# Week02\_Challenge1

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#### Librerías

```
import numpy as np
In [72]:
         import pandas as pd
         import matplotlib.pyplot as plt
In [73]: # Crear el dataframe con los datos proporcionados
         data = {
              'Attendance': [80, 65, 95, 95, 85, 75, 90, 65],
              'Homework': [75, 70, 85, 100, 65, 55, 90, 80],
              'Pass': ['yes', 'no', 'yes', 'yes', 'no', 'no', 'yes', 'yes'],
             'Reference': ['yes', 'no', 'yes', 'no', 'no', 'no', 'yes', 'no']
         df = pd.DataFrame(data)
         # Convertir la columna 'Pass' a valores binarios (1 para 'yes' y 0 para 'no')
         df['Pass'] = df['Pass'].apply(lambda x: 1 if x == 'yes' else 0)
         # Convertir la columna 'Reference' a valores binarios (1 para 'yes' y 0 para 'no')
         df['Reference'] = df['Reference'].apply(lambda x: 1 if x == 'yes' else 0)
         df
```

Out[73]:		Attendance	Homework	Pass	Reference
	0	80	75	1	1
	1	65	70	0	0
	2	95	85	1	1
	3	95	100	1	0
	4	85	65	0	0
	5	75	55	0	0
	6	90	90	1	1
	7	65	80	1	0

## Regresión Logística

Las interpretaciones de la regresión logística sin dividir los datos en entrenamiento y prueba se realizaron con los siguientes parámetros:

- theta0\_init = 0
- theta1\_init = 0
- alpha = 0.01

• iterations = 1000

```
In [74]: # Inicializar parámetros
                                   theta0_init = 0
                                    theta1_init = 0
                                    alpha = 0.001
                                    iterations = 100
In [75]: def sigmoid(z):
                                                  return 1 / (1 + np.exp(-z)) #h_0
                                    def logistic_regression(X, y, theta0_init, theta1_init, alpha, iterations):
                                                  theta0 = theta0_init
                                                  theta1 = theta1_init
                                                  n = len(y)
                                                  for i in range(iterations):
                                                                z = theta0 + theta1 * X
                                                                h = sigmoid(z)
                                                                 # Calcula los gradientes
                                                                 gradient0 = (1/n) * np.sum(h - y)
                                                                 gradient1 = (1/n) * np.sum((h - y) * X)
                                                                 # Actualiza los parámetros
                                                                theta0 -= alpha * gradient0
                                                                 theta1 -= alpha * gradient1
                                                                 # Calcula el costo
                                                                 cost = -1/n * np.sum(y * np.log(h) + (1 - y) * np.log(1 - h))
                                                                 print(f"Iteración {i+1}: theta0 = {theta0}, theta1 = {theta1}, alpha = {alp
                                                  return theta0, theta1
                                                                 #if (i+1) % 100 == 0:
                                                                                \#print(f"Iteración \{i+1\}: theta0 = \{theta0\}, theta1 = \{theta1\}, alpha = \{theta2\}, alpha = \{theta3\}, alpha = \{theta4\}, 
                                                  #return theta0, theta1
```

## Clasificador que estime si un estudiante aprueba o no el curso

#### **Característica Attendance**

```
In [76]: # Característica 'Attendance'
X_attendance = df['Attendance'].values
y = df['Pass'].values

# Aplicar La regresión logística
theta0_attendance, theta1_attendance = logistic_regression(X_attendance, y, theta0_
print(f"Parámetros finales para 'Attendance': theta0 = {theta0_attendance}, theta1
# Predicción para cada muestra en el conjunto de datos
z_attendance = theta0_attendance + theta1_attendance * X_attendance
# La función sigmoide convierte los valores en probabilidades, que estarán en el ray_pred_attendance = sigmoid(z_attendance) >= 0.5
```

```
Week02 Challenge1
Iteración 1: theta0 = 0.000125, theta1 = 0.0125, alpha = 0.001, costo = 0.69314718
05599453
Iteración 2: theta0 = 1.6774744101258067e-05, theta1 = 0.005731771382780238, alpha
= 0.001, costo = 0.6626468899832987
Iteración 3: theta0 = 2.7517936758486686e-05, theta1 = 0.008772057291519551, alpha
= 0.001, costo = 0.6488738740281816
Iteración 4: theta0 = -1.811919945807841e-05, theta1 = 0.007155822163269516, alpha
= 0.001, costo = 0.6467995990723467
Iteración 5: theta0 = -3.429325865979461e-05, theta1 = 0.007971518988365676, alpha
= 0.001, costo = 0.6461367044720583
Iteración 6: theta0 = -6.548650301519383e-05, theta1 = 0.007547052419716091, alpha
= 0.001, costo = 0.6459785607463455
Iteración 7: theta0 = -8.88939007099455e-05, theta1 = 0.007765355888775023, alpha
= 0.001, costo = 0.6459336950088266
Iteración 8: theta0 = -0.0001163073638280061, theta1 = 0.007652889517823268, alpha
= 0.001, costo = 0.6459215826568705
Iteración 9: theta0 = -0.00014165080485685857, theta1 = 0.007711316956683345, alph
a = 0.001, costo = 0.6459178159176626
Iteración 10: theta0 = -0.00016806142291422686, theta1 = 0.007681625481971009, alp
ha = 0.001, costo = 0.6459163212803277
Iteración 11: theta0 = -0.0001939210131491527, theta1 = 0.007697418480019387, alph
a = 0.001, costo = 0.6459154246120901
Iteración 12: theta0 = -0.00022006470074993617, theta1 = 0.0076897461640195895, al
pha = 0.001, costo = 0.6459146881079009
Iteración 13: theta0 = -0.0002460616079617348, theta1 = 0.007694182860554269, alph
a = 0.001, costo = 0.645913994120134
Iteración 14: theta0 = -0.00027213407405826526, theta1 = 0.007692371724135016, alp
ha = 0.001, costo = 0.6459133114746313
Iteración 15: theta0 = -0.0002981673769284784, theta1 = 0.007693784484466364, alph
a = 0.001, costo = 0.6459126318508952
Iteración 16: theta0 = -0.0003242207127504387, theta1 = 0.0076935337674039535, alp
ha = 0.001, costo = 0.6459119530364352
Iteración 17: theta0 = -0.00035026353712198016, theta1 = 0.0076941413900905335, al
pha = 0.001, costo = 0.6459112744417974
Iteración 18: theta0 = -0.0003763116106707815, theta1 = 0.007694306118696579, alph
a = 0.001, costo = 0.6459105959100955
Iteración 19: theta0 = -0.00040235680110797887, theta1 = 0.0076946993737810375, al
pha = 0.001, costo = 0.6459099173995517
Iteración 20: theta0 = -0.00042840330460258117, theta1 = 0.007694974710050427, alp
ha = 0.001, costo = 0.645909238899044
Iteración 21: theta0 = -0.00045444895598457875, theta1 = 0.0076953108889197454, al
pha = 0.001, costo = 0.645908560405611
Iteración 22: theta0 = -0.0004804948724551772, theta1 = 0.00769561567175937, alpha
= 0.001, costo = 0.6459078819184638
Iteración 23: theta0 = -0.000506540477553966, theta1 = 0.007695936652481039, alpha
= 0.001, costo = 0.645907203437393
Iteración 24: theta0 = -0.0005325860687274376, theta1 = 0.007696249273271815, alph
a = 0.001, costo = 0.6459065249623425
Iteración 25: theta0 = -0.0005586314924980507, theta1 = 0.007696566205618732, alph
a = 0.001, costo = 0.6459058464932972
Iteración 26: theta0 = -0.0005846768280587123, theta1 = 0.007696880911217172, alph
a = 0.001, costo = 0.6459051680302532
Iteración 27: theta0 = -0.00061072203454775, theta1 = 0.007697196763735274, alpha
= 0.001, costo = 0.6459044895732096
Iteración 28: theta0 = -0.0006367671330498643, theta1 = 0.00769751202241173, alpha
= 0.001, costo = 0.6459038111221658
Iteración 29: theta0 = -0.0006628121126863762, theta1 = 0.00769782758545211, alpha
= 0.001, costo = 0.6459031326771217
Iteración 30: theta0 = -0.0006888569790712073, theta1 = 0.007698142989396806, alph
a = 0.001, costo = 0.6459024542380774
Iteración 31: theta0 = -0.0007149017293083641, theta1 = 0.007698458473383317, alph
a = 0.001, costo = 0.6459017758050325
Iteración 32: theta0 = -0.0007409463648928281, theta1 = 0.007698773914020752, alph
```

