro - Abril N 2do - Junio C	58.00	1ro - Abril 2do - Junio			1,13	1ro - Abril N 2do - Junio C		56.30	1.03	2do - Junio C		1.27	1ro - Abril 2do - Junio		28.00	0.48
2do - Junio C	67.50	2do - Junio	C	76.60	1.22	2do - Junio C		69.80	1.01	2do - Junio C	85.50		2do - Junio		31.30	0.46
2do - Junio C	71.50	2do - Junio	_	87.40	1.10	2do - Junio E	•	72.30	0.81	2do - Junio C	96.60		2do - Junio		32.90	0.46
2do - Junio C	68.90	2do - Junio		75.60	1.47	2do - Junio C		55.60	1.14	2do-Junio G	30.00	1.82	2do - Junio		30.20	0.44
2do - Junio L	59.40	2do - Junio		87.40	0.99	2do - Junio L	•	67.90	1.12	2do-Junio L	108.00		2do - Junio		34.70	0.58
2do - Junio L	64.50	2do - Junio		64.00	1.15	2do - Junio L	•	72.50	0.94	2do-Junio L	91.80	1.35	2do - Junio		30.00	0.47
2do - Junio I	57.50	2do - Junio		66.30	1.37	2do - Junio II	·	54.20	1.02	2do Junio M	77.40		2do - Junio		25.90	0.45
2do - Junio F	61.90	2do - Junio		84.50	1.47	2do - Junio F		63.40	1.47	2do-Junio R	93.30		2do - Junio		26.60	0.43
2do - Junio S	61.50	2do - Junio		90.40	1.24	2do - Junio S		90.40	0.78	2do-Junio S	96.40		2do - Junio		38.00	0.62
2do - Junio S	89.00	2do - Junio		110.80	1.32	2do Junio S		69.80	1.06	2do Junio S	133.30		2do - Junio		32.90	0.37
2do - Junio V	79.90	2do - Junio		105.60	1.26	2do - Junio N		84.40	1.20	2do-Junio V	144.90			v	38.30	0.48
2do - Junio F	72.40	2do - Junio	F	91.50	1.29	2do - Junio F		86.90	1.01	2do Junio F	109.30		2do - Junio	_	38.30	0.53
2do - Junio N	60.70	2do - Junio	İV	78.30	1.28	2do Junio V		61.10	0.89	2do Junio V	72.30		2do - Junio	V	31.30	0.52
2do - Junio N	78.00	2do - Junio	N	100.00	0.00	2do - Junio N		69.80	0.97	2do - Junio M		0.00	2do - Junio	M	27.80	0.36
3ro - Agosto C	58.00	3ro - Agosto	Īċ.		1.27	3ro - Agosto (56.30	1.03	3ro - Agosto C		1.42	3ro - Agosto	C.	28.00	0.48
3ro - Agosto (67.50	3ro - Agosto	lč	85.60	1.26	3ro - Agosto (69.80	1.05	3ro - Agosto C	95.70		3ro - Agosto	C	32.60	0.48
3ro - Agosto [71.50	3ro - Agosto	lō.	90.00	1.35	3ro - Agosto		75.30	0.81	3ro - Agosto □	116.90		_	Ď	36.10	0.50
3ro - Agosto (68.90	3ro - Agosto	i c	93.30	2.16	3ro - Agosto (55.60	1.14	3ro - Agosto €	84.40	1.94	_	G	34.40	0.50
3ro - Agosto L	59.40	3ro - Agosto	L	128.10	0.99	3ro - Agosto L		67.90	1.12	3ro - Agosto L	115.20	1.42	3ro - Agosto	Li	34.70	0.58
3ro - Agosto L	64.50	3ro - Agosto	L	64.00	1.34	3ro - Agosto L		72.50	0.94	3ro - Agosto L	91.80	1.54	3ro - Agosto	Li	30.00	0.47
3ro - Agosto N	57.50	3ro - Agosto	M	77.20	1.37	3ro - Agosto N	V	54.20	1.02	3ro - Agosto M	88.60	1.51	3ro - Agosto	M	25.90	0.45
3ro - Agosto F	61.90	3ro - Agosto	F	84.50	1.47	3ro - Agosto F		63.40	1.47	3ro - Agosto R	93.30	1.57	3ro - Agosto	В	26.60	0.43
3ro - Agosto S	61.50	3ro - Agosto	s	90.40	1.29	3ro - Agosto S		90.40	0.78	3ro - Agosto S	96.40	1.60	3ro - Agosto	S	38.00	0.62
3ro - Agosto S	89.00	3ro - Agosto	s	115.00	1.32	3ro - Agosto 9		69.80	1.06	3ro - Agosto S	142.50	1.81	3ro - Agosto	S	37.40	0.42
3ro - Agosto ∖	79.90	3ro - Agosto	V	105.60	1.37	3ro - Agosto N		84.40	1.16	3ro - Agosto V	144.90	1.52	3ro - Agosto	V.	38.30	0.48
3ro - Agosto F	72.40	3ro - Agosto	F	98.90	1.44	3ro - Agosto F		84.20	1.05	3ro - Agosto F	109.80	1.39	3ro - Agosto	FI	43.40	0.60
3ro - Agosto ∖	60.70	3ro - Agosto	V	87.20	1.48	3ro - Agosto N	ı	63.90	0.89	3ro - Agosto V	84.40	1.42	3ro - Agosto	V.	34.40	0.57
3ro - Agosto N	78.00	3ro - Agosto	M	115.20	0.00	3ro - Agosto N	Y	69.80	0.00	3ro - Agosto M	110.50	0.00	3ro - Agosto	M	31.30	0.40
4to - Noviembre (60.20	4to - Noviembr	e C		1.24	4to - Noviembre (1.00	4to - Noviembre C		1.55	4to - Noviembre	С		0.00
4to - Noviembre (: 66.70	4to - Noviembr	e C	82.70	1.26	4to - Noviembre (66.70	1.09	4to - Noviembre C	103.60		4to - Noviembre	С	34.40	0.52
4to - Noviembre [69.90	4to - Noviembr		88.00	1.33	4to - Noviembre [76.00	0.81	4to - Noviembre 🛘	111.60		4to - Noviembre		36.40	0.52
4to - Noviembre (68.90	4to - Noviembr	e C	91.90	0.00	4to - Noviembre (55.60	0.00	4to - Noviembre C	90.00	0.00	4to - Noviembre	G	34.00	0.49
4to - Noviembre L	61.30	4to - Noviembr	_		1.27	4to - Noviembre L			1.16	4to - Noviembre L		1.36	4to - Noviembre	_		0.00
4to - Noviembre L	64.40	4to - Noviembr	<u> </u>	81.50	1.20	4to - Noviembre L		74.90	1.01	4to - Noviembre L	87.90		4to - Noviembre	_	36.10	0.56
4to - Noviembre N	59.10	4to - Noviembr		71.00	1.36	4to - Noviembre N	Y	59.80	1.05	4to - Noviembre №	88.00		4to - Noviembre		31.30	0.53
4to - Noviembre F	61.30	4to - Noviembr		83.30	1.64	4to - Noviembre F		64.50	1.30	4to - Noviembre R	92.40		4to - Noviembre		30.10	0.49
4to - Noviembre S	61.20	4to - Noviembr		100.60	1.16	4to - Noviembre S		79.80	0.83	4to - Noviembre S	98.30		4to - Noviembre		40.00	0.65
4to - Noviembre S	89.00	4to - Noviembr	-	103.50	1.31	4to - Noviembre S		73.70	1.02	4to - Noviembre S	151.80		4to - Noviembre		39.20	0.44
4to - Noviembre \	79.50	4to - Noviembr	-	104.40	1.36	4to - Noviembre \	ı		1.16	4to - Noviembre V	132.70		4to - Noviembre	-	40.00	0.50
4to - Noviembre F	72.50	4to - Noviembr	e F	98.90	1.38	4to - Noviembre F		84.30	1.00	4to - Noviembre F	111.10	1.30	4to - Noviembre	FI	40.00	0.55

