

# Brunner-Munzel検定について

- 黒木玄
- 2022-08-05, 2022-09-19, 2023-09-13

## 文献

- E. Brunner and U. Munzel. The nonparametric Behrens-Fisher problem: Asymptotic theory and a small-sample approximation. *Biometrical Journal*, 42:17–25, 2000. [pdf ([https://www.researchgate.net/profile/Edgar-Brunner/publication/264799502\\_Nonparametric\\_Hypotheses\\_and\\_Rank\\_Statistics\\_for\\_Unbalanced\\_Factorial\\_Designs/links/5f/Hypotheses-and-Rank-Statistics-for-Unbalanced-Factorial-Designs.pdf](https://www.researchgate.net/profile/Edgar-Brunner/publication/264799502_Nonparametric_Hypotheses_and_Rank_Statistics_for_Unbalanced_Factorial_Designs/links/5f/Hypotheses-and-Rank-Statistics-for-Unbalanced-Factorial-Designs.pdf))]
- Karin Neubert and Edgar Brunner, A studentized permutation test for the non-parametric Behrens-Fisher problem, *Computational Statistics and Data Analysis*, Vol. 51, pp. 5192-5204 (2007). <https://doi.org/10.1016/j.csda.2006.05.024> (<https://doi.org/10.1016/j.csda.2006.05.024>)
- Claus P. Nowak, Markus Pauly, Edgar Brunner. The nonparametric Behrens-Fisher problem in small samples. <https://arxiv.org/abs/2208.01231> (<https://arxiv.org/abs/2208.01231>)

## 目次

- ▼ 1 準備
  - [1.1 パッケージの読み込みなど](#)
  - [1.2 組み合わせの生成子](#)
  - [1.3 Welchのt検定](#)
  - [1.4 単峰型の函数が正の値になる場所を見つける函数](#)
  - [1.5 2つの分布が「互角」になるシフトの仕方を求める函数](#)
- ▼ 2 Brunner-Munzel検定
  - [2.1 Brunner-Munzel検定の実装](#)
  - [2.2 よく使われているっぽいテストデータで正しく実装されているかを確認](#)
  - [2.3 組み合わせの生成子](#)
  - [2.4 Brunner-Munzel検定のpermutation版の実装](#)
  - [2.5 permutation版が正しく実装されているかの確認](#)
- 3 計算例
- ▼ 4 Brunner-Munzel検定とWelchのt検定の比較
  - [4.1 第一種の過誤の確率](#)
  - [4.2 Brunner-Munzel検定は中央値に関する検定ではないことの証拠](#)
  - [4.3 BM検定による互角シフトの信頼区間とWelchのt検定による平均の差の信頼区間の比較](#)
- 5 小サンプルでのpermutation版の検定とBM検定とWelchのt検定の比較

## 1 準備

### 1.1 パッケージの読み込みなど

```
In [1]: 1 using Base.Threads
2 using BenchmarkTools
3 using Distributions
4 using PrettyPrinting
5 using QuadGK
6 using Random
7 using RCall
8 using Roots
9 using StatsBase
10 using StatsFuns
11 using StatsPlots
12 default(fmt=:png, size=(400, 250),
13         titlefontsize=10, guidefontsize=8, tickfontsize=6)
14
15 x ≈ y = x < y || x ≈ y
16 x ≈ y = x > y || x ≈ y
17 safemul(x, y) = x == 0 ? x : isnan(x) ? typeof(x)(Inf) : x*y
18 safediv(x, y) = x == 0 ? x : isnan(y) ? zero(y) : x/y
```

Out[1]: safediv (generic function with 1 method)

## 1.2 組み合わせの生成子

```
In [2]:  1 """
2     nextcombination!(n, t, c = typeof(t)[min(t-1, i) for i in 1:t])
3
4 '[1,2,...,n]' からの重複無しの 't' 個の組み合わせ 'c' をすべて生成したい。
5
6 'nextcombination!(n, t, c)' は配列で表現された組み合わせ 'c' をその次の組み合わせに書き換
7
8 初期条件を 'c = typeof(t)[min(t-1, i) for i in 1:t]' にすると,
9 'binomial(n, t)' 回の 'nextcombination!(n, t, c)' ですべての組み合わせが生成される。
10 """
11 function nextcombination!(n, t, c = typeof(t)[min(t-1, i) for i in 1:t])
12     t == 0 && return c
13     @inbounds for i in t:-1:1
14         c[i] += 1
15         c[i] > (n - (t - i)) && continue
16         for j in i+1:t
17             c[j] = c[j-1] + 1
18         end
19         break
20     end
21     c
22 end
23
24 """
25 mycombinations!(n::Integer, t, c)
26
27 事前に割り当てられた組み合わせを格納する配列 'c' を使って,
28 '[1,2,...,n]' からの重複無しの 't' 個の組み合わせのすべてを生成する生成子を返す。
29 """
30 function mycombinations!(n::Integer, t, c)
31     for i in 1:t c[i] = min(t - 1, i) end
32     (nextcombination!(n, t, c) for _ in 1:binomial(n, t))
33 end
34
35 """
36 mycombinations!(a, t, c)
37
38 事前に割り当てられた組み合わせを格納する配列 'c' を使って,
39 配列 'a' からのインデックスに重複がない 't' 個の組み合わせのすべてを生成する生成子を返す。
40 """
41 function mycombinations!(a, t, c)
42     t < 0 && (t = length(a) + 1)
43     (view(a, indices) for indices in mycombinations!(length(a), t, c))
44 end
45
46 """
47 mycombinations(x, t)
48
49 'x' が整数ならば '[1,2,...,x]' からの, 'x' が配列ならば 'x' からの,
50 インデックスに重複がない 't' 個の組み合わせのすべてを生成する生成子を返す。
51 """
52 mycombinations(x, t) = mycombinations!(x, t, Vector{typeof(t)}(undef, t))
```