a = 0.001, costo = 0.6459010973779874

```
Iteración 33: theta0 = -0.0007669908850539123, theta1 = 0.007699089374976688, alph
a = 0.001, costo = 0.6459004189569417
Iteración 34: theta0 = -0.0007930352901899752, theta1 = 0.007699404823399781, alph
a = 0.001, costo = 0.6458997405418956
Iteración 35: theta0 = -0.0008190795800961687, theta1 = 0.007699720276240743, alph
a = 0.001, costo = 0.6458990621328489
Iteración 36: theta0 = -0.0008451237548788882, theta1 = 0.0077000357247533294, alp
ha = 0.001, costo = 0.6458983837298015
Iteración 37: theta0 = -0.0008711678144839333, theta1 = 0.00770035117345044, alpha
= 0.001, costo = 0.6458977053327535
Iteración 38: theta0 = -0.000897211758939968, theta1 = 0.007700666620003511, alpha
= 0.001, costo = 0.6458970269417046
Iteración 39: theta0 = -0.0009232555882329, theta1 = 0.007700982065614048, alpha =
0.001, costo = 0.6458963485566549
Iteración 40: theta0 = -0.0009492993023706982, theta1 = 0.007701297509662108, alph
a = 0.001, costo = 0.6458956701776047
Iteración 41: theta0 = -0.0009753429013499484, theta1 = 0.007701612952467566, alph
a = 0.001, costo = 0.6458949918045535
Iteración 42: theta0 = -0.00100138638517311, theta1 = 0.0077019283938653945, alpha
= 0.001, costo = 0.6458943134375013
Iteración 43: theta0 = -0.0010274297538396115, theta1 = 0.007702243833940742, alph
a = 0.001, costo = 0.6458936350764481
Iteración 44: theta0 = -0.0010534730073504454, theta1 = 0.007702559272649677, alph
a = 0.001, costo = 0.6458929567213938
Iteración 45: theta0 = -0.0010795161457057975, theta1 = 0.007702874710014889, alph
a = 0.001, costo = 0.6458922783723386
Iteración 46: theta0 = -0.0011055591689062693, theta1 = 0.007703190146024672, alph
a = 0.001, costo = 0.6458916000292821
Iteración 47: theta0 = -0.0011316020769522482, theta1 = 0.007703505580685066, alph
a = 0.001, costo = 0.6458909216922246
Iteración 48: theta0 = -0.001157644869844232, theta1 = 0.007703821013992969, alpha
= 0.001, costo = 0.6458902433611657
Iteración 49: theta0 = -0.0011836875475826614, theta1 = 0.007704136445949989, alph
a = 0.001, costo = 0.6458895650361057
Iteración 50: theta0 = -0.0012097301101680067, theta1 = 0.007704451876555301, alph
a = 0.001, costo = 0.6458888867170441
Iteración 51: theta0 = -0.0012357725576007228, theta1 = 0.0077047673058093374, alp
ha = 0.001, costo = 0.6458882084039814
Iteración 52: theta0 = -0.0012618148898812724, theta1 = 0.007705082733711893, alph
a = 0.001, costo = 0.6458875300969171
Iteración 53: theta0 = -0.0012878571070101149, theta1 = 0.007705398160263064, alph
a = 0.001, costo = 0.6458868517958514
Iteración 54: theta0 = -0.0013138992089877107, theta1 = 0.007705713585462818, alph
a = 0.001, costo = 0.6458861735007841
Iteración 55: theta0 = -0.0013399411958145196, theta1 = 0.007706029009311182, alph
a = 0.001, costo = 0.6458854952117152
Iteración 56: theta0 = -0.0013659830674910022, theta1 = 0.007706344431808147, alph
a = 0.001, costo = 0.6458848169286446
Iteración 57: theta0 = -0.0013920248240176186, theta1 = 0.0077066598529537235, alp
ha = 0.001, costo = 0.6458841386515725
Iteración 58: theta0 = -0.001418066465394829, theta1 = 0.007706975272747911, alpha
= 0.001, costo = 0.6458834603804986
Iteración 59: theta0 = -0.0014441079916230933, theta1 = 0.007707290691190722, alph
a = 0.001, costo = 0.6458827821154228
Iteración 60: theta0 = -0.0014701494027028722, theta1 = 0.0077076061082821615, alp
ha = 0.001, costo = 0.6458821038563453
Iteración 61: theta0 = -0.0014961906986346255, theta1 = 0.00770792152402222, alpha
= 0.001, costo = 0.6458814256032658
Iteración 62: theta0 = -0.0015222318794188136, theta1 = 0.007708236938410924, alph
a = 0.001, costo = 0.6458807473561845
Iteración 63: theta0 = -0.0015482729450558964, theta1 = 0.00770855235144826, alpha
= 0.001, costo = 0.645880069115101
Iteración 64: theta0 = -0.0015743138955463345, theta1 = 0.007708867763134246, alph
a = 0.001, costo = 0.6458793908800156
```

```
Week02 Challenge1
Iteración 65: theta0 = -0.0016003547308905879, theta1 = 0.007709183173468874, alph
a = 0.001, costo = 0.645878712650928
Iteración 66: theta0 = -0.0016263954510891165, theta1 = 0.007709498582452167, alph
a = 0.001, costo = 0.6458780344278385
Iteración 67: theta0 = -0.001652436056142381, theta1 = 0.007709813990084104, alpha
= 0.001, costo = 0.6458773562107466
Iteración 68: theta0 = -0.0016784765460508412, theta1 = 0.00771012939636472, alpha
= 0.001, costo = 0.6458766779996526
Iteración 69: theta0 = -0.0017045169208149576, theta1 = 0.0077104448012939985, alp
ha = 0.001, costo = 0.6458759997945562
Iteración 70: theta0 = -0.00173055718043519, theta1 = 0.0077107602048719515, alpha
= 0.001, costo = 0.6458753215954576
Iteración 71: theta0 = -0.001756597324911999, theta1 = 0.007711075607098588, alpha
= 0.001, costo = 0.6458746434023563
Iteración 72: theta0 = -0.0017826373542458446, theta1 = 0.007711391007973896, alph
a = 0.001, costo = 0.6458739652152528
Iteración 73: theta0 = -0.0018086772684371868, theta1 = 0.007711706407497907, alph
a = 0.001, costo = 0.6458732870341466
Iteración 74: theta0 = -0.0018347170674864863, theta1 = 0.007712021805670598, alph
a = 0.001, costo = 0.6458726088590382
Iteración 75: theta0 = -0.0018607567513942026, theta1 = 0.0077123372024920055, alp
ha = 0.001, costo = 0.645871930689927
Iteración 76: theta0 = -0.0018867963201607966, theta1 = 0.007712652597962095, alph
a = 0.001, costo = 0.6458712525268132
Iteración 77: theta0 = -0.001912835773786728, theta1 = 0.0077129679920809125, alph
a = 0.001, costo = 0.6458705743696966
Iteración 78: theta0 = -0.001938875112272457, theta1 = 0.007713283384848431, alpha
= 0.001, costo = 0.6458698962185773
Iteración 79: theta0 = -0.001964914335618444, theta1 = 0.0077135987762646735, alph
a = 0.001, costo = 0.6458692180734553
Iteración 80: theta0 = -0.001990953443825149, theta1 = 0.0077139141663296355, alph
a = 0.001, costo = 0.6458685399343302
Iteración 81: theta0 = -0.0020169924368930326, theta1 = 0.007714229555043332, alph
a = 0.001, costo = 0.6458678618012025
Iteración 82: theta0 = -0.0020430313148225546, theta1 = 0.007714544942405754, alph
a = 0.001, costo = 0.6458671836740717
Iteración 83: theta0 = -0.002069070077614175, theta1 = 0.007714860328416919, alpha
= 0.001, costo = 0.6458665055529379
Iteración 84: theta0 = -0.002095108725268355, theta1 = 0.007715175713076822, alpha
= 0.001, costo = 0.6458658274378011
Iteración 85: theta0 = -0.0021211472577855536, theta1 = 0.007715491096385478, alph
a = 0.001, costo = 0.6458651493286611
Iteración 86: theta0 = -0.0021471856751662313, theta1 = 0.007715806478342888, alph
a = 0.001, costo = 0.6458644712255182
Iteración 87: theta0 = -0.0021732239774108486, theta1 = 0.007716121858949052, alph
a = 0.001, costo = 0.6458637931283719
Iteración 88: theta0 = -0.0021992621645198656, theta1 = 0.007716437238203979, alph
a = 0.001, costo = 0.6458631150372223
Iteración 89: theta0 = -0.0022253002364937425, theta1 = 0.00771675261610767, alpha
= 0.001, costo = 0.6458624369520695
Iteración 90: theta0 = -0.0022513381933329394, theta1 = 0.0077170679926601435, alp
ha = 0.001, costo = 0.6458617588729133
Iteración 91: theta0 = -0.0022773760350379164, theta1 = 0.007717383367861385, alph
a = 0.001, costo = 0.6458610807997537
Iteración 92: theta0 = -0.0023034137616091336, theta1 = 0.007717698741711418, alph
a = 0.001, costo = 0.6458604027325907
Iteración 93: theta0 = -0.0023294513730470518, theta1 = 0.0077180141142102286, alp
ha = 0.001, costo = 0.6458597246714242
Iteración 94: theta0 = -0.0023554888693521304, theta1 = 0.007718329485357839, alph
a = 0.001, costo = 0.6458590466162539
Iteración 95: theta0 = -0.00238152625052483, theta1 = 0.0077186448551542376, alpha
= 0.001, costo = 0.6458583685670803
Iteración 96: theta0 = -0.002407563516565611, theta1 = 0.007718960223599447, alpha
= 0.001, costo = 0.645857690523903
```

```
Iteración 97: theta0 = -0.0024336006674749336, theta1 = 0.007719275590693456, alph a = 0.001, costo = 0.6458570124867219

Iteración 98: theta0 = -0.0024596377032532575, theta1 = 0.0077195909564362845, alph a = 0.001, costo = 0.645856334455537

Iteración 99: theta0 = -0.002485674623901043, theta1 = 0.007719906320827926, alpha = 0.001, costo = 0.6458556564303485

Iteración 100: theta0 = -0.0025117114294187504, theta1 = 0.0077202216838683875, al pha = 0.001, costo = 0.645854978411156

Parámetros finales para 'Attendance': theta0 = -0.0025117114294187504, theta1 = 0.0077202216838683875
```

- El parámetro theta1 es positivo (0.037), lo que indica que, a medida que aumenta la asistencia (Attendance), la probabilidad de que el estudiante apruebe también aumenta. Sin embargo, el valor de theta1 es bastante pequeño, lo que sugiere que el impacto de la asistencia en la probabilidad de aprobación es bajo.
- El valor de theta0 es negativo (-0.279), lo que implica que, cuando Attendance es 0, la función sigmoide arrojaría una probabilidad menor que 0.5, indicando una tendencia a no aprobar.

