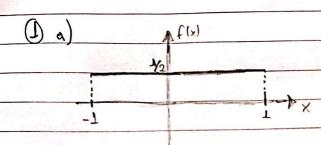
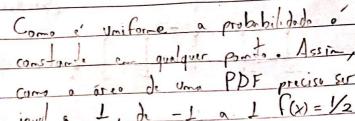
STQQSS

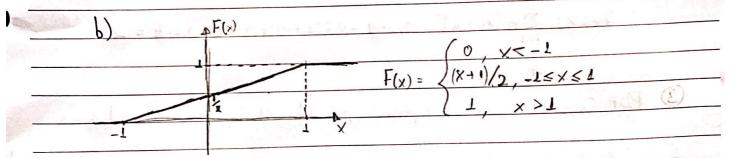
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PPGEE 2249 - Aprendizado de Magrina.

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c)
$$P(-0,2 < X < 0,2)$$

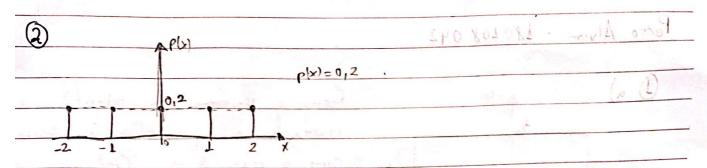
= $F(0,2) - F(-0,2)$
= $0,2+1 - (-0,2+1)$
2 2

$$\frac{1}{2} = \frac{1}{2} = \frac{1}$$

$$E[x^{2}] = \int_{-\infty}^{\infty} x^{2} f(x) dx = x^{3} \int_{-1}^{1} = \frac{1}{6} - (-1) = \frac{1}{6}$$

$$E[x^4] = \int_{-\infty}^{\infty} x^4 f(x) \, dx = \frac{x^5}{10} \frac{1}{10} = \frac{1}{10} - \frac{(-1)}{10} = \frac{1}{5}$$

$$V_{or} = E[(x - E(x))^2] = \int_{-\infty}^{\infty} (x - 0)^2 f(x) dx = E[x^2] = 1$$



$$E[x] = \sum_{i=1}^{6} x_i \rho(x_i) = -2 \cdot 0, 2 - 1 \cdot 0, 2 + 0 + 1 \cdot 0, 2 + 2 \cdot 0, 2 = 0$$

$$V_{\text{or}}[x] = \sum_{i=1}^{\infty} P_i(x_i - \mu)^2 = 0, 2 \cdot 4 + 0, 2 \cdot L + 0 + 0, 2 \cdot L + 0, 2 \cdot 4 = 2$$

$$f_{X_{1}X_{2}}(X_{1}, X_{2}) = \frac{1}{2\pi\sigma_{X_{1}} \nabla_{X_{2}} \sqrt{1-\rho^{2}}} \left(\frac{x_{1} - \mu_{X_{1}}}{2(1-\rho^{2})} \left(\frac{x_{1} - \mu_{X_{1}}}{\sigma_{X_{1}}} \right)^{2} - 2\rho \left(\frac{x_{1} - \mu_{X_{1}}}{\sigma_{X_{1}}} \right) \left(\frac{x_{2} - \mu_{X_{2}}}{\sigma_{X_{2}}} \right) + \left(\frac{x_{2} - \mu_{X_{2}}}{\sigma_{X_{2}}} \right)^{2} \right)$$

$$P = P_{\text{eorSom}} \quad \text{correlation} \quad \text{coeficient} = \quad \frac{Cov(X_1, X_2)}{\sigma_{\overline{X}_1} \sigma_{\overline{X}_2}} \quad \text{in} \quad \text{i$$

$$\frac{\sqrt{0} | \frac{1}{1} | \frac{1}{$$

$$\frac{(2(X_{1},X_{2})=\frac{1}{4\pi\sqrt{1.84}}-\exp\left[-\frac{1}{3.68}\left[(X_{1}+2)^{2}+0.4(X_{1}+2)(X_{2}-1)+(X_{2}-1)^{2}\right]}{3.68}$$

☐ AlvimPedro / ppge_machine_learning Public <> Code Issues Pull requests Actions Projects Wiki Security ✓ Insights Settings Beta Try the new code view ⊮ main 🕶 ppge_machine_learning / lista1 / ex4.ipynb Go to file AlvimPedro Lista 1 Latest commit 77a5ac8 5 minutes ago (1) History A 1 contributor Q 282 lines (282 sloc) 7.21 KB Raw Blame <> In [38]: import pandas as pd import math df = pd.read_csv('housing.csv') df.head() Out[38]: longitude latitude housing_median_age total_rooms total_bedrooms population households median_income median_house_value -122.23 37.88 41.0 880.0 129.0 322.0 126.0 8.3252 452600.0 -122.22 1 37.86 21.0 7099.0 1106.0 2401.0 1138.0 8.3014 358500.0 -122.24 37.85 52.0 1467.0 190.0 496.0 177.0 7.2574 352100.0 341300.0 -122.25 37.85 52.0 1274.0 235.0 558.0 219.0 5.6431 -122.25 37.85 52.0 1627.0 280.0 565.0 259.0 3.8462 342200.0 Choosing median_income and median_house_value Mean In [39]: m1 = df['median_income'].mean() m2 = df['median_house_value'].mean() Variance In [40]: v1 = df['median_income'].var() v2 = df['median_house_value'].var() Standard Deviation In [41]: s1 = math.sqrt(v1)s2 = math.sqrt(v2)Covariance In [42]: cov = ((df['median_income'] - m1)*(df['median_house_value'] - m2)).sum() / len(df) Correlation In [44]: corr = cov / (s1*s2)In [52]: print(f'A correlação entre media de renda do bairro e a media de preço das casas do bairro é de: {corr:.4f} A correlação entre media de renda do bairro e a media de preço das casas do bairro é de: 0.6880 In []:

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