Математическая и прикладная статистика

Лабораторная работа

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Вариант

Количество гласных букв = 10 Количество согласных букв = 16 Вариант = 06

Исходный код

https://github.com/DenisBelobrotski/AppliedStatisticsLab (https://github.com/DenisBelobrotski/AppliedStatisticsLab)

Задание 1

Модули

In [1]:

```
import math
import numpy as np
from scipy import stats
from scipy import special
import matplotlib.pyplot as plt
# %matplotlib inline
```

Чтение данных (data_1_var_06.txt)

In [2]:

```
data file name = "data 1 var 06.txt"
def parse_data(file_name):
    result data = []
    result data len = 0
    all_data_parsed = False
    with open(file_name, "r") as file:
        numbers = file.read().split(",")
        numbers len = len(numbers)
        for current in numbers:
            result data.append(float(current.strip()))
        result_data_len = len(result_data)
        all_data_parsed = (numbers_len == result_data_len)
    return result data, result data len, all data parsed
data, dataLen, allDataParsed = parse_data(data_file_name)
print("Parsed numbers count = " + str(dataLen))
print("Is all data parsed = " + str(allDataParsed))
sortedData = sorted(data)
```

Parsed numbers count = 345 Is all data parsed = True

а) Выборочное среднее

```
X_1,\dots,X_n - выборка ar{X}=rac{1}{n}\sum_i^n X_i
```

In [3]:

```
def calc_sample_mean(in_data):
    in_data_len = len(in_data)
    mean_sum = 0.0
    for number in in_data:
        mean_sum += number
    return mean_sum / in_data_len

sampleMean = calc_sample_mean(data)
print("Sample mean = " + str(sampleMean))
```

Sample mean = 3.5859199999999984

б) Выборочная дисперсия

In [4]:

```
def calc_variance_sum(in_data, mean):
    variance_sum = 0.0
    for number in in_data:
        variance_sum += (number - mean)**2
    return variance_sum

def calc_sample_variance(in_data, mean):
    in_data_len = len(in_data)
    return calc_variance_sum(in_data, mean) / in_data_len

sampleVariance = calc_sample_variance(data, sampleMean)
print("Sample variance = " + str(sampleVariance))
```

Sample variance = 1.2073530875130443

в) Исправленная дисперсия

In [5]:

```
def calc_unbiased_sample_variance(in_data, mean):
    in_data_len = len(in_data)
    return calc_variance_sum(in_data, mean) / (in_data_len - 1)

unbiasedSampleVariance = calc_unbiased_sample_variance(data, sampleMean)
print("Unbiased sample variance = " + str(unbiasedSampleVariance))
```

Unbiased sample variance = 1.210862834860466

г) Размах выборки

In [6]:

```
dataMin = sortedData[0]
dataMax = sortedData[-1]
dataRange = dataMax - dataMin
print("Min = " + str(dataMin))
print("Max = " + str(dataMax))
print("Range = " + str(dataRange))
Min = 0.6819
```

```
Min = 0.6819
Max = 6.5951
Range = 5.913200000000001
```

д) Медиана

In [7]:

```
median = sortedData[dataLen // 2]
if dataLen % 2 == 0:
    median += sortedData[dataLen // 2 - 1]
    median /= 2

print("Median = " + str(median))
```

Median = 3.5788

е) Квартили

In [8]:

```
lowerQuartile = sortedData[dataLen // 4]
upperQuartile = sortedData[3 * dataLen // 4]

print("Q1 = " + str(lowerQuartile))
print("Q2 = " + str(median))
print("Q3 = " + str(upperQuartile))

Q1 = 2.8036
Q2 = 3.5788
Q3 = 4.3214
```

ж) Выборочная квантиль уровня 1/3

In [9]:

```
quantile_1_3 = sortedData[dataLen // 3]
print("Q = " + str(quantile_1_3))
```

Q = 3.0461

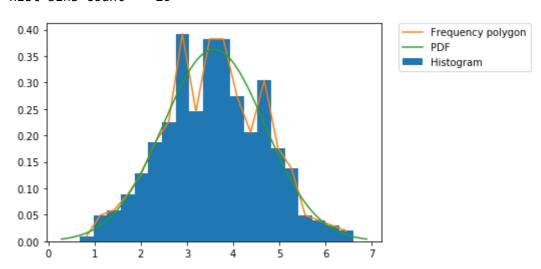
з) Гистограмма, полигон частот, плотность нормального распределения.