#### Caracacterística Homework

```
In [77]: # Característica 'Homework'
X_homework = df['Homework'].values

# Aplicar la regresión Logística
theta0_homework, theta1_homework = logistic_regression(X_homework, y, theta0_init,

print(f"Parámetros finales para 'Homework': theta0 = {theta0_homework}, theta1 = {t

# Predicción
z_homework = theta0_homework + theta1_homework * X_homework
y_pred_homework = sigmoid(z_homework) >= 0.5
```

```
Week02 Challenge1
Iteración 1: theta0 = 0.000125, theta1 = 0.015, alpha = 0.001, costo = 0.693147180
5599453
Iteración 2: theta0 = -9.888626090384368e-06, theta1 = 0.009365959780589291, alpha
= 0.001, costo = 0.6320533494123544
Iteración 3: theta0 = -5.817822262977785e-05, theta1 = 0.010563634054966617, alpha
= 0.001, costo = 0.6188651974291306
Iteración 4: theta0 = -0.00012629384015839902, theta1 = 0.010191339646144555, alph
a = 0.001, costo = 0.618374102959544
Iteración 5: theta0 = -0.00018830626366062763, theta1 = 0.01030198210747851, alpha
= 0.001, costo = 0.6183207801479456
Iteración 6: theta0 = -0.00025212240598776817, theta1 = 0.010269827529988906, alph
a = 0.001, costo = 0.6183125413350568
Iteración 7: theta0 = -0.000315397505991416, theta1 = 0.010280462277617532, alpha
= 0.001, costo = 0.6183081399915199
Iteración 8: theta0 = -0.00037883358586768967, theta1 = 0.010278324177518708, alph
a = 0.001, costo = 0.6183040859837956
Iteración 9: theta0 = -0.00044222103536468733, theta1 = 0.010280003225604722, alph
a = 0.001, costo = 0.6183000628991894
Iteración 10: theta0 = -0.000505622491563601, theta1 = 0.01028054191562476, alpha
= 0.001, costo = 0.6182960426246693
Iteración 11: theta0 = -0.0005690192415714695, theta1 = 0.010281421308680464, alph
a = 0.001, costo = 0.6182920226472572
Iteración 12: theta0 = -0.0006324168762693125, theta1 = 0.010282198906463756, alph
a = 0.001, costo = 0.6182880027427294
Iteración 13: theta0 = -0.0006958137253107289, theta1 = 0.010283006911104894, alph
a = 0.001, costo = 0.618283982891068
Iteración 14: theta0 = -0.0007592102877372595, theta1 = 0.010283805825463035, alph
a = 0.001, costo = 0.6182799630904858
Iteración 15: theta0 = -0.000822606414457681, theta1 = 0.01028460744974439, alpha
= 0.001, costo = 0.6182759433408229
Iteración 16: theta0 = -0.0008860021500174818, theta1 = 0.01028540825852503, alpha
= 0.001, costo = 0.6182719236420641
Iteración 17: theta0 = -0.0009493974811113776, theta1 = 0.010286209305052857, alph
a = 0.001, costo = 0.6182679039942077
Iteración 18: theta0 = -0.0010127924117174427, theta1 = 0.010287010274663623, alph
a = 0.001, costo = 0.618263884397253
Iteración 19: theta0 = -0.001076186940650245, theta1 = 0.010287811261364702, alpha
= 0.001, costo = 0.6182598648511992
Iteración 20: theta0 = -0.0011395810682669726, theta1 = 0.01028861223707115, alpha
= 0.001, costo = 0.6182558453560462
Iteración 21: theta0 = -0.0012029747944639515, theta1 = 0.0102894132101734, alpha
= 0.001, costo = 0.6182518259117926
Iteración 22: theta0 = -0.0012663681192751908, theta1 = 0.010290214178164843, alph
a = 0.001, costo = 0.6182478065184382
Iteración 23: theta0 = -0.0013297610426935668, theta1 = 0.010291015141794344, alph
a = 0.001, costo = 0.6182437871759825
Iteración 24: theta0 = -0.0013931535647242444, theta1 = 0.010291816100838205, alph
a = 0.001, costo = 0.6182397678844246
Iteración 25: theta0 = -0.0014565456853687171, theta1 = 0.010292617055363287, alph
a = 0.001, costo = 0.6182357486437642
Iteración 26: theta0 = -0.0015199374046295758, theta1 = 0.01029341800534963, alpha
= 0.001, costo = 0.6182317294540003
Iteración 27: theta0 = -0.001583328722509083, theta1 = 0.010294218950803225, alpha
= 0.001, costo = 0.6182277103151324
Iteración 28: theta0 = -0.0016467196390095994, theta1 = 0.010295019891722313, alph
a = 0.001, costo = 0.61822369122716
Iteración 29: theta0 = -0.001710110154133457, theta1 = 0.010295820828107436, alpha
= 0.001, costo = 0.6182196721900826
Iteración 30: theta0 = -0.0017735002678829959, theta1 = 0.010296621759958456, alph
a = 0.001, costo = 0.6182156532038994
Iteración 31: theta0 = -0.001836889980260554, theta1 = 0.01029742268727545, alpha
= 0.001, costo = 0.6182116342686095
```

Iteración 32: theta0 = -0.0019002792912684698, theta1 = 0.010298223610058402, alph

a = 0.001, costo = 0.6182076153842129

```
Week02 Challenge1
Iteración 33: theta0 = -0.001963668200909082, theta1 = 0.010299024528307357, alpha
= 0.001, costo = 0.6182035965507087
Iteración 34: theta0 = -0.002027056709184729, theta1 = 0.010299825442022315, alpha
= 0.001, costo = 0.6181995777680959
Iteración 35: theta0 = -0.002090444816097749, theta1 = 0.010300626351203308, alpha
= 0.001, costo = 0.6181955590363745
Iteración 36: theta0 = -0.002153832521650481, theta1 = 0.010301427255850348, alpha
= 0.001, costo = 0.6181915403555435
Iteración 37: theta0 = -0.0022172198258452636, theta1 = 0.010302228155963452, alph
a = 0.001, costo = 0.6181875217256023
Iteración 38: theta0 = -0.0022806067286844354, theta1 = 0.010303029051542647, alph
a = 0.001, costo = 0.6181835031465505
Iteración 39: theta0 = -0.002343993230170335, theta1 = 0.010303829942587942, alpha
= 0.001, costo = 0.6181794846183872
Iteración 40: theta0 = -0.0024073793303053016, theta1 = 0.010304630829099364, alph
a = 0.001, costo = 0.6181754661411122
Iteración 41: theta0 = -0.0024707650290916737, theta1 = 0.010305431711076927, alph
a = 0.001, costo = 0.6181714477147244
Iteración 42: theta0 = -0.00253415032653179, theta1 = 0.010306232588520653, alpha
= 0.001, costo = 0.6181674293392234
Iteración 43: theta0 = -0.0025975352226279896, theta1 = 0.010307033461430562, alph
a = 0.001, costo = 0.6181634110146086
Iteración 44: theta0 = -0.002660919717382611, theta1 = 0.01030783432980666, alpha
= 0.001, costo = 0.6181593927408795
Iteración 45: theta0 = -0.002724303810797994, theta1 = 0.010308635193648987, alpha
= 0.001, costo = 0.618155374518035
Iteración 46: theta0 = -0.0027876875028764766, theta1 = 0.010309436052957546, alph
a = 0.001, costo = 0.6181513563460752
Iteración 47: theta0 = -0.0028510707936203984, theta1 = 0.01031023690773236, alpha
= 0.001, costo = 0.6181473382249989
Iteración 48: theta0 = -0.0029144536830320983, theta1 = 0.01031103775797345, alpha
= 0.001, costo = 0.6181433201548057
Iteración 49: theta0 = -0.0029778361711139158, theta1 = 0.01031183860368083, alpha
= 0.001, costo = 0.6181393021354948
Iteración 50: theta0 = -0.0030412182578681895, theta1 = 0.010312639444854523, alph
a = 0.001, costo = 0.6181352841670661
Iteración 51: theta0 = -0.003104599943297259, theta1 = 0.010313440281494552, alpha
= 0.001, costo = 0.6181312662495184
Iteración 52: theta0 = -0.0031679812274034635, theta1 = 0.010314241113600927, alph
a = 0.001, costo = 0.6181272483828513
Iteración 53: theta0 = -0.0032313621101891423, theta1 = 0.010315041941173666, alph
a = 0.001, costo = 0.6181232305670641
Iteración 54: theta0 = -0.0032947425916566345, theta1 = 0.0103158427642128, alpha
= 0.001, costo = 0.6181192128021564
Iteración 55: theta0 = -0.0033581226718082794, theta1 = 0.01031664358271833, alpha
= 0.001, costo = 0.6181151950881274
Iteración 56: theta0 = -0.0034215023506464167, theta1 = 0.010317444396690294, alph
a = 0.001, costo = 0.6181111774249766
Iteración 57: theta0 = -0.0034848816281733857, theta1 = 0.010318245206128703, alph
a = 0.001, costo = 0.6181071598127033
Iteración 58: theta0 = -0.003548260504391526, theta1 = 0.010319046011033567, alpha
= 0.001, costo = 0.6181031422513068
Iteración 59: theta0 = -0.0036116389793031767, theta1 = 0.010319846811404918, alph
a = 0.001, costo = 0.6180991247407867
Iteración 60: theta0 = -0.0036750170529106776, theta1 = 0.010320647607242767, alph
a = 0.001, costo = 0.6180951072811421
Iteración 61: theta0 = -0.0037383947252163686, theta1 = 0.010321448398547138, alph
a = 0.001, costo = 0.6180910898723726
Iteración 62: theta0 = -0.003801771996222589, theta1 = 0.010322249185318044, alpha
= 0.001, costo = 0.6180870725144776
Iteración 63: theta0 = -0.0038651488659316785, theta1 = 0.010323049967555507, alph
a = 0.001, costo = 0.6180830552074563
Iteración 64: theta0 = -0.003928525334345977, theta1 = 0.010323850745259544, alpha
= 0.001, costo = 0.6180790379513081
```

```
Week02 Challenge1
Iteración 65: theta0 = -0.003991901401467823, theta1 = 0.01032465151843018, alpha
= 0.001, costo = 0.6180750207460327
Iteración 66: theta0 = -0.004055277067299559, theta1 = 0.010325452287067428, alpha
= 0.001, costo = 0.618071003591629
Iteración 67: theta0 = -0.004118652331843522, theta1 = 0.010326253051171306, alpha
= 0.001, costo = 0.6180669864880968
Iteración 68: theta0 = -0.004182027195102053, theta1 = 0.010327053810741834, alpha
= 0.001, costo = 0.6180629694354351
Iteración 69: theta0 = -0.0042454016570774925, theta1 = 0.010327854565779037, alph
a = 0.001, costo = 0.6180589524336437
Iteración 70: theta0 = -0.00430877571777218, theta1 = 0.010328655316282926, alpha
= 0.001, costo = 0.6180549354827216
Iteración 71: theta0 = -0.0043721493771884555, theta1 = 0.010329456062253523, alph
a = 0.001, costo = 0.6180509185826686
Iteración 72: theta0 = -0.0044355226353286585, theta1 = 0.010330256803690844, alph
a = 0.001, costo = 0.6180469017334835
Iteración 73: theta0 = -0.00449889549219513, theta1 = 0.010331057540594914, alpha
= 0.001, costo = 0.6180428849351662
Iteración 74: theta0 = -0.00456226794779021, theta1 = 0.010331858272965747, alpha
= 0.001, costo = 0.6180388681877159
Iteración 75: theta0 = -0.004625640002116238, theta1 = 0.01033265900080336, alpha
= 0.001, costo = 0.618034851491132
Iteración 76: theta0 = -0.004689011655175554, theta1 = 0.010333459724107777, alpha
= 0.001, costo = 0.6180308348454138
Iteración 77: theta0 = -0.004752382906970499, theta1 = 0.010334260442879018, alpha
= 0.001, costo = 0.6180268182505606
Iteración 78: theta0 = -0.004815753757503413, theta1 = 0.010335061157117093, alpha
= 0.001, costo = 0.6180228017065722
Iteración 79: theta0 = -0.004879124206776637, theta1 = 0.010335861866822027, alpha
= 0.001, costo = 0.6180187852134474
Iteración 80: theta0 = -0.00494249425479251, theta1 = 0.010336662571993841, alpha
= 0.001, costo = 0.6180147687711861
Iteración 81: theta0 = -0.005005863901553374, theta1 = 0.010337463272632547, alpha
= 0.001, costo = 0.6180107523797873
Iteración 82: theta0 = -0.005069233147061567, theta1 = 0.010338263968738172, alpha
= 0.001, costo = 0.6180067360392507
Iteración 83: theta0 = -0.005132601991319432, theta1 = 0.01033906466031073, alpha
= 0.001, costo = 0.6180027197495755
Iteración 84: theta0 = -0.005195970434329307, theta1 = 0.01033986534735024, alpha
= 0.001, costo = 0.617998703510761
Iteración 85: theta0 = -0.005259338476093535, theta1 = 0.01034066602985672, alpha
= 0.001, costo = 0.6179946873228067
Iteración 86: theta0 = -0.005322706116614455, theta1 = 0.010341466707830187, alpha
= 0.001, costo = 0.617990671185712
Iteración 87: theta0 = -0.0053860733558944085, theta1 = 0.010342267381270673, alph
a = 0.001, costo = 0.6179866550994764
Iteración 88: theta0 = -0.005449440193935736, theta1 = 0.01034306805017818, alpha
= 0.001, costo = 0.6179826390640989
Iteración 89: theta0 = -0.0055128066307407785, theta1 = 0.010343868714552736, alph
a = 0.001, costo = 0.6179786230795792
Iteración 90: theta0 = -0.005576172666311876, theta1 = 0.010344669374394357, alpha
= 0.001, costo = 0.6179746071459167
Iteración 91: theta0 = -0.00563953830065137, theta1 = 0.010345470029703062, alpha
= 0.001, costo = 0.6179705912631104
Iteración 92: theta0 = -0.0057029035337616005, theta1 = 0.010346270680478865, alph
a = 0.001, costo = 0.6179665754311601
Iteración 93: theta0 = -0.005766268365644909, theta1 = 0.010347071326721802, alpha
= 0.001, costo = 0.6179625596500651
Iteración 94: theta0 = -0.005829632796303636, theta1 = 0.010347871968431874, alpha
= 0.001, costo = 0.6179585439198247
Iteración 95: theta0 = -0.005892996825740123, theta1 = 0.010348672605609104, alpha
= 0.001, costo = 0.6179545282404383
Iteración 96: theta0 = -0.005956360453956711, theta1 = 0.010349473238253516, alpha
```