Relative			
quat			
			1.47
			1.47
			1.37
			1.32
			1.29
			1.28
			1.26
			1.24
			1.22
	П		1.15
	sŀ		1.13
			1.10
			0.99
		N/A	

Team Evolution





Variables Data Analyses Edit







1 1 2022

Bondy V S Filtors 0



















	🚱 Equipo	Fecha	∂ a Atleta	PC kg	Sent 1RM	Sent rel	i 🎸 Ti
1	UNLaM	Abril 2023	Camila Mann	55.0			_
2	UNLaM	Abril 2023	Camila Palczi	69.2	70.7	1.02	
3	UNLaM	Abril 2023	Dana Kokil	74.0	90.7	1.23	
4	UNLaM	Abril 2023	Guadalupe Di	68.5	78.8	1.15	
5	UNLaM	Abril 2023	Lorena Reinoso	59.4	80.0	1.35	
6	UNLaM	Abril 2023	Luana Alfonso	64.2	65.2	1.02	
7	UNLaM	Abril 2023	Milagros Jacue	56.9			
8	UNLaM	Abril 2023	Rocio Brandan	61.4	69.8	1.14	
9	UNLaM	Abril 2023	Sabrina Torino	63.7	90.4	1.42	
10	UNLaM	Abril 2023	Sara Agüero	100.0	77.9	0.78	
11	UNLaM	Abril 2023	Sheila Miguel	88.2	106.6	1.21	
12	UNLaM	Abril 2023	Florencia Diaz	71.5	77.9	1.09	
13	UNLaM	Abril 2023	Valentina Noya	60.9	78.3	1.29	
14	UNLaM	Abril 2023	Vanina Nieva	77.8	91.8	1.18	
15	UNLaM	Abril 2023	Martina Fojo	76.2			
16	UNLaM	Junio 2023	Camila Mann	58.0			
17	UNLaM	Junio 2023	Camila Palczi	67.5	76.6	1.13	
18	UNLaM	Junio 2023	Dana Kokil	71.5	87.4	1.22	
19	UNLaM	Junio 2023	Guadalupe Di	68.9	75.6	1.10	
20	UNLaM	Junio 2023	Lorena Reinoso	59.4			
21	UNLaM	Junio 2023	Luana Alfonso	64.5			
22	UNLaM	Junio 2023	Milagros Jacue	57.5	66.3	1.15	
23	UNLaM	Junio 2023	Rocio Brandan	61.9	84.5	1.37	
24	UNLaM	Junio 2023	Sabrina Torino	61.5	90.4	1.47	
25	UNLaM	Junio 2023	Sheila Miguel	89.0	110.8	1.24	
26	UNLaM	Junio 2023	Florencia Diaz	79.9	105.6	1.32	
27	UNLaM	Junio 2023	Valentina Noya	72.4	91.5	1.26	
28	UNLaM	Junio 2023	Vanina Nieva	60.7	78.3	1.29	
							-