PDF - Probability Density Function (Функция плотности распределения)

In [10]:

```
def norm pdf(x, mu, sigma):
    return (1.0 / (sigma * math.sqrt(2 * math.pi))) * math.exp(-0.5 * ((x - mu) / sigma
)**2)
def norm_pdf_list(x, mu, sigma):
    result = x.copy()
    for i in range(len(x)):
        result[i] = norm_pdf(result[i], mu, sigma)
    return result
def draw_pdf(hist_bins_count, mu, sigma):
    n, bins, patches = plt.hist(data, hist_bins_count, density=True, label="Histogram")
    bins = np.delete(bins, -1)
    for i in range(hist bins count):
        bins[i] += patches[i].get_width() / 2
    plt.plot(bins, n, label="Frequency polygon")
    nodes_count = 100
    nodes = np.linspace(mu - 3 * sigma, mu + 3 * sigma, nodes_count)
    plt.plot(nodes, norm pdf list(nodes, mu, sigma), label="PDF")
    plt.legend(bbox_to_anchor=(1.05, 1), loc='upper left', borderaxespad=0.0)
    print("mu = ", str(mu))
    print("sigma = ", str(sigma))
    print("hist bins count = ", str(hist_bins_count))
draw_pdf(20, sampleMean, math.sqrt(unbiasedSampleVariance))
```

mu = 3.585919999999984 sigma = 1.1003921277710351 hist bins count = 20



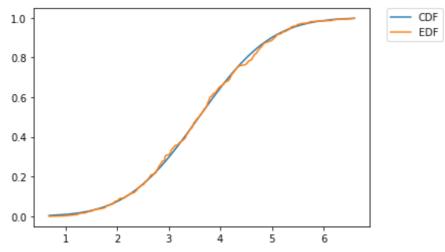
и) Эмпирическая функция распределения и функция распределения нормального закона

CDF - Cumulative distribution function (Теоритическая функция распределения)

EDF - Empirical distribution function (Эмпирическая функция распределения)

In [11]:

```
def norm cdf(x, mu, sigma):
    return 0.5 * (1 + math.erf((x - mu) / (sigma * 1.4142)))
def norm_cdf_list(in_data, mu, sigma):
    out_data = in_data.copy()
    for i in range(len(out_data)):
        out_data[i] = norm_cdf(out_data[i], mu, sigma)
    return out_data
def edf(in data):
    out_data = in_data.copy()
    out_data_len = len(out_data)
    for i in range(out_data_len):
        out_data[i] = i / out_data_len
    return out_data
def draw_cdf(mu, sigma):
    plt.plot(sortedData, norm_cdf_list(sortedData, mu, sigma), label="CDF")
    plt.plot(sortedData, edf(sortedData), label="EDF")
    plt.legend(bbox_to_anchor=(1.05, 1), loc='upper left', borderaxespad=0.0)
draw cdf(sampleMean, math.sqrt(unbiasedSampleVariance))
```



к) Доверительные интервалы для среднего и дисперсии с вероятностью 0.99

In [12]:

```
def confidence interval mean(mean, variance, data len, probability):
    bias = math.sqrt(variance / (data_len - 1)) * stats.t.ppf(1 - probability / 2, data
_len - 1)
    return mean - bias, mean + bias
def confidence interval_variance(variance, data_len, probability):
    lower bound = (data len - 1) * variance / stats.chi2.ppf((1 + probability) / 2, dat
a len - 1)
    upper_bound = (data_len - 1) * variance / stats.chi2.ppf((1 - probability) / 2, dat
a len - 1)
    return lower bound, upper bound
print("Mean confidence interval = " + str(confidence_interval_mean(sampleMean, unbiased
SampleVariance, dataLen, 0.99)))
print("Sample mean = " + str(sampleMean))
print()
print("Variance confidence interval = " + str(confidence_interval_variance(unbiasedSamp
leVariance, dataLen, 0.99)))
print("Unbiased sample variance = " + str(unbiasedSampleVariance))
```

```
Mean confidence interval = (3.5851758590018594, 3.5866641409981375)

Sample mean = 3.585919999999984

Variance confidence interval = (1.0029449329842604, 1.4865973537588315)

Unbiased sample variance = 1.210862834860466
```