= 0.001, costo = 0.6179505126119053

```
Iteración 97: theta0 = -0.006019723680955739, theta1 = 0.010350273866365124, alpha = 0.001, costo = 0.6179464970342248

Iteración 98: theta0 = -0.00608308650673955, theta1 = 0.010351074489943948, alpha = 0.001, costo = 0.6179424815073968

Iteración 99: theta0 = -0.006146448931310486, theta1 = 0.01035187510899001, alpha = 0.001, costo = 0.6179384660314201

Iteración 100: theta0 = -0.006209810954670886, theta1 = 0.010352675723503325, alpha = 0.001, costo = 0.6179344506062945

Parámetros finales para 'Homework': theta0 = -0.006209810954670886, theta1 = 0.010352675723503325
```

- El parámetro theta1 es positivo (0.196), lo que indica que a medida que el puntaje de tareas (Homework) aumenta, también lo hace la probabilidad de que el estudiante apruebe. Comparado con Attendance, el valor de theta1 para Homework es más grande, lo que sugiere que esta variable podría tener un mayor impacto en la probabilidad de aprobar.
- El valor de theta0 es negativo (-0.689), lo que nuevamente indica que, con un puntaje de tarea de 0, la probabilidad de aprobar sería baja.

## Métricas de desempeño

- Verdaderos Positivos (TP): Estudiantes que aprueban y el modelo predice correctamente que aprueban.
- Verdaderos Negativos (TN): Estudiantes que no aprueban y el modelo predice correctamente que no aprueban.
- Falsos Positivos (FP): Estudiantes que no aprueban pero el modelo predice incorrectamente que aprueban.
- Falsos Negativos (FN): Estudiantes que aprueban pero el modelo predice incorrectamente que no aprueban.

```
In [78]: # Crear La matriz de confusión y calcular Las métricas
def confusion_matrix_metrics(y_true, y_pred):
    TP = np.sum((y_true == 1) & (y_pred == 1))
    TN = np.sum((y_true == 0) & (y_pred == 0))
    FP = np.sum((y_true == 0) & (y_pred == 1))
    FN = np.sum((y_true == 1) & (y_pred == 0))

accuracy = (TP + TN) / (TP + TN + FP + FN)
    precision = TP / (TP + FP) if (TP + FP) > 0 else 0
    recall = TP / (TP + FN) if (TP + FN) > 0 else 0
    f1 = 2 * (precision * recall) / (precision + recall) if (precision + recall) >
    return accuracy, precision, recall, f1
```

```
In [79]: # Métricas para 'Attendance'
    accuracy_attendance, precision_attendance, recall_attendance, f1_attendance = confu

print(f"Métricas para 'Attendance':")
    print(f"Accuracy: {accuracy_attendance}")
    print(f"Precision: {precision_attendance}")
```

```
print(f"Recall: {recall_attendance}")
         print(f"F1 Score: {f1_attendance}")
         Métricas para 'Attendance':
         Accuracy: 0.625
         Precision: 0.625
         Recall: 1.0
         F1 Score: 0.7692307692307693
In [80]: # Métricas para 'Homework'
         accuracy_homework, precision_homework, recall_homework, f1_homework = confusion_mat
         print(f"Métricas para 'Homework':")
         print(f"Accuracy: {accuracy_homework}")
         print(f"Precision: {precision_homework}")
         print(f"Recall: {recall_homework}")
         print(f"F1 Score: {f1_homework}")
         Métricas para 'Homework':
         Accuracy: 0.625
         Precision: 0.625
         Recall: 1.0
         F1 Score: 0.7692307692307693
```

Los resultados y las predicciones son similares en Attendance y Homework, por lo que las métricas son las mismas.

#### ¿Cuál es mejor? ¿Le ganan a la referencia?

- La métrica de Acurracy es mayor en el modelo de Reference (0.75) en comparación con Attendance y Homework. Esto suguiere que Reference predice correctamente si un estudiante aprueba o no en un porcentaje mayor de los casos.
- Precision es ligeramente más alta en Attendance y Homework (0.625) que en Reference (0.6). Esto indica que cuando Attendance y Homework predicen que un estudiante aprobará, estan en lo correcto un 62.5% de las veces comparado con el 60% de Reference.
- El Recall es de 1.0 para los tres modelos. Esto indica que todos los modelos identifican correctamente todos los casos en los que un estudiante aprueba. Sin embargo, un alto Recall acompañado de un menor Precision podría indicar que el modelo esta realizando muchas predicciones positivas, incluso algunas incorrectas.
- El F1 Score balancea Precision y Recall, este es más alto en Attendence y Homework (0.769) en comparación con Reference (0.75). Esto significa que, aunque Reference tiene

un mejor Accuracy, los modelos basados en Attendance y Homework tienen un mejor equilibrio entre precisión y sensibilidad.

Si se prioriza el Accuracy y el objetivo es maximizar la cantidad de predicciones correctas en general, el modelo basado en Reference es el mejor. En cambio, si se valora un equilibrio entre Precision y Recall, los modelos basados en Attendance o Homework podrían ser mejores opciones, ya que tienen un F1 Score ligeramente más alto.

#### Modelo

Para trabajar os modelos al dividir el conjunto de datos, se optó por manejar los siguientes parámetros:

- theta0 init = 0
- theta1 init = 0
- alpha = 0.001
- iterations = 100

Se decidió obtener el valor de la función de costo de cada iteeración para observar el continuo comportamiento al probar con diferentes parámetros.

La func ión de costo devuelve un valor no negativo, que cuanto menor sea, mejor será el ajuste del modelo a los datos. Si el costo es cercano a 0, significa que el modelo está prediciendo muy bien los resultados. Por el contrario, si el costo es alto, indica que el modelo ésta haciendo muchas predicciones incorrectas y que no se ajusta bien a los datos.

Al probar con una tasa más pequeña pero manteniendo el resto de los parámetros como se menciona, existe una gran cantidad de valoes nan en el valor de la función de costo. Cuando se probo con una tasa mas grande como 0.1, 0.05 y 0.5, el valor del costo era muy lejano a 0.