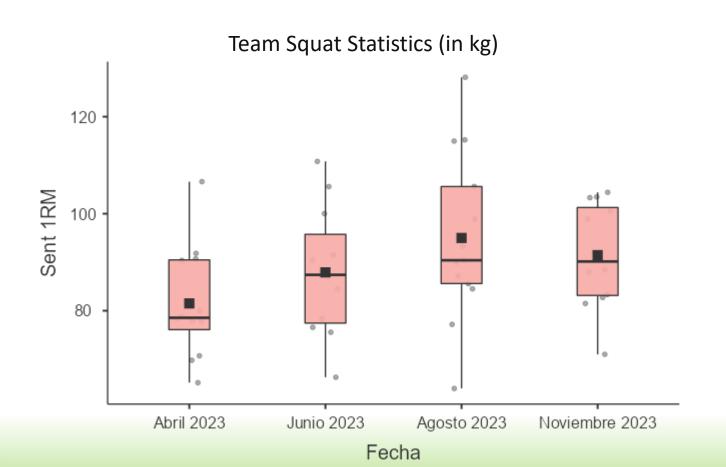
Results

Descriptives

Descriptives

	Fecha	PC kg	Sent 1RM	Despegue 1RM	Press 1RN
N	Abril 2023	15	12	10	12
	Junio 2023	14	11	9	11
	Agosto 2023	14	13	13	14
	Noviembre 2023	14	12	12	12
Missing	Abril 2023	0	3	5	3
	Junio 2023	0	3	5	3
	Agosto 2023	0	1	1	0
	Noviembre 2023	0	2	2	2
Mean	Abril 2023	69.8	81.5	90.4	30.1
	Junio 2023	67.9	87.9	101	32.8
	Agosto 2023	67.9	95.0	106	33.6
	Noviembre 2023	68.6	91.5	105	35.8
Std. error mean	Abril 2023	3.15	3.33	6.00	1.09
	Junio 2023	2.51	4.10	8.13	1.31
	Agosto 2023	2.51	4.82	5.59	1.33
	Noviembre 2023	2.40	3.11	5.85	0.992
95% CI mean lower bound	Abril 2023	63.6	75.0	78.7	28.0
	Junio 2023	63.0	79.9	85.1	30.2
	Agosto 2023	63.0	85.6	94.8	31.1
	Noviembre 2023	63.9	85.4	93.2	33.9
95% CI mean upper bound	Abril 2023	76.0	88.0	102	32.2
	Junio 2023	72.8	96.0	117	35.3

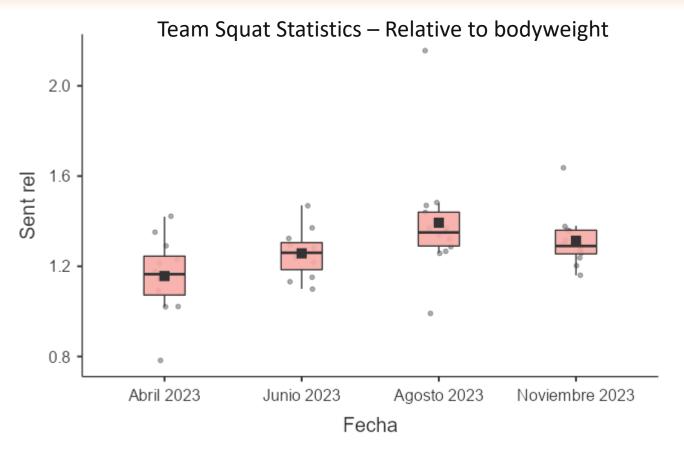
Squat

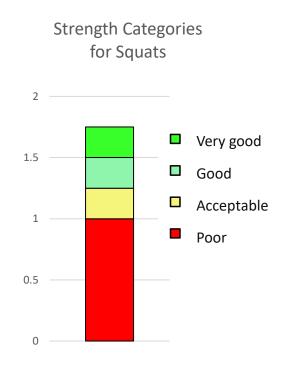


The team's average squat strength increased throughout the year during the strength and power phases of the sport calendar. After the maintenance phase, which coincided with the Metropolitan Tournament, the strength level for this exercise slightly decreased within the expected parameters. However, the team finished the year with higher lower body strength compared to the beginning of the year. This places them in a more advantageous situation to tackle 2024 and the upcoming sport demands.