Задание 2

In [13]:

```
significance = 0.05
test_hists_bins_count = 20
kolmogorov_distribution_quantile = 1.36
a1_distribution_quantile = 0.46
a2_distribution_quantile = 2.49
chi2_distribution_quantile = stats.chi2.ppf(1 - significance, test_hists_bins_count - 1
)
norm_distribution_quantile = stats.norm.ppf(1 - significance)
```

а) При помощи критерия Колмогорова проверить гипотезу о том, что данные имеют нормальный закон распределения со средним 10 и дисперсией 5.5. Уровень занчимости 0.05.

In [14]:

passed = False

passed = False

```
def kolmogorov test normal(sorted data, mean, variance, quantile):
    data_len = len(sorted data)
    cdf result = norm cdf list(sortedData, mean, math.sqrt(variance))
    edf result = edf(sortedData)
    max distance = 0
    for i in range(data len):
        # max_distance = max(max_distance, math.fabs(cdf_result[i] - edf_result[i]))
        max_distance = max(max_distance, math.fabs(cdf_result[i] - (i + 1) / data_len))
    result_statistic = math.sqrt(data_len) * max_distance
    return result statistic, (result statistic < quantile)</pre>
kolmogorov test normal statistic, kolmogorov test normal passed = \
    kolmogorov_test_normal(sortedData, 10.0, 5.5, kolmogorov_distribution_quantile)
print("Kolmogorov test:")
print("statistic = ", str(kolmogorov_test_normal_statistic))
print("quantile = ", str(kolmogorov_distribution_quantile))
print("passed = ", str(kolmogorov_test_normal_passed))
Kolmogorov test:
statistic = 17.604856029131053
quantile = 1.36
```

б) При помощи критерия Крамера-Мизеса проверить гипотезу о том, что данные имеют нормальный закон распределения со средним 10 и дисперсией 5.5. Уровень занчимости 0.05.

```
In [15]:
def cramer_mises_test_normal(sorted_data, mean, variance, quntile):
    data len = len(sorted data)
    cdf result = norm cdf list(sortedData, mean, math.sqrt(variance))
    result_statistic = 1 / (12 * data_len)
    for i in range(data_len):
        result statistic += (cdf result[i] - (2 * i + 1)/(2 * data len))**2
    return result statistic, (result statistic < quntile)</pre>
cramer_mises_test_normal_statistic, cramer_mises_test_normal_passed = \
    cramer_mises_test_normal(sortedData, 10.0, 5.5, a1_distribution_quantile)
print("Cramer-Mises test:")
print("statistic = ", str(cramer mises test normal statistic))
print("quantile = ", str(a1_distribution_quantile))
print("passed = ", str(cramer mises test normal passed))
Cramer-Mises test:
statistic = 111.35222279248605
quantile = 0.46
```

в) При помощи критерия Андерсона-Дарлинга проверить гипотезу о том, что данные имеют нормальный закон распределения со средним 10 и дисперсией 5.5. Уровень занчимости 0.05.

In [16]:

```
def anderson darling test normal(sorted data, mean, variance, guntile):
    data len = len(sorted data)
    cdf result = norm cdf list(sortedData, mean, math.sqrt(variance))
    result statistic = -data len
    for i in range(data_len):
        coef = (2 * i + 1)/(2 * data_len)
        result statistic += -2 * (coef * math.log(cdf result[i]) + (1 - coef) * math.lo
g(1 - cdf_result[i]))
    return result_statistic, (result_statistic < quntile)</pre>
anderson darling test normal statistic, anderson darling test normal passed = \
    anderson darling test normal(sortedData, 10.0, 5.5, a2 distribution quantile)
print("Anderson-Darling test:")
print("statistic = ", str(anderson_darling_test_normal_statistic))
print("quantile = ", str(a2_distribution_quantile))
print("passed = ", str(anderson_darling_test_normal_passed))
Anderson-Darling test:
statistic = 1402.6543578852445
quantile = 2.49
passed = False
```