Cuando se tanteo con las iteraciones mayores a 100, el valor de la función de costo reducia, y volvía a crecer, o en otros casos, se obtenian valores muy altos.

Los parámetros seleccionados, no muestran valores nan y todos son muy cercanos a 0 y van disminuyendo gradualmente, sin emabargo dejan de disminuir significativamente en las primeras iteraciones, lo que puede indicar que el modelo a alcanzado un mínimo local o ha convergido. En este caso, continuar con el entrenamiento quozá no mejore el modelo.

```
In [82]: # Dividir el conjunto de datos en entrenamiento
def train_test_split(df, test_size=0.25):
    np.random.seed(42)
    shuffled_indices = np.random.permutation(len(df))
    test_set_size = int(len(df) * test_size)
    test_indices = shuffled_indices[:test_set_size]
    train_indices = shuffled_indices[test_set_size:]
    return df.iloc[train_indices], df.iloc[test_indices]

# Dividir el conjunto de datos
train_df, test_df = train_test_split(df, test_size=0.25)
In [83]: # Característica 'Attendance' para el conjunto de entrenamiento
X_attendance_train = train_df['Attendance'].values
```

y train values = train df['Pass'].values

```
# Aplicar la regresión logística en el conjunto de entrenamiento
theta0_attendance, theta1_attendance = logistic_regression(X_attendance_train, y_tr
print(f"Parámetros finales para 'Attendance': theta0 = {theta0_attendance}, theta1

# Evaluar el modelo en el conjunto de prueba para 'Attendance'
X_attendance_test = test_df['Attendance'].values
y_test_values = test_df['Pass'].values
z_attendance_test = theta0_attendance + theta1_attendance * X_attendance_test
y_pred_attendance_test = sigmoid(z_attendance_test) >= 0.5
```

```
Week02 Challenge1
= 0.001, costo = 0.6931471805599452
Iteración 2: theta0 = 0.00025208675024002153, theta1 = 0.021164906717516856, alpha
= 0.001, costo = 0.4911277667283197
Iteración 3: theta0 = 0.00022951608454240585, theta1 = 0.018946256483969165, alpha
= 0.001, costo = 0.45605994643487024
Iteración 4: theta0 = 0.00023116114839000581, theta1 = 0.018784031746452528, alpha
= 0.001, costo = 0.4533676013222499
Iteración 5: theta0 = 0.00023470221207197705, theta1 = 0.018783078766200112, alpha
= 0.001, costo = 0.4533543407298557
Iteración 6: theta0 = 0.00023825396961689075, theta1 = 0.018783036201710568, alpha
= 0.001, costo = 0.4533543276974191
Iteración 7: theta0 = 0.00024180572835025722, theta1 = 0.018782994449018636, alpha
= 0.001, costo = 0.4533543150806408
Iteración 8: theta0 = 0.00024535747873859735, theta1 = 0.018782952697128368, alpha
= 0.001, costo = 0.4533543024639223
Iteración 9: theta0 = 0.000248909220773686, theta1 = 0.018782910945338217, alpha =
0.001, costo = 0.453354289847263
Iteración 10: theta0 = 0.0002524609544555379, theta1 = 0.018782869193647588, alpha
= 0.001, costo = 0.4533542772306632
Iteración 11: theta0 = 0.0002560126797841746, theta1 = 0.01878282744205647, alpha
= 0.001, costo = 0.45335426461412265
Iteración 12: theta0 = 0.0002595643967596179, theta1 = 0.01878278569056487, alpha
= 0.001, costo = 0.45335425199764157
Iteración 13: theta0 = 0.00026311610538188945, theta1 = 0.018782743939172782, alph
a = 0.001, costo = 0.45335423938121966
Iteración 14: theta0 = 0.000266667805651011, theta1 = 0.018782702187880217, alpha
= 0.001, costo = 0.4533542267648572
Iteración 15: theta0 = 0.0002702194975670042, theta1 = 0.01878266043668716, alpha
= 0.001, costo = 0.45335421414855415
Iteración 16: theta0 = 0.0002737711811298908, theta1 = 0.018782618685593618, alpha
= 0.001, costo = 0.45335420153231026
Iteración 17: theta0 = 0.0002773228563396925, theta1 = 0.01878257693459959, alpha
= 0.001, costo = 0.45335418891612594
Iteración 18: theta0 = 0.000280874523196431, theta1 = 0.018782535183705075, alpha
= 0.001, costo = 0.45335417630000074
Iteración 19: theta0 = 0.00028442618170012815, theta1 = 0.018782493432910084, alph
a = 0.001, costo = 0.453354163683935
Iteración 20: theta0 = 0.0002879778318508054, theta1 = 0.0187824516822146, alpha =
0.001, costo = 0.4533541510679285
Iteración 21: theta0 = 0.00029152947364848454, theta1 = 0.018782409931618622, alph
a = 0.001, costo = 0.45335413845198147
Iteración 22: theta0 = 0.0002950811070931874, theta1 = 0.018782368181122163, alpha
= 0.001, costo = 0.4533541258360939
Iteración 23: theta0 = 0.0002986327321849356, theta1 = 0.01878232643072522, alpha
= 0.001, costo = 0.4533541132202653
Iteración 24: theta0 = 0.0003021843489237507, theta1 = 0.018782284680427783, alpha
= 0.001, costo = 0.4533541006044964
Iteración 25: theta0 = 0.0003057359573096547, theta1 = 0.018782242930229866, alpha
= 0.001, costo = 0.4533540879887866
Iteración 26: theta0 = 0.00030928755734266903, theta1 = 0.01878220118013145, alpha
= 0.001, costo = 0.4533540753731361
Iteración 27: theta0 = 0.0003128391490228156, theta1 = 0.01878215943013255, alpha
= 0.001, costo = 0.45335406275754525
Iteración 28: theta0 = 0.0003163907323501161, theta1 = 0.01878211768023317, alpha
= 0.001, costo = 0.45335405014201346
Iteración 29: theta0 = 0.00031994230732459213, theta1 = 0.018782075930433295, alph
a = 0.001, costo = 0.4533540375265411
Iteración 30: theta0 = 0.0003234938739462654, theta1 = 0.018782034180732928, alpha
= 0.001, costo = 0.4533540249111282
Iteración 31: theta0 = 0.0003270454322151577, theta1 = 0.01878199243113208, alpha
= 0.001, costo = 0.4533540122957745
```

Iteración 32: theta0 = 0.00033059698213129063, theta1 = 0.018781950681630733, alph

a = 0.001, costo = 0.4533539996804801

```
Week02 Challenge1
Iteración 33: theta0 = 0.0003341485236946859, theta1 = 0.018781908932228895, alpha
= 0.001, costo = 0.45335398706524527
Iteración 34: theta0 = 0.00033770005690536537, theta1 = 0.018781867182926574, alph
a = 0.001, costo = 0.45335397445006953
Iteración 35: theta0 = 0.0003412515817633506, theta1 = 0.018781825433723757, alpha
= 0.001, costo = 0.453353961834953
Iteración 36: theta0 = 0.00034480309826866336, theta1 = 0.01878178368462045, alpha
= 0.001, costo = 0.4533539492198962
Iteración 37: theta0 = 0.00034835460642132537, theta1 = 0.018781741935616655, alph
a = 0.001, costo = 0.45335393660489853
Iteración 38: theta0 = 0.00035190610622135834, theta1 = 0.01878170018671237, alpha
= 0.001, costo = 0.4533539239899603
Iteración 39: theta0 = 0.00035545759766878387, theta1 = 0.018781658437907592, alph
a = 0.001, costo = 0.4533539113750815
Iteración 40: theta0 = 0.00035900908076362374, theta1 = 0.018781616689202318, alph
a = 0.001, costo = 0.45335389876026166
Iteración 41: theta0 = 0.00036256055550589963, theta1 = 0.018781574940596554, alph
a = 0.001, costo = 0.45335388614550143
Iteración 42: theta0 = 0.00036611202189563324, theta1 = 0.018781533192090298, alph
a = 0.001, costo = 0.4533538735308005
Iteración 43: theta0 = 0.0003696634799328463, theta1 = 0.018781491443683546, alpha
= 0.001, costo = 0.4533538609161587
Iteración 44: theta0 = 0.0003732149296175606, theta1 = 0.018781449695376304, alpha
= 0.001, costo = 0.4533538483015765
Iteración 45: theta0 = 0.00037676637094979774, theta1 = 0.018781407947168576, alph
a = 0.001, costo = 0.4533538356870537
Iteración 46: theta0 = 0.00038031780392957945, theta1 = 0.01878136619906035, alpha
= 0.001, costo = 0.45335382307259
Iteración 47: theta0 = 0.0003838692285569274, theta1 = 0.01878132445105163, alpha
= 0.001, costo = 0.4533538104581856
Iteración 48: theta0 = 0.0003874206448318633, theta1 = 0.018781282703142412, alpha
= 0.001, costo = 0.4533537978438407
Iteración 49: theta0 = 0.0003909720527544089, theta1 = 0.018781240955332706, alpha
= 0.001, costo = 0.4533537852295551
Iteración 50: theta0 = 0.00039452345232458586, theta1 = 0.0187811992076225, alpha
= 0.001, costo = 0.45335377261532894
Iteración 51: theta0 = 0.0003980748435424159, theta1 = 0.0187811574600118, alpha =
0.001, costo = 0.4533537600011619
Iteración 52: theta0 = 0.0004016262264079207, theta1 = 0.018781115712500608, alpha
= 0.001, costo = 0.45335374738705436
Iteración 53: theta0 = 0.00040517760092112205, theta1 = 0.018781073965088917, alph
a = 0.001, costo = 0.453353734773006
Iteración 54: theta0 = 0.00040872896708204164, theta1 = 0.018781032217776737, alph
a = 0.001, costo = 0.453353722159017
Iteración 55: theta0 = 0.00041228032489070104, theta1 = 0.018780990470564054, alph
a = 0.001, costo = 0.45335370954508747
Iteración 56: theta0 = 0.0004158316743471221, theta1 = 0.01878094872345088, alpha
= 0.001, costo = 0.45335369693121713
Iteración 57: theta0 = 0.0004193830154513264, theta1 = 0.018780906976437205, alpha
= 0.001, costo = 0.4533536843174062
Iteración 58: theta0 = 0.0004229343482033359, theta1 = 0.018780865229523044, alpha
= 0.001, costo = 0.45335367170365465
Iteración 59: theta0 = 0.00042648567260317183, theta1 = 0.01878082348270837, alpha
= 0.001, costo = 0.45335365908996217
Iteración 60: theta0 = 0.00043003698865085635, theta1 = 0.018780781735993207, alph
a = 0.001, costo = 0.4533536464763295
Iteración 61: theta0 = 0.000433588296346411, theta1 = 0.018780739989377546, alpha
= 0.001, costo = 0.4533536338627556
Iteración 62: theta0 = 0.0004371395956898576, theta1 = 0.018780698242861396, alpha
= 0.001, costo = 0.45335362124924133
Iteración 63: theta0 = 0.00044069088668121753, theta1 = 0.018780656496444736, alph
```