Squat





The relative strength for the squat shows a similar trend compared to the absolute values. On average, the team is better equipped to start 2024 in the category considered 'Good', which is higher than 125% of their own bodyweight, compared to April 2023 when their values were 'Acceptable', or between 100% and 125% of their own weight.



Project 2: Edith Cowan University

Edith Cowan University Monitoring Unit (Dr. Greg Haff) - 2021







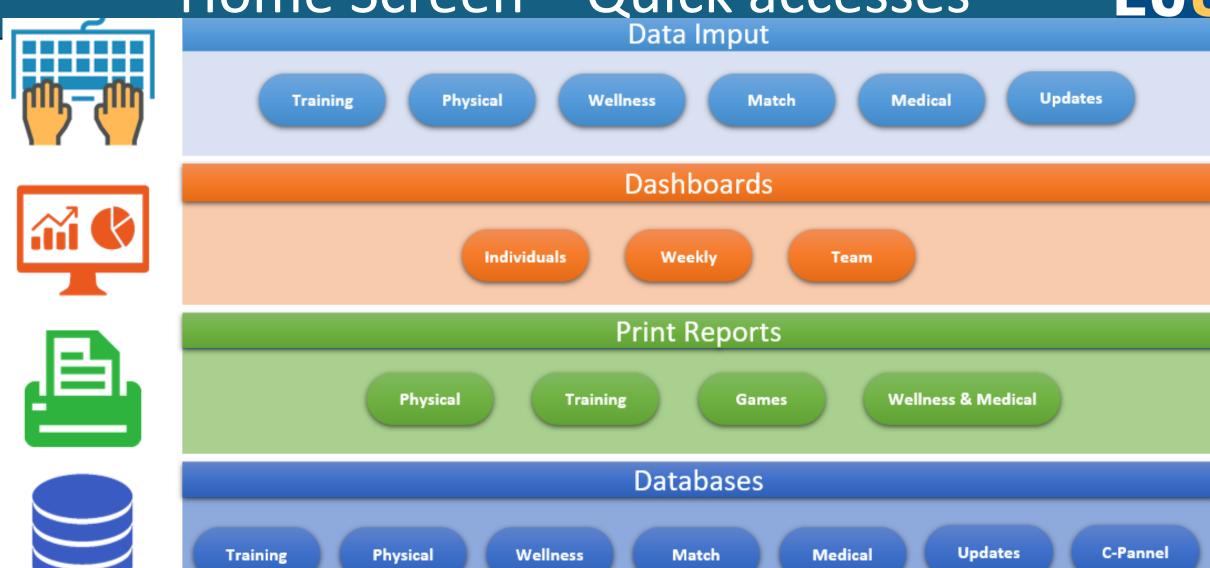
Excel



Visual Basic

Home Screen – Quick accesses

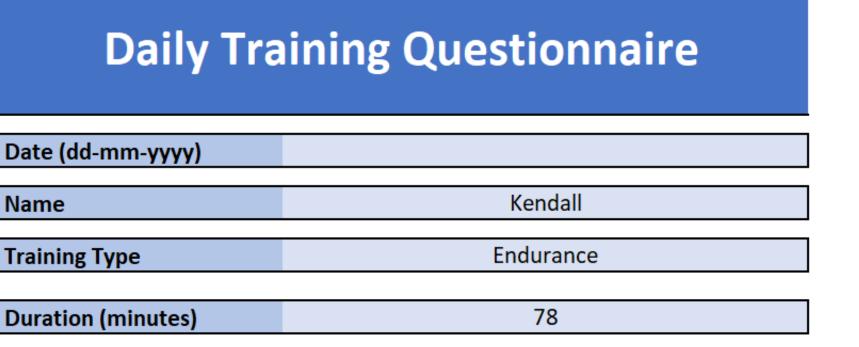




Home D.I.Train D.I.Phys D.I.Well D.I.Match D.I.Med D.I.Upd Dash Ind d.calc Dash Weekly w.calc Dash Team t.calc Phys Rep Phys-re

Data Input Tab: Training information



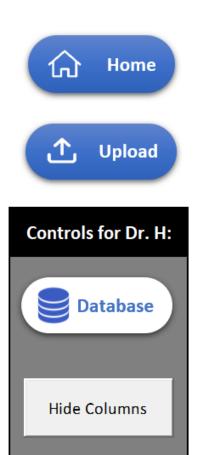


02 03 04 05 06 07 •8 09 010

1 - "aint notin' but a peanut"; 10 - "completely extenuating"