Чтение данных второй выборки (data 1 var 07.txt)

In [17]:

```
data2_file_name = "data_1_var_07.txt"

data2, data2Len, allData2Parsed = parse_data(data2_file_name)

print("Parsed numbers count = " + str(data2Len))
print("Is all data parsed = " + str(allData2Parsed))
sortedData2 = sorted(data2)

Parsed numbers count = 349
```

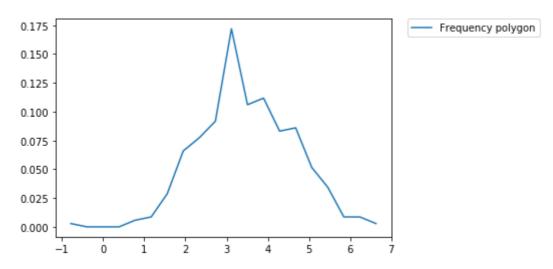
Is all data parsed = True

г) При помощи критерия хи-квадрат проверить гипотезу, что вторая выборка имеет нормальное распределение. Уровень значимости 0.05.

In [18]:

```
def create hist(in data, intervals count, lower bound, upper bound, normalized = False
):
    out hist = [0] * intervals count
    intervals_positions = [0] * intervals_count
hist_step = (upper_bound - lower_bound) / intervals_count
    cur border = lower bound + hist step
    cur_interval_index = 0
    in_data_len = len(in_data)
    for i in range(intervals count):
        intervals positions[i] = lower bound + hist step * i
    for current_number in in_data:
        if current number < cur border:</pre>
            out_hist[cur_interval_index] += 1
        elif cur interval index < intervals count - 1:
            cur interval index += 1
            cur border += hist step
    if normalized:
        for i in range(intervals_count):
            out_hist[i] /= in_data_len
    return out_hist, intervals_positions
def pearson_test_normal(in_data, bins_count, quantile):
    mean = calc_sample_mean(in_data)
    sqrd variance = math.sqrt(calc unbiased sample variance(in data, mean))
    in data len = len(in data)
    in_data_hist, _ = create_hist(in_data, bins_count, in_data[0], in_data[-1], False)
    step = (in_data[-1] - in_data[0]) / bins_count
    cur_border = in_data[0]
    chi_sum = 0
    for i in range(bins_count):
        prev border = cur border
        cur_border = in_data[0] + (i + 1) * step
        expected_count = in_data_len * (norm_cdf(cur_border, mean, sqrd_variance) - nor
m_cdf(prev_border, mean, sqrd_variance))
        chi sum += (in data hist[i] - expected count)**2 / expected count
    return chi sum, (chi sum < quantile)</pre>
hist, interval_positions = create_hist(sortedData2, test_hists_bins_count, sortedData2[
0], sortedData2[-1], True)
plt.plot(interval positions, hist, label="Frequency polygon")
_ = plt.legend(bbox_to_anchor=(1.05, 1), loc='upper left', borderaxespad=0.0)
pearson_test_normal_statistic, pearson_test_normal_passed = \
    pearson_test_normal(sortedData2, test_hists_bins_count, chi2_distribution_quantile)
print("Pearson test:")
print("hist intervals = ", str(test hists bins count))
print("statistic = ", str(pearson_test_normal_statistic))
print("quantile = ", str(chi2_distribution_quantile))
print("passed = ", str(pearson_test_normal_passed))
```

```
Pearson test:
hist intervals = 20
statistic = 30.06662489471957
quantile = 30.14352720564616
passed = True
```



Сравнение эмпирических функций распределения двух выборок

In [19]:

```
edf_1 = edf(sortedData)
edf_2 = edf(sortedData2)
plt.plot(sortedData, edf_1, label="data_1_var_06.txt")
plt.plot(sortedData2, edf_2, label="data_1_var_07.txt")
= plt.legend(bbox_to_anchor=(1.05, 1), loc='upper left', borderaxespad=0.0)
                                                          data 1 var 06.txt
1.0
                                                         data 1 var 07.txt
0.8
0.6
0.4
0.2
```

д) Проверить гипотезу однородности выборок при помощи критерия Колмогорова-Смирнова. Уровень значимости 0.05.

