

■ Wilcoxon 検定の棄却確率の推定

```

In[1]:= m[n_, k_] := 0 /; k < 0
        m[n_, k_] := 0 /; k > n (n + 1) / 2
        m[1, 0] = m[1, 1] = 1;
        m[n_, k_] := m[n,  $\frac{n(n+1)}{2} - k$ ] /; k >  $\frac{n(n+1)}{4}$ 
        m[n_, k_] := m[n, k] = m[n - 1, k] + m[n - 1, k - n]

In[6]:= cW[n_, a_] := Module[{i, p, d},
        i = n (n + 1) / 2;
        d = 2^n;
        p = m[n, 1] / d;
        While[p < a, i = i - 1; p = p + m[n, i] / d];
        {i, p}]

In[7]:= cW[16, 0.10]

Out[7]= {93,  $\frac{433}{4096}$ }

In[8]:= N[ $\frac{433}{4096}$ ]

Out[8]= 0.105713

In[9]:= N[Sum[m[16, i] / (2^16), {i, 93, 8*17}]]

Out[9]= 0.105713

In[10]:= outW = 93

Out[10]= 93

In[11]:= <<Statistics`ContinuousDistributions`

In[12]:= dataC[n_, a_, b_] := RandomArray[CauchyDistribution[a, b], n]

In[13]:= x = dataC[5, 0, 1]

Out[13]= {-3.41631, 0.875673, 0.12594, -3.24563, 0.929477}

In[14]:= y = Abs[x]

Out[14]= {3.41631, 0.875673, 0.12594, 3.24563, 0.929477}

In[15]:= z = Ordering[y]

Out[15]= {3, 2, 5, 4, 1}

In[16]:= Sort[y]

Out[16]= {0.12594, 0.875673, 0.929477, 3.24563, 3.41631}

In[17]:= s[x_] := 1 /; x > 0
        s[x_] := 0 /; x ≤ 0

In[19]:= Table[s[x[[i]]], {i, 5}]

Out[19]= {0, 1, 1, 0, 1}

```

```
In[20]:= wilcoxon[x_List] := Module[{z},
  z = Ordering[Abs[x]];
  Sum[1 * s[x[[z[[1]]]]], {1, Length[x]}]
```

```
In[21]:= x
```

```
Out[21]= {-3.41631, 0.875673, 0.12594, -3.24563, 0.929477}
```

```
In[22]:= Sort[y]
```

```
Out[22]= {0.12594, 0.875673, 0.929477, 3.24563, 3.41631}
```

```
In[23]:= wilcoxon[x]
```

```
Out[23]= 6
```

```
In[24]:= wilcoxon[dataC[16, 0, 1]]
```

```
Out[24]= 87
```

```
In[25]:= wilcoxon[dataC[16, 1, 1]]
```

```
Out[25]= 77
```

```
In[26]:= s[x_, c_] := 1 /; x ≥ c
```

```
s[x_, c_] := 0 /; x < c
```

```
In[28]:= ? s
```

Global`*s*

```
s[x_] := 1 /; x > 0
```

```
s[x_] := 0 /; x ≤ 0
```

```
s[x_, c_] := 1 /; x ≥ c
```

```
s[x_, c_] := 0 /; x < c
```

```
In[29]:= Table[s[wilcoxon[dataC[16, 1, 1]], cutW], {20}]
```

```
Out[29]= {0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1}
```

```
In[30]:= Sim[k_, n_, a_, b_, c_] :=
```

```
Module[{p, sig}, p = Sum[s[wilcoxon[dataC[n, a, b]], c], {k}] / k; {p, 2  $\sqrt{\frac{p(1-p)}{k}}$ }]
```

```
In[31]:= N[Sim[10000, 16, 0, 1, cutW]]
```

```
Out[31]= {0.107, 0.00618226}
```

```
In[32]:= dataN[n_, a_, b_] := RandomArray[NormalDistribution[a, b], n]
```

General::spell1 : スベル間違いの可能性があります。新規シンボル*dataN*はすでにあるシンボル*dataC*に似ています。 詳細

```
In[33]:= student[x_List] := Module[{n, b, s2}, n = Length[x]; b = Sum[x[[1]], {1, n}] / n;
```

```
s2 = Sum[x[[1]]^2, {1, n}] / n; s2 = (s2 - b^2)  $\frac{n}{n-1}$ ; b  $\sqrt{\frac{n}{s2}}$ ]
```

```
In[34]:= Quantile[NormalDistribution[0, 1], 0.95]
```

```
Out[34]= 1.64485
```

```
In[35]:= CDF[NormalDistribution[0, 1], %]
```

```
Out[35]= 0.95
```

```
In[36]:= cS[n_, a_] := Quantile[StudentTDistribution[n - 1], a]
```

```
In[37]:= cutS = cS[16, 1 - 0.105]
```

General::spell1 : スベル間違いの可能性がありますが、新規シンボル"cutS"はすでにあるシンボル"cutW"に似ています。 詳細

```
Out[37]= 1.30971
```

```
In[38]:= Sim[k_, n_, a_, b_, c_, test_, data_] :=
```

```
Module[{p, sig}, p = Sum[s[test[data[n, a, b]], c], {k}] / k; {p, 2  $\sqrt{\frac{p(1-p)}{k}}$ }]
```

```
In[39]:= N[Sim[10000, 16, 0, 1, cutW, wilcoxon, dataC]]
```

```
Out[39]= {0.111, 0.00628264}
```

```
In[40]:= N[Sim[10000, 16, 0, 1, cutS, student, dataN]]
```

```
Out[40]= {0.1086, 0.00622273}
```

```
In[41]:= N[Sim[10000, 16, 0, 1, cutW, wilcoxon, dataN]]
```

```
Out[41]= {0.1049, 0.00612849}
```

```
In[42]:= N[Sim[10000, 16, 0.5, 1, cutS, student, dataN]]
```

```
Out[42]= {0.7509, 0.00864984}
```

```
In[43]:= cB[n_, a_] := Module[{i, p, d},
```

```
  i = n;
```

```
  d = 2^n;
```

```
  p = Binomial[n, i] / d;
```

```
  While[p < a, i = i - 1; p = p + Binomial[n, i] / d];
```

```
  {i, p}]
```

```
In[44]:= cB[16, 0.105]
```

```
Out[44]= {11,  $\frac{6885}{65536}$ }
```

```
In[45]:= N[%]
```

```
Out[45]= {11., 0.105057}
```

```
In[46]:= N[Sum[Binomial[16, i], {i, 11, 16}] / 2^16]
```

```
Out[46]= 0.105057
```

```
In[47]:= cutT = 11
```

General::spell : スベル間違いの可能性がありますが、新規シンボル"cutT"はすでにあるシンボル{cutS, cutW}に似ています。 詳細

```
Out[47]= 11
```

```
In[48]:= tukey[x_List] := Sum[s[x[[1]]], {1, Length[x]}]
```

```
In[49]:= x = dataN[16, 0, 1]
```

```
Out[49]= {-0.856295, 1.6623, -1.76112, -0.575696, 1.05413, 0.574969, -0.108474, 0.465186,  
1.01578, 1.16251, 1.98214, 0.540122, 0.416425, 0.441777, 0.456553, -0.41737}
```

```
In[50]:= tukey[x]
```

```
Out[50]= 11
```

```
In[51]:= N[Sim[10000, 16, 0, 1, outT, tukey, dataN]]
```

```
Out[51]= {0.1073, 0.00618989}
```

```
In[52]:= N[Sim[10000, 16, 0.5, 1, outT, tukey, dataN]]
```

```
Out[52]= {0.6226, 0.00969473}
```