Iteración 64: theta0 = 0.0004442421693205127, theta1 = 0.01878061475012758, alpha

a = 0.001, costo = 0.4533536086357864

= 0.001, costo = 0.45335359602239095

```
Iteración 65: theta0 = 0.0004477934436077649, theta1 = 0.018780573003909926, alpha
= 0.001, costo = 0.4533535834090546
Iteración 66: theta0 = 0.00045134470954299576, theta1 = 0.01878053125779178, alpha
= 0.001, costo = 0.45335357079577765
Iteración 67: theta0 = 0.00045489596712622695, theta1 = 0.01878048951177313, alpha
= 0.001, costo = 0.45335355818255996
Iteración 68: theta0 = 0.0004584472163574802, theta1 = 0.018780447765853983, alpha
= 0.001, costo = 0.45335354556940166
Iteración 69: theta0 = 0.00046199845723677706, theta1 = 0.01878040602003433, alpha
= 0.001, costo = 0.45335353295630243
Iteración 70: theta0 = 0.0004655496897641395, theta1 = 0.018780364274314187, alpha
= 0.001, costo = 0.45335352034326293
Iteración 71: theta0 = 0.0004691009139395891, theta1 = 0.018780322528693542, alpha
= 0.001, costo = 0.4533535077302825
Iteración 72: theta0 = 0.00047265212976314755, theta1 = 0.01878028078317239, alpha
= 0.001, costo = 0.4533534951173615
Iteración 73: theta0 = 0.00047620333723483654, theta1 = 0.018780239037750737, alph
a = 0.001, costo = 0.4533534825044998
Iteración 74: theta0 = 0.00047975453635467784, theta1 = 0.018780197292428587, alph
a = 0.001, costo = 0.4533534698916975
Iteración 75: theta0 = 0.0004833057271226931, theta1 = 0.01878015554720594, alpha
= 0.001, costo = 0.45335345727895443
Iteración 76: theta0 = 0.000486856909538904, theta1 = 0.018780113802082787, alpha
= 0.001, costo = 0.45335344466627087
Iteración 77: theta0 = 0.0004904080836033323, theta1 = 0.018780072057059134, alpha
= 0.001, costo = 0.45335343205364653
Iteración 78: theta0 = 0.0004939592493159997, theta1 = 0.01878003031213498, alpha
= 0.001, costo = 0.45335341944108143
Iteración 79: theta0 = 0.0004975104066769278, theta1 = 0.01877998856731032, alpha
= 0.001, costo = 0.4533534068285756
Iteración 80: theta0 = 0.0005010615556861384, theta1 = 0.01877994682258516, alpha
= 0.001, costo = 0.4533533942161292
Iteración 81: theta0 = 0.0005046126963436532, theta1 = 0.018779905077959493, alpha
= 0.001, costo = 0.45335338160374217
Iteración 82: theta0 = 0.0005081638286494939, theta1 = 0.018779863333433327, alpha
= 0.001, costo = 0.4533533689914145
Iteración 83: theta0 = 0.0005117149526036823, theta1 = 0.01877982158900666, alpha
= 0.001, costo = 0.453353356379146
Iteración 84: theta0 = 0.0005152660682062399, theta1 = 0.01877977984467949, alpha
= 0.001, costo = 0.4533533437669368
Iteración 85: theta0 = 0.0005188171754571885, theta1 = 0.018779738100451807, alpha
= 0.001, costo = 0.4533533311547871
Iteración 86: theta0 = 0.0005223682743565498, theta1 = 0.018779696356323636, alpha
= 0.001, costo = 0.45335331854269667
Iteración 87: theta0 = 0.0005259193649043454, theta1 = 0.018779654612294947, alpha
= 0.001, costo = 0.4533533059306655
Iteración 88: theta0 = 0.0005294704471005972, theta1 = 0.018779612868365762, alpha
= 0.001, costo = 0.4533532933186939
Iteración 89: theta0 = 0.0005330215209453268, theta1 = 0.018779571124536064, alpha
= 0.001, costo = 0.4533532807067811
Iteración 90: theta0 = 0.0005365725864385558, theta1 = 0.018779529380805866, alpha
= 0.001, costo = 0.45335326809492815
Iteración 91: theta0 = 0.0005401236435803061, theta1 = 0.01877948763717516, alpha
= 0.001, costo = 0.45335325548313415
Iteración 92: theta0 = 0.0005436746923705994, theta1 = 0.018779445893643953, alpha
= 0.001, costo = 0.4533532428713996
Iteración 93: theta0 = 0.0005472257328094572, theta1 = 0.018779404150212238, alpha
= 0.001, costo = 0.4533532302597244
Iteración 94: theta0 = 0.0005507767648969015, theta1 = 0.01877936240688002, alpha
= 0.001, costo = 0.45335321764810854
Iteración 95: theta0 = 0.0005543277886329535, theta1 = 0.01877932066364729, alpha
= 0.001, costo = 0.4533532050365519
Iteración 96: theta0 = 0.0005578788040176354, theta1 = 0.018779278920514054, alpha
= 0.001, costo = 0.4533531924250548
```

Iteración 97: theta0 = 0.0005614298110509687, theta1 = 0.018779237177480312, alpha = 0.001, costo = 0.4533531798136169Iteración 98: theta0 = 0.0005649808097329751, theta1 = 0.018779195434546064, alpha = 0.001, costo = 0.4533531672022382Iteración 99: theta0 = 0.0005685318000636763, theta1 = 0.018779153691711306, alpha = 0.001, costo = 0.4533531545909189Iteración 100: theta0 = 0.000572082782043094, theta1 = 0.01877911194897604, alpha = 0.001, costo = 0.4533531419796591Parámetros finales para 'Attendance': theta0 = 0.000572082782043094, theta1 = 0.01877911194897604

```
In [84]: # Característica 'Homework' para el conjunto de entrenamiento
X_homework_train = train_df['Homework'].values

# Aplicar la regresión logística en el conjunto de entrenamiento
theta0_homework, theta1_homework = logistic_regression(X_homework_train, y_train_va
print(f"Parámetros finales para 'Homework': theta0 = {theta0_homework}, theta1 = {t

# Evaluar el modelo en el conjunto de prueba para 'Homework'
X_homework_test = test_df['Homework'].values
z_homework_test = theta0_homework + theta1_homework * X_homework_test
y_pred_homework_test = sigmoid(z_homework_test) >= 0.5
```

```
Week02 Challenge1
= 0.001, costo = 0.6931471805599452
Iteración 2: theta0 = 0.0002452163682669277, theta1 = 0.02579766724899144, alpha =
0.001, costo = 0.411725718035884
Iteración 3: theta0 = 0.000187955971738443, theta1 = 0.02368526602628352, alpha =
0.001, costo = 0.3958708796005844
Iteración 4: theta0 = 0.00014818963579660917, theta1 = 0.02300211406629224, alpha
= 0.001, costo = 0.39288117797952016
Iteración 5: theta0 = 0.00011459171663526482, theta1 = 0.022824109717528754, alpha
= 0.001, costo = 0.3925843355386902
Iteración 6: theta0 = 8.264627128332796e-05, theta1 = 0.022781525830082944, alpha
= 0.001, costo = 0.3925635793349409
Iteración 7: theta0 = 5.110156420009472e-05, theta1 = 0.02277178454373626, alpha =
0.001, costo = 0.3925614508067139
Iteración 8: theta0 = 1.965148713780879e-05, theta1 = 0.022769794870605937, alpha
= 0.001, costo = 0.3925604000892998
Iteración 9: theta0 = -1.1776351544951803e-05, theta1 = 0.022769622705635352, alph
a = 0.001, costo = 0.39255940918092413
Iteración 10: theta0 = -4.319893328848943e-05, theta1 = 0.022769876024418025, alph
a = 0.001, costo = 0.3925584215614722
Iteración 11: theta0 = -7.462023393206303e-05, theta1 = 0.022770228913497163, alph
a = 0.001, costo = 0.3925574341261746
Iteración 12: theta0 = -0.00010604118391663829, theta1 = 0.022770605101354962, alp
ha = 0.001, costo = 0.3925564467049079
Iteración 13: theta0 = -0.000137462000965157, theta1 = 0.022770986740488396, alpha
= 0.001, costo = 0.39255545928835583
Iteración 14: theta0 = -0.0001688827360236577, theta1 = 0.022771369654683776, alph
a = 0.001, costo = 0.39255447187600817
Iteración 15: theta0 = -0.00020030340101327931, theta1 = 0.022771752866734545, alp
ha = 0.001, costo = 0.392553484467837
Iteración 16: theta0 = -0.0002317239987235893, theta1 = 0.022772136147980605, alph
a = 0.001, costo = 0.3925524970638408
Iteración 17: theta0 = -0.000263144529807426, theta1 = 0.02277251944491686, alpha
= 0.001, costo = 0.3925515096640196
Iteración 18: theta0 = -0.00029456499441764597, theta1 = 0.02277290274502341, alph
a = 0.001, costo = 0.3925505222683732
Iteración 19: theta0 = -0.0003259853925901128, theta1 = 0.02277328604537067, alpha
= 0.001, costo = 0.3925495348769016
Iteración 20: theta0 = -0.00035740572433331445, theta1 = 0.022773669345273116, alp
ha = 0.001, costo = 0.39254854748960477
Iteración 21: theta0 = -0.00038882598964933295, theta1 = 0.022774052644570348, alp
ha = 0.001, costo = 0.39254756010648284
Iteración 22: theta0 = -0.0004202461885387514, theta1 = 0.022774435943224834, alph
a = 0.001, costo = 0.39254657272753574
Iteración 23: theta0 = -0.0004516663210018022, theta1 = 0.022774819241227786, alph
a = 0.001, costo = 0.3925455853527633
Iteración 24: theta0 = -0.0004830863870386356, theta1 = 0.02277520253857716, alpha
= 0.001, costo = 0.3925445979821656
Iteración 25: theta0 = -0.0005145063866493828, theta1 = 0.022775585835272463, alph
a = 0.001, costo = 0.3925436106157427
Iteración 26: theta0 = -0.0005459263198341704, theta1 = 0.022775969131313592, alph
a = 0.001, costo = 0.39254262325349454
Iteración 27: theta0 = -0.0005773461865931239, theta1 = 0.02277635242670052, alpha
= 0.001, costo = 0.3925416358954209
Iteración 28: theta0 = -0.0006087659869263688, theta1 = 0.022776735721433232, alph
a = 0.001, costo = 0.3925406485415222
Iteración 29: theta0 = -0.0006401857208340302, theta1 = 0.02277711901551174, alpha
= 0.001, costo = 0.392539661191798
Iteración 30: theta0 = -0.0006716053883162335, theta1 = 0.022777502308936038, alph
a = 0.001, costo = 0.3925386738462484
Iteración 31: theta0 = -0.0007030249893731037, theta1 = 0.022777885601706127, alph
a = 0.001, costo = 0.3925376865048736
Iteración 32: theta0 = -0.0007344445240047663, theta1 = 0.02277826889382201, alpha
= 0.001, costo = 0.39253669916767336
```