In [20]:

```
def apply edf(in data, in data edf, argument):
    in_data_len = len(in_data)
    result = 0
    if argument > in data[-1]:
        result = 1
    elif argument > in_data[0]:
        for i in range(in_data_len):
            if argument <= in_data[i]:</pre>
                percentage = (argument - in_data[i - 1]) / (in_data[i] - in_data[i - 1
])
                result = in data edf[i - 1] + percentage * (in data edf[i] - in data ed
f[i - 1])
                break
    else:
        result = 0
    return result
def calc_smirnov_test_distance(in_data_1, in_data_2, in_data_2_edf):
    in_data_1_len = len(in_data_1)
    distance = 0
    for i in range(in_data_1_len):
        distance = max(distance, math.fabs(
            (i + 1) / in_data_1_len -
            apply_edf(in_data_2, in_data_2_edf, in_data_1[i])))
    return distance
def smirnov_test(in_data_1, in_data_2, quantile):
    in_data_1_len = len(in_data_1)
    in_data_2_len = len(in_data_2)
    in edf 1 = edf(in data 1)
    in_edf_2 = edf(in_data_2)
    distance_positive = calc_smirnov_test_distance(in_data_1, in_data_2, in_edf_2)
    distance_negative = calc_smirnov_test_distance(in_data_2, in_data_1, in_edf_1)
    distance = max(distance_positive, distance_negative)
    statistic coef = math.sqrt(in data 1 len * in data 2 len / (in data 1 len + in data
_2_len))
    statistic = statistic coef * distance
    return statistic, (statistic < quantile)</pre>
smirnov test statistic, smirnov test passed = \
    smirnov test(sortedData, sortedData2, kolmogorov distribution quantile)
print("Smirnov test:")
print("statistic = ", str(smirnov_test_statistic))
print("quantile = ", str(kolmogorov_distribution_quantile))
print("passed = ", str(smirnov_test_passed))
Smirnov test:
statistic = 0.7671199254742549
quantile = 1.36
```

passed = True

е) Проверить гипотезу однородности выборок при помощи критерия Розенблатта. Уровень значимости 0.05.

In [21]:

```
def calc ranks(in data):
    out_data = in_data.copy()
    out_data_len = len(out_data)
    for i in range(out data len):
        out data[i] = i
    return out_data
def rosenblatt_test(in_data_1, in_data_2, quantile):
    in_data_1_len = len(in_data_1)
    in data 2 len = len(in data 2)
    ranks_1 = calc_ranks(in_data_1)
    ranks 2 = calc ranks(in data 2)
    sum 1 = 0
    for i in range(in_data_1_len):
        sum_1 += (ranks_1[i] - (i + 1))**2
    sum_1 /= in_data_2_len
    sum 2 = 0
    for j in range(in_data_2_len):
        sum_2 += (ranks_2[j] - (j + 1))**2
    sum_2 /= in_data_1_len
    squared_omega = 1. / (in_data_1_len * in_data_2_len) * ((1. / 6) + sum_1 + sum_2) -
    statistic = (in_data_1_len * in_data_2_len) / (in_data_1_len + in_data_2_len) * squ
ared_omega
    mean = 1. / 6 * (1 + 1 / (in_data_1_len + in_data_2_len))
    variance = 1. / 45 * (1 + 1 / (in_data_1_len + in_data_2_len)) * (1 + 1 / (in_data_
1_len + in_data_2_len) - 3. / 4 * (1 / in_data_1_len + 1 / in_data_2_len))
    corrected_statistic = (statistic - mean) / math.sqrt(45. * variance) + 1. / 6
    return corrected_statistic, (corrected_statistic < quantile)</pre>
# https://en.wikipedia.org/wiki/Cram%C3%A9r%E2%80%93von Mises criterion
def rosenblatt_test_2(in_data_1, in_data_2, quantile):
    in_data_1_len = len(in_data_1)
    in data 2 len = len(in data 2)
    ranks 1 = calc ranks(in data 1)
    ranks_2 = calc_ranks(in_data_2)
    sum_1 = 0
    for i in range(in_data_1_len):
        sum 1 += (ranks 1[i] - (i + 1))**2
    sum 2 = 0
    for j in range(in_data_2_len):
        sum_2 += (ranks_2[j] - (j + 1))**2
    n = in data 1 len
   m = in_data_2_len
    u = n * sum_1 + m * sum_2
    t = u / (n * m * (n + m)) - (4 * m * n - 1) / (6 * (m + n))
    return t, (t < quantile)</pre>
```

```
rosenblatt_test_statistic, rosenblatt_test_passed = \
    rosenblatt_test(sortedData, sortedData2, a2_distribution_quantile)
print("Rosenblatt test:")
print("statistic = ", str(rosenblatt_test_statistic))
print("quantile = ", str(a2_distribution_quantile))
print("passed = ", str(rosenblatt_test_passed))
```

```
Rosenblatt test:

statistic = -115.74373038178214

quantile = 2.49

passed = True
```