Out[2]: mycombinations

### 1.3 Welchのt検定

```
In [3]: M
1 function tvalue_welch(m, x̄, sx², n, ȳ, sy²; Δμ=0)
2     (x̄ - ȳ - Δμ) / √(sx²/m + sy²/n)
3 end
4
5 function tvalue_welch(x, y; Δμ=0)
6     m, x̄, sx² = length(x), mean(x), var(x)
7     n, ȳ, sy² = length(y), mean(y), var(y)
8     tvalue_welch(m, x̄, sx², n, ȳ, sy²; Δμ)
9 end
10
11 function degree_of_freedom_welch(m, sx², n, sy²)
12     (sx²/m + sy²/n)^2 / ((sx²/m)^2/(m-1) + (sy²/n)^2/(n-1))
13 end
14
15 function degree_of_freedom_welch(x, y)
16     m, sx² = length(x), var(x)
17     n, sy² = length(y), var(y)
18     degree_of_freedom_welch(m, sx², n, sy²)
19 end
20
21 function pvalue_welch(m, x̄, sx², n, ȳ, sy²; Δμ=0)
22     t = tvalue_welch(m, x̄, sx², n, ȳ, sy²; Δμ)
23     v = degree_of_freedom_welch(m, sx², n, sy²)
24     2ccdf(TDist(v), abs(t))
25 end
26
27 function pvalue_welch(x, y; Δμ=0)
28     m, x̄, sx² = length(x), mean(x), var(x)
29     n, ȳ, sy² = length(y), mean(y), var(y)
30     pvalue_welch(m, x̄, sx², n, ȳ, sy²; Δμ)
31 end
32
33 function confint_welch(m, x̄, sx², n, ȳ, sy²; α=0.05)
34     v = degree_of_freedom_welch(m, sx², n, sy²)
35     c = quantile(TDist(v), 1-α/2)
36     SEhat = √(sx²/m + sy²/n)
37     [x̄-ȳ-c*SEhat, x̄-ȳ+c*SEhat]
38 end
39
40 function confint_welch(x, y; α=0.05)
41     m, x̄, sx² = length(x), mean(x), var(x)
42     n, ȳ, sy² = length(y), mean(y), var(y)
43     confint_welch(m, x̄, sx², n, ȳ, sy²; α)
44 end
```

Out[3]: confint\_welch (generic function with 2 methods)

### 1.4 単峰型の函数が正の値になる場所を見つける函数

```
In [4]: M
1 function findpositive(f, a, b; maxsplit = 30)
2     @assert f(a) < 0
3     @assert f(b) < 0
4     c = (a + b)/2
5     f(c) > 0 && return c
6     w = b - a
7     for k in 2:maxsplit
8         for d in range(w/2^(k+1), w/2-w/2^(k+1), step=w/2^k)
9             x = c + d
10            f(x) > 0 && return x
11            x = c - d
12            f(x) > 0 && return x
13        end
14    end
15    error("k > maxsplit = $maxsplit")
16 end
17
18 f(x) = abs(x) < 1e-4 ? 1.0 : -1.0
19
20 @time findpositive(f, -100abs(randn()), 20abs(randn()))
```

0.000373 seconds

Out[4]: -5.2804658903227164e-5

Loading web-font STIX-Web/Size1/Regular

## 1.5 2つの分布が「互角」になるシフトの仕方を求める函数

In [5]:

```
1 """
2     prob_x_le_y(distx::UnivariateDistribution, disty::UnivariateDistribution;
3         a = 0.0)
4
5     この函数は，連続分布 `distx`，`disty` と実数 `a` について，
6     `distx` と `disty` に従って生成される乱数をそれぞれ X, Y と書くとき，
7      $X \leq Y + a$  が成立する確率を返す。
8 """
9 function prob_x_le_y(distx::UnivariateDistribution, disty::UnivariateDistribution,
10     a = 0.