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Iteración 33: theta0 = -0.0007658639922113466, theta1 = 0.02277865218528368, alpha
= 0.001, costo = 0.39253571183464764
Iteración 34: theta0 = -0.0007972833939929696, theta1 = 0.02277903547609115, alpha
= 0.001, costo = 0.3925347245057965
Iteración 35: theta0 = -0.0008287027293497608, theta1 = 0.022779418766244407, alph
a = 0.001, costo = 0.39253373718112006
Iteración 36: theta0 = -0.0008601219982818452, theta1 = 0.02277980205574346, alpha
= 0.001, costo = 0.39253274986061804
Iteración 37: theta0 = -0.0008915412007893483, theta1 = 0.022780185344588307, alph
a = 0.001, costo = 0.3925317625442907
Iteración 38: theta0 = -0.0009229603368723951, theta1 = 0.022780568632778955, alph
a = 0.001, costo = 0.3925307752321377
Iteración 39: theta0 = -0.0009543794065311112, theta1 = 0.022780951920315387, alph
a = 0.001, costo = 0.39252978792415927
Iteración 40: theta0 = -0.0009857984097656216, theta1 = 0.022781335207197627, alph
a = 0.001, costo = 0.39252880062035533
Iteración 41: theta0 = -0.0010172173465760517, theta1 = 0.022781718493425654, alph
a = 0.001, costo = 0.3925278133207257
Iteración 42: theta0 = -0.0010486362169625264, theta1 = 0.022782101778999492, alph
a = 0.001, costo = 0.39252682602527084
Iteración 43: theta0 = -0.0010800550209251714, theta1 = 0.022782485063919118, alph
a = 0.001, costo = 0.39252583873399033
Iteración 44: theta0 = -0.001111473758464112, theta1 = 0.02278286834818455, alpha
= 0.001, costo = 0.39252485144688404
Iteración 45: theta0 = -0.0011428924295794733, theta1 = 0.022783251631795776, alph
a = 0.001, costo = 0.3925238641639524
Iteración 46: theta0 = -0.0011743110342713804, theta1 = 0.022783634914752805, alph
a = 0.001, costo = 0.392522876885195
Iteración 47: theta0 = -0.0012057295725399588, theta1 = 0.022784018197055632, alph
a = 0.001, costo = 0.39252188961061185
Iteración 48: theta0 = -0.0012371480443853337, theta1 = 0.02278440147870426, alpha
= 0.001, costo = 0.3925209023402032
Iteración 49: theta0 = -0.0012685664498076304, theta1 = 0.02278478475969869, alpha
= 0.001, costo = 0.3925199150739689
Iteración 50: theta0 = -0.001299984788806974, theta1 = 0.02278516804003892, alpha
= 0.001, costo = 0.39251892781190884
Iteración 51: theta0 = -0.00133140306138349, theta1 = 0.02278555131972496, alpha =
0.001, costo = 0.392517940554023
Iteración 52: theta0 = -0.0013628212675373035, theta1 = 0.0227859345987568, alpha
= 0.001, costo = 0.39251695330031167
Iteración 53: theta0 = -0.0013942394072685399, theta1 = 0.022786317877134445, alph
a = 0.001, costo = 0.39251596605077455
Iteración 54: theta0 = -0.0014256574805773243, theta1 = 0.022786701154857886, alph
a = 0.001, costo = 0.3925149788054115
Iteración 55: theta0 = -0.0014570754874637822, theta1 = 0.02278708443192714, alpha
= 0.001, costo = 0.39251399156422295
Iteración 56: theta0 = -0.0014884934279280387, theta1 = 0.022787467708342203, alph
a = 0.001, costo = 0.39251300432720837
Iteración 57: theta0 = -0.0015199113019702193, theta1 = 0.022787850984103065, alph
a = 0.001, costo = 0.3925120170943681
Iteración 58: theta0 = -0.001551329109590449, theta1 = 0.022788234259209738, alpha
= 0.001, costo = 0.3925110298657021
Iteración 59: theta0 = -0.0015827468507888533, theta1 = 0.02278861753366222, alpha
= 0.001, costo = 0.39251004264121014
Iteración 60: theta0 = -0.0016141645255655572, theta1 = 0.022789000807460512, alph
a = 0.001, costo = 0.3925090554208924
Iteración 61: theta0 = -0.0016455821339206862, theta1 = 0.02278938408060461, alpha
= 0.001, costo = 0.39250806820474865
Iteración 62: theta0 = -0.0016769996758543655, theta1 = 0.022789767353094514, alph
a = 0.001, costo = 0.3925070809927792
Iteración 63: theta0 = -0.0017084171513667206, theta1 = 0.022790150624930228, alph
a = 0.001, costo = 0.3925060937849837
Iteración 64: theta0 = -0.0017398345604578765, theta1 = 0.02279053389611176, alpha
= 0.001, costo = 0.39250510658136234
```

```
Iteración 65: theta0 = -0.0017712519031279586, theta1 = 0.022790917166639092, alph
a = 0.001, costo = 0.39250411938191504
Iteración 66: theta0 = -0.0018026691793770922, theta1 = 0.022791300436512243, alph
a = 0.001, costo = 0.3925031321866418
Iteración 67: theta0 = -0.0018340863892054025, theta1 = 0.022791683705731206, alph
a = 0.001, costo = 0.39250214499554265
Iteración 68: theta0 = -0.0018655035326130148, theta1 = 0.02279206697429598, alpha
= 0.001, costo = 0.3925011578086174
Iteración 69: theta0 = -0.0018969206096000545, theta1 = 0.022792450242206573, alph
a = 0.001, costo = 0.3925001706258663
Iteración 70: theta0 = -0.0019283376201666468, theta1 = 0.022792833509462974, alph
a = 0.001, costo = 0.3924991834472892
Iteración 71: theta0 = -0.001959754564312917, theta1 = 0.02279321677606519, alpha
= 0.001, costo = 0.39249819627288585
Iteración 72: theta0 = -0.0019911714420389904, theta1 = 0.022793600042013223, alph
a = 0.001, costo = 0.3924972091026565
Iteración 73: theta0 = -0.002022588253344992, theta1 = 0.022793983307307072, alpha
= 0.001, costo = 0.39249622193660116
Iteración 74: theta0 = -0.002054004998231048, theta1 = 0.022794366571946733, alpha
= 0.001, costo = 0.39249523477471976
Iteración 75: theta0 = -0.002085421676697283, theta1 = 0.022794749835932212, alpha
= 0.001, costo = 0.3924942476170121
Iteración 76: theta0 = -0.002116838288743822, theta1 = 0.02279513309926351, alpha
= 0.001, costo = 0.3924932604634786
Iteración 77: theta0 = -0.002148254834370791, theta1 = 0.02279551636194063, alpha
= 0.001, costo = 0.3924922733141188
Iteración 78: theta0 = -0.0021796713135783146, theta1 = 0.022795899623963566, alph
a = 0.001, costo = 0.3924912861689328
Iteración 79: theta0 = -0.0022110877263665186, theta1 = 0.022796282885332315, alph
a = 0.001, costo = 0.39249029902792065
Iteración 80: theta0 = -0.0022425040727355283, theta1 = 0.022796666146046897, alph
a = 0.001, costo = 0.3924893118910822
Iteración 81: theta0 = -0.002273920352685469, theta1 = 0.022797049406107293, alpha
= 0.001, costo = 0.3924883247584179
Iteración 82: theta0 = -0.0023053365662164655, theta1 = 0.022797432665513505, alph
a = 0.001, costo = 0.39248733762992716
Iteración 83: theta0 = -0.0023367527133286436, theta1 = 0.022797815924265545, alph
a = 0.001, costo = 0.3924863505056102
Iteración 84: theta0 = -0.0023681687940221284, theta1 = 0.022798199182363404, alph
a = 0.001, costo = 0.39248536338546686
Iteración 85: theta0 = -0.0023995848082970454, theta1 = 0.022798582439807088, alph
a = 0.001, costo = 0.39248437626949745
Iteración 86: theta0 = -0.0024310007561535196, theta1 = 0.022798965696596594, alph
a = 0.001, costo = 0.39248338915770165
Iteración 87: theta0 = -0.0024624166375916763, theta1 = 0.022799348952731925, alph
a = 0.001, costo = 0.3924824020500797
Iteración 88: theta0 = -0.0024938324526116414, theta1 = 0.02279973220821308, alpha
= 0.001, costo = 0.39248141494663125
Iteración 89: theta0 = -0.0025252482012135397, theta1 = 0.022800115463040057, alph
a = 0.001, costo = 0.3924804278473566
Iteración 90: theta0 = -0.0025566638833974965, theta1 = 0.02280049871721287, alpha
= 0.001, costo = 0.3924794407522556
Iteración 91: theta0 = -0.002588079499163637, theta1 = 0.0228008819707315, alpha =
0.001, costo = 0.39247845366132783
Iteración 92: theta0 = -0.002619495048512087, theta1 = 0.02280126522359596, alpha
= 0.001, costo = 0.3924774665745743
Iteración 93: theta0 = -0.0026509105314429715, theta1 = 0.022801648475806246, alph
a = 0.001, costo = 0.3924764794919938
Iteración 94: theta0 = -0.0026823259479564158, theta1 = 0.022802031727362358, alph
a = 0.001, costo = 0.39247549241358737
Iteración 95: theta0 = -0.0027137412980525453, theta1 = 0.022802414978264303, alph
a = 0.001, costo = 0.39247450533935435
Iteración 96: theta0 = -0.0027451565817314854, theta1 = 0.022802798228512076, alph
a = 0.001, costo = 0.3924735182692947
```