ж) Проверить гипотезу однородности выборок при помощи критерия Уилкоксона (Манна-Уитни). Уровень значимости 0.05.

In [22]:

```
def wilcoxon_mann_whitney_test(in_data_1, in_data_2, quantile):
    n = len(in_data_1)
    m = len(in_data_2)
    result sum = 0
    for i in range(n):
        for j in range(m):
            if in_data_1[i] < in_data_2[j]:</pre>
                result_sum += 1
    mean = (n * m) / 2
    variance = n * m * (n + m + 1) / 12
    statistic = (result_sum - mean) / math.sqrt(variance)
    return statistic, (statistic < quantile)</pre>
wilcoxon_mann_whitney_test_statistic, wilcoxon_mann_whitney_test_passed = \
    wilcoxon mann whitney test(sortedData, sortedData2, norm distribution quantile)
print("Wilcoxon-Mann-Whitney test:")
print("statistic = ", str(wilcoxon_mann_whitney_test_statistic))
print("quantile = ", str(norm_distribution_quantile))
print("passed = ", str(wilcoxon mann whitney test passed))
```

```
Wilcoxon-Mann-Whitney test:
statistic = 0.8026188969873727
quantile = 1.6448536269514722
passed = True
```

In [23]:

```
def pearson test 2 experiments(in data 1, in data 2, bins count, quantile):
    in_data_1_len = len(in_data_1)
    in data 2 len = len(in data 2)
    in data len = in data 1 len + in data 2 len
    lower_bound = min(in_data_1[0], in_data_2[0])
    upper_bound = max(in_data_1[-1], in_data_2[-1])
    hist_1, _ = create_hist(in_data_1, bins_count, lower_bound, upper_bound, False)
    hist_2, _ = create_hist(in_data_2, bins_count, lower_bound, upper_bound, False)
    sum 1 = 0
    for j in range(bins_count):
        frequency = (hist_1[j] + hist_2[j]) / in_data_len
        if frequency < 1.0e-05:</pre>
            continue
        sum_1 += (hist_1[j] - in_data_1_len * frequency)**2 / (in_data_1_len * frequenc
y)
    sum_2 = 0
    for j in range(bins_count):
        frequency = (hist_1[j] + hist_2[j]) / in_data_len
        if frequency < 1.0e-05:</pre>
            continue
        sum_2 += (hist_2[j] - in_data_2_len * frequency)**2 / (in_data_2_len * frequenc
y)
    statistic = sum 1 + sum 2
    return statistic, (statistic < quantile)</pre>
pearson_test_2_experiments_statistic, pearson_test_2_experiments_passed = \
    pearson_test_2_experiments(sortedData, sortedData2, test_hists_bins_count, chi2_dis
tribution quantile)
print("Pearson test:")
print("statistic = ", str(pearson_test_2_experiments_statistic))
print("quantile = ", str(chi2_distribution_quantile))
print("passed = ", str(pearson_test_2_experiments passed))
```

Pearson test:

```
statistic = 18.04004731707197
quantile = 30.14352720564616
passed = True
```