Training

Session RPE



Individual Dashboard



Controls

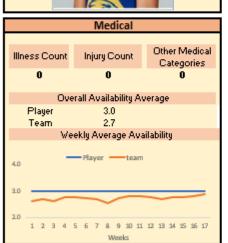
Select athlete to visualize data:

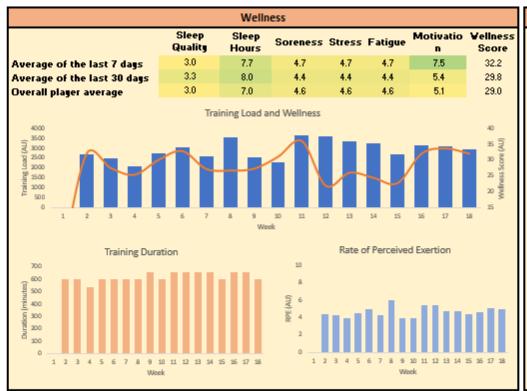
French

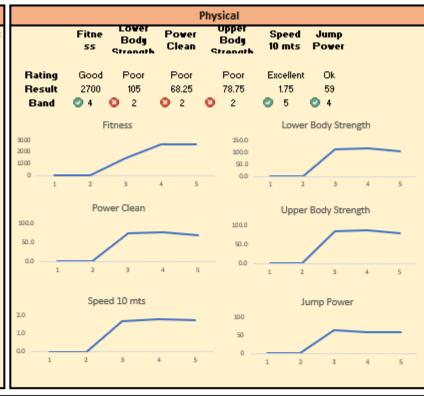


Performance Profile: French

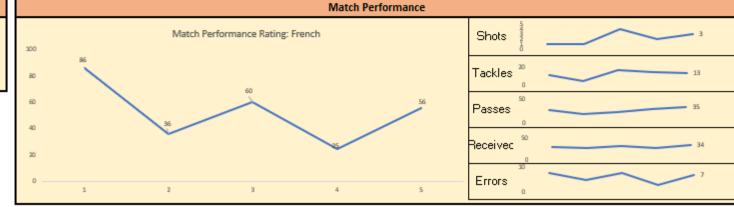








	Updates							
	ate	Priority	Area	Comment				
3040	4/2022	2	Performance	Good COD technique				
29/0	4/2022	2	Medical	Covid 19 positive				
27/0	4/2022	2	Performance	good passes session				
26/0	4/2022	2	Medical	Covid 19 positive				
22/0	4/2022	2	Performance	Didn't hit tackle # goal				



Weekly Dashboard





D.I.Upd

D.I.Match

Dash Weekly

w.calc

t.calc

Team Dashboard





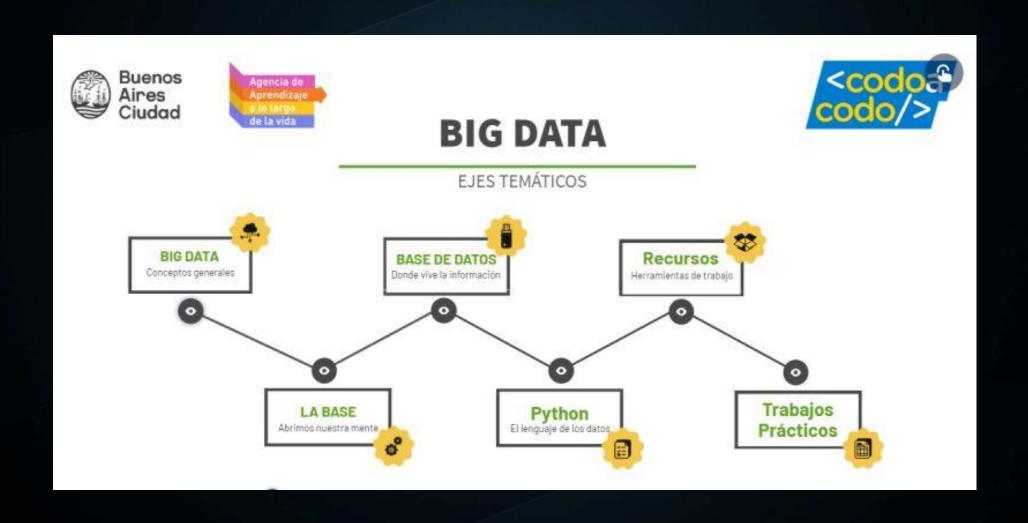
Wellness and Medical Report



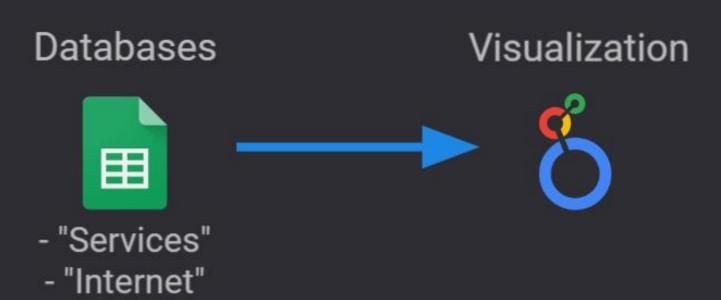
	ontrols ete to visualize				
Blitz					
elect athlete 1 2 3 4 5 6 7 8 9 10	and week to pri				



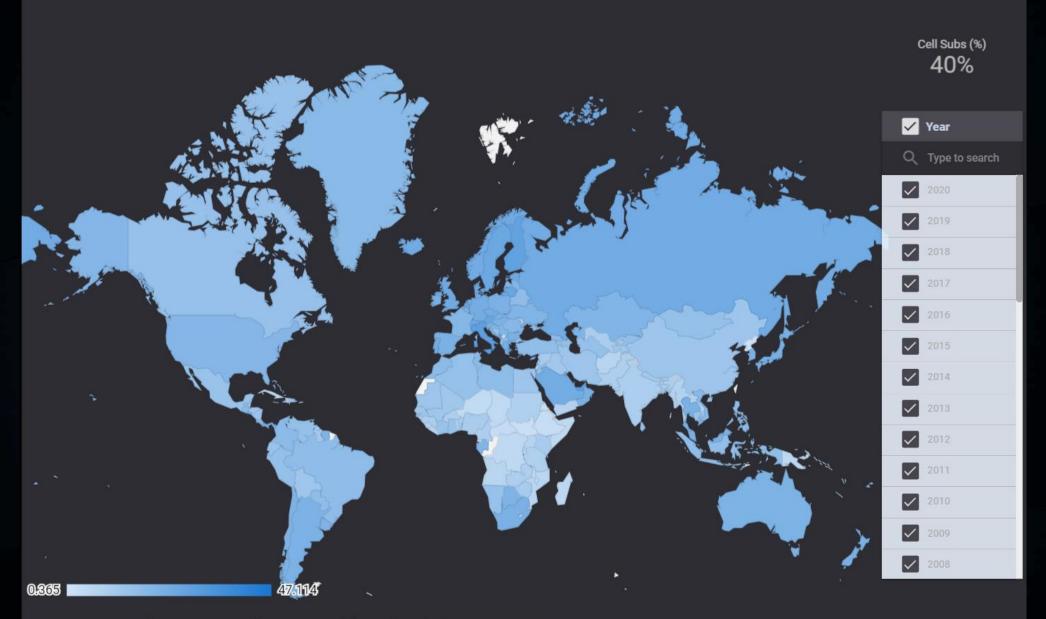
Project 3: Big Data – 2023



Sheets + Looker Studio

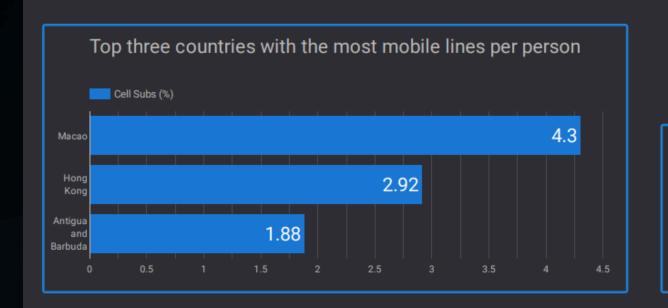


World mobile subscriptions percentage by country, from 1980 to 2020



Over 100% denotes more than one mobile subscription per person

Country with the most mobile lines per person in 2020: Macao

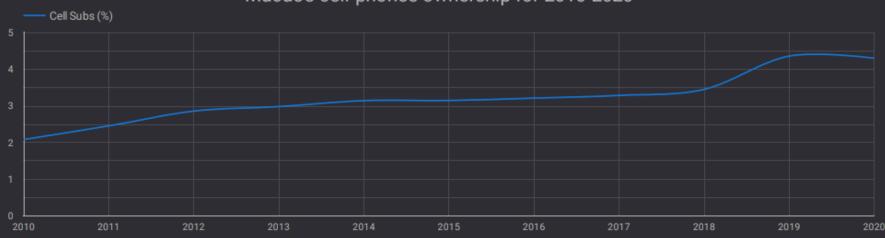




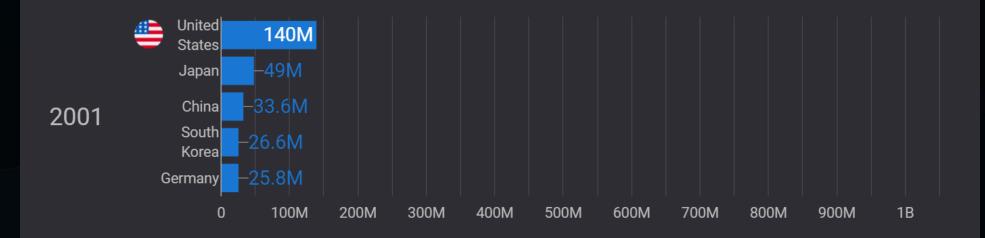
Macao's average mobile line's ownership per person:

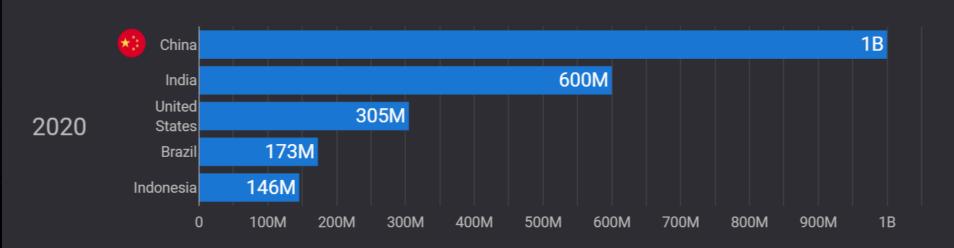
4.3

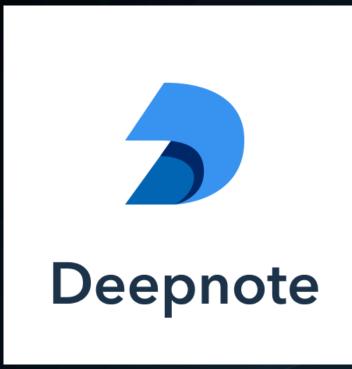
Macao's cell-phones ownership for 2010-2020



Top five countries with the highest number of internet users









Dataset: "exams"

Basic Operations

Review data types

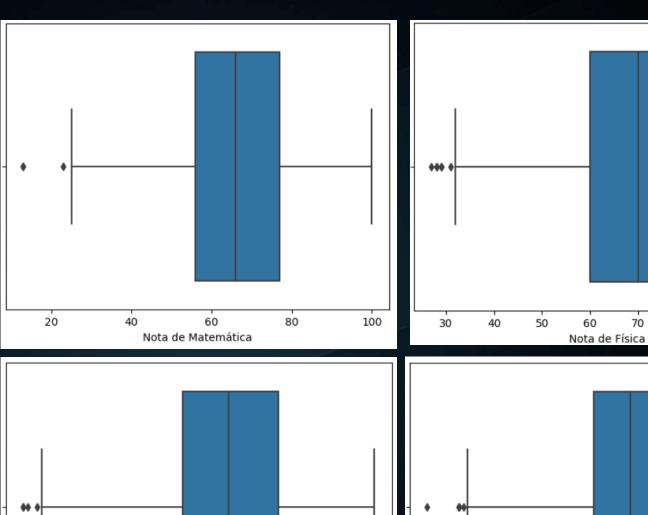
```
df.dtypes
                                 object
id
gender
                                 object
race/ethnicity
                                object
parental level of education
                                object
                                 object
lunch
                                object
employed
                                object
test preparation course
                                float64
math score
physics score
                                float64
                               float64
chemistry score
algebra_score
                                float64
dtype: object
```

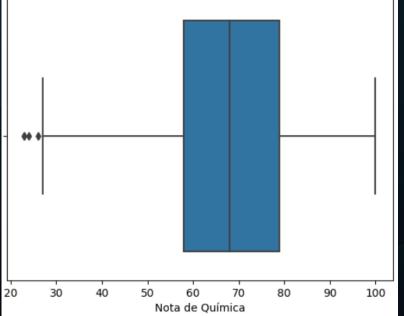
Drop duplicates

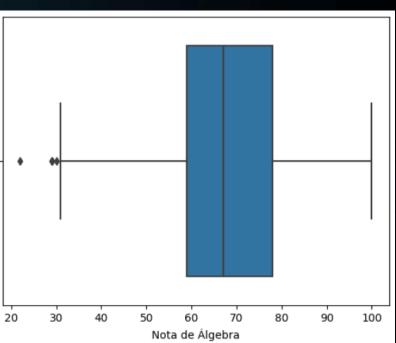
```
print(f'Original: {df.id.count()} filas')
   duplicate_rows_df = df[df.duplicated()] #"duplicated???"
   print(f'Cantidad de filas duplicadas: {duplicate_rows_df.id.count()}')
   #eliminar duplicados
   df = df.drop_duplicates()
   #print (df.head())
   # Filas despues de eliminar duplicados
   print(f'Final: {df.id.count()} filas')
11
Original: 1018 filas
Cantidad de filas duplicadas: 18
Final: 1000 filas
```

Detect outliers

```
#materias renombradas:
#"math score": "Nota de Matemática",
#"physics score": "Nota de Física",
#"chemistry score": "Nota de Química",
#"algebra_score": "Nota de Álgebra"
sns.boxplot(x=df['Nota de Matemática'])
plt.show()
sns.boxplot(x=df['Nota de Física'])
plt.show()
sns.boxplot(x=df['Nota de Química'])
plt.show()
sns.boxplot(x=df['Nota de Álgebra'])
plt.show()
```







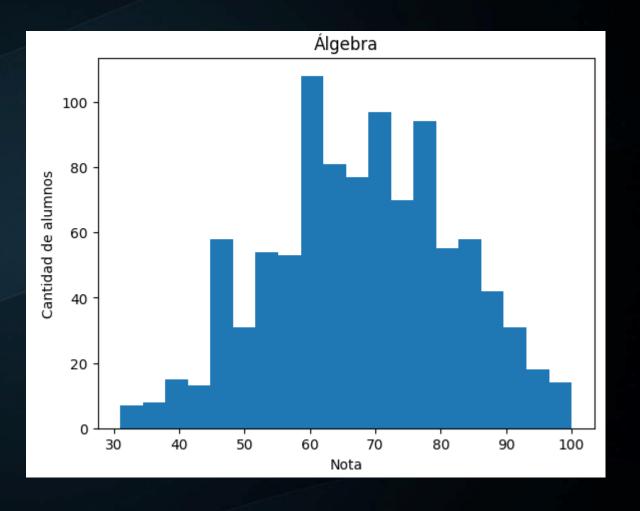
70

90

100

Frequencies: Plot histograms

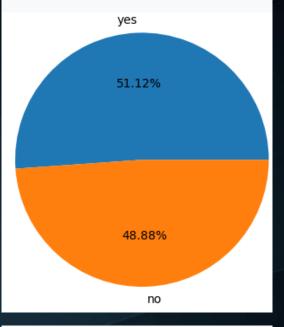
```
#"math score": "Nota de Matemática",
 2 #"physics score": "Nota de Física",
 3 #"chemistry score": "Nota de Química",
   #"algebra_score": "Nota de Álgebra"
   #Algebra
   plt.hist(df['Nota de Álgebra'], bins=20)
 8 plt.title("Álgebra")
   plt.ylabel("Cantidad de alumnos")
   plt.xlabel("Nota")
    plt.show()
12
   #Chemistry
   plt.hist(df['Nota de Química'], bins=20)
   plt.title("Química")
   plt.ylabel("Cantidad de alumnos")
   plt.xlabel("Nota")
   plt.show()
   #Math
   plt.hist(df['Nota de Matemática'], bins=20)
   plt.title("Matemática")
   plt.ylabel("Cantidad de alumnos")
   plt.xlabel("Nota")
   plt.show()
26
   #Physics
   plt.hist(df['Nota de Física'], bins=20)
   plt.title("Fisica")
   plt.ylabel("Cantidad de alumnos")
   plt.xlabel("Nota")
   plt.show()
33
34
```



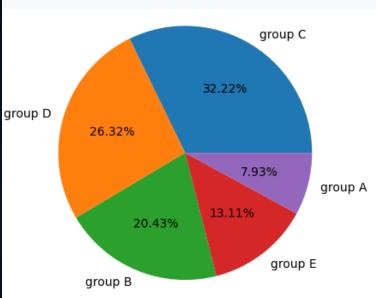
Categorical Values Exploration

```
1 #pandas.value_counts()--> devuelve usa sere con valores unic
3 #Nombres actualizados:
     "gender" : "Género",
5 # "race/ethnicity": "Etnia",
6 #"parental level of education": "Educación de los padres",
7 #"lunch": "Almuerzo",
8 #"employed": "Empleado/a",
9 #"test preparation course": "Curso preparatorio",
10
11 #torta: 'Género'
12 labels = df['Género'].value_counts().index
13 sizes = df['Género'].value_counts()
  plt.pie(sizes, labels=labels, autopct='%1.2f%%')
15 plt.title('Género')
16 plt.show()
17
   #torta: 'Etnia'
19 labels = df['Etnia'].value_counts().index
   sizes = df['Etnia'].value_counts()
   plt.pie(sizes, labels=labels, autopct='%1.2f%%')
22 plt.title('Etnia')
   plt.show()
24
25 #torta: 'Empleado/a'
  labels = df['Empleado/a'].value_counts().index
27 sizes = df['Empleado/a'].value_counts()
  plt.pie(sizes, labels=labels, autopct='%1.2f%%')
   plt.title('¿Está empleado?')
   plt.show()
31
   #torta: 'Curso preparatorio'
33 labels = df['Curso preparatorio'].value_counts().index
34 sizes = df['Curso preparatorio'].value_counts()
  plt.pie(sizes, labels=labels, autopct='%1.2f%%')
36 plt.title('¿Tomó el curso preparatorio?')
37 plt.show()
```

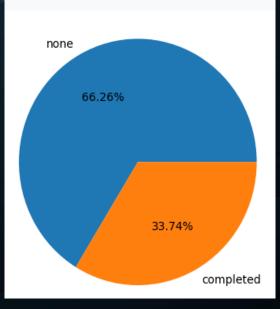
Employed?



Ethnic Groups



Prep Course Taken



Lunch Type