```
Iteración 97: theta0 = -0.002776571798993361, theta1 = 0.02280318147810568, alpha = 0.001, costo = 0.39247253120340864

Iteración 98: theta0 = -0.0028079869498382976, theta1 = 0.022803564727045107, alpha = 0.001, costo = 0.39247154414169616

Iteración 99: theta0 = -0.002839402034266421, theta1 = 0.022803947975330374, alpha = 0.001, costo = 0.39247055708415723

Iteración 100: theta0 = -0.0028708170522778556, theta1 = 0.022804331222961466, alpha = 0.001, costo = 0.3924695700307917

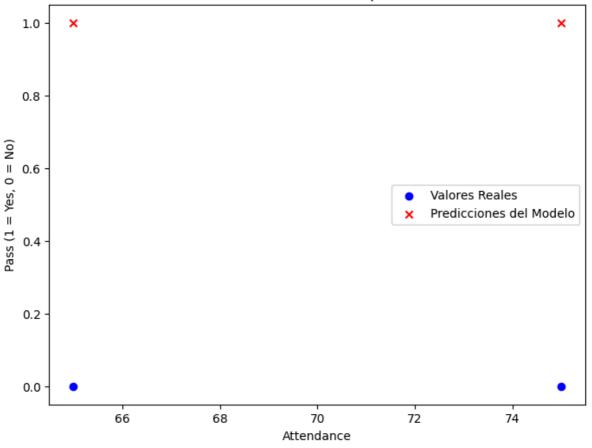
Parámetros finales para 'Homework': theta0 = -0.0028708170522778556, theta1 = 0.022804331222961466
```

## Prueba de Implementación

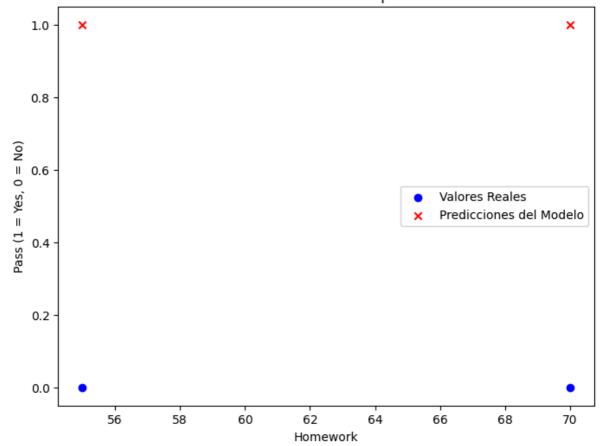
Utiliza el modelo entrenado para hacer predecir las salidas del subconjunto de prueba, y compara contra los datos reales en una gráfica.

```
# Predecir las salidas del subconjunto de prueba para 'Attendance'
In [85]:
         y_pred_attendance_test = sigmoid(theta0_attendance + theta1_attendance * X_attendar
         # Predecir las salidas del subconjunto de prueba para 'Homework'
         y_pred_homework_test = sigmoid(theta0_homework + theta1_homework * X_homework_test)
         # Graficar los resultados para 'Attendance'
In [86]:
         plt.figure(figsize=(8, 6))
         plt.scatter(X_attendance_test, y_test_values, color='blue', label='Valores Reales')
         plt.scatter(X_attendance_test, y_pred_attendance_test, color='red', marker='x', lak
         plt.xlabel('Attendance')
         plt.ylabel('Pass (1 = Yes, 0 = No)')
         plt.title('Predicciones vs Valores Reales para Attendance')
         plt.legend()
         plt.show()
         # Graficar los resultados para 'Homework'
         plt.figure(figsize=(8, 6))
         plt.scatter(X homework test, y test values, color='blue', label='Valores Reales')
         plt.scatter(X_homework_test, y_pred_homework_test, color='red', marker='x', label='
         plt.xlabel('Homework')
         plt.ylabel('Pass (1 = Yes, 0 = No)')
         plt.title('Predicciones vs Valores Reales para Homework')
         plt.legend()
         plt.show()
```





### Predicciones vs Valores Reales para Homework



En el caso de Attendance, parece que las observaciones con asistencia baja (cerca de 65) no pasan (clase 0), mientras que las observaciones con asistencia mayor (cerca de 75) pasan

(clase 1). En este caso, el modelo predice correctamente las observaciones, ya que las cruces rojas se superponen a los puntos azules correspondientes.

Para la variable Homework, las observaciones más bajas (cerca de 55) no pasan (clase 0) y las observaciones con puntajes más altos (cerca de 70) pasan (clase 1). Similar a Attendance, el modelo predice correctamente las observaciones en este caso, con las cruces rojas superpuestas a los puntos azules.

# Valor de la función de costo para el subconjunto de entrenamiento, y para el subconjunto de prueba.

```
In [87]: # Función de costo
def compute_cost(X, y, theta0, theta1):
    n = len(y)
    z = theta0 + theta1 * X
    h = sigmoid(z)
    cost = -1/n * np.sum(y * np.log(h) + (1 - y) * np.log(1 - h))
    return cost
```

In [88]: # Calcular la función de costo para el conjunto de entrenamiento y prueba para 'Att
 train\_cost\_attendance = compute\_cost(X\_attendance\_train, y\_train\_values, theta0\_att
 test\_cost\_attendance = compute\_cost(X\_attendance\_test, y\_test\_values, theta0\_attendance)
 print(f"Función de costo para 'Attendance' - Entrenamiento: {train\_cost\_attendance})

Función de costo para 'Attendance' - Entrenamiento: 0.45335312936845834, Prueba: 1.5536369948909567

In [89]: # Calcular la función de costo para el conjunto de entrenamiento y prueba para 'Hon
train\_cost\_homework = compute\_cost(X\_homework\_train, y\_train\_values, theta0\_homework
test\_cost\_homework = compute\_cost(X\_homework\_test, y\_test\_values, theta0\_homework,
print(f"Función de costo para 'Homework' - Entrenamiento: {train\_cost\_homework}, Pr

Función de costo para 'Homework' - Entrenamiento: 0.3924685829815997, Prueba: 1.64 07157462830257

Tanto en Attendance y Homework, la función del costo en el conjunto de prueba es significativamente mayor que en el conjunto de entrenamiento. Esto sugiere que el modelo puede estar sobreajustado (overfitting) a los datos de entrenamiento. En otras palabras, el modelo ha aprendido a predecir muy bien los ejemplos en el conjunto de entrenamiento, pero no generaliza bien a ejemplos nuevos que no ha visto antes, como los del conjunto de prueba.

## Métricas de desempeño

```
In [90]: # Métricas para 'Attendance'
attendance_accuracy, attendance_precision, attendance_recall, attendance_f1 = confu

print(f"Métricas para 'Attendance':")
print(f"Accuracy: {attendance_accuracy}")
print(f"Precision: {attendance_precision}")
print(f"Recall: {attendance_recall}")
print(f"F1 Score: {attendance_f1}")
```

```
Métricas para 'Attendance':
         Accuracy: 0.0
         Precision: 0.0
         Recall: 0
         F1 Score: 0
In [91]: # Métricas para 'Homework'
         homework_accuracy, homework_precision, homework_recall, homework_f1 = confusion_mat
         print(f"Métricas para 'Homework':")
         print(f"Accuracy: {homework_accuracy}")
         print(f"Precision: {homework_precision}")
         print(f"Recall: {homework_recall}")
         print(f"F1 Score: {homework_f1}")
         Métricas para 'Homework':
         Accuracy: 0.0
         Precision: 0.0
         Recall: 0
         F1 Score: 0
In [92]: # Métricas para Reference
         accuracy_reference, precision_reference, recall_reference, f1_reference = confusior
         print(f"Métricas para la referencia:")
         print(f"Accuracy: {accuracy_reference}")
         print(f"Precision: {precision_reference}")
         print(f"Recall: {recall_reference}")
         print(f"F1 Score: {f1_reference}")
         Métricas para la referencia:
         Accuracy: 0.75
         Precision: 0.6
         Recall: 1.0
         F1 Score: 0.749999999999999
```

#### ¿Cuál es mejor? ¿Le ganan a la referencia?

Los clasificadores basados en Attendance y Homework son claramente inferiores al clasificador de referencia. Ninguna de las métricas para Attendance o Homework es positiva, lo que indica que estos modelos no pueden hacer predicciones útiles. El clasificador basado en la referencia es significativamente mejor que los otros dos modelos en términos de todas las métricas importantes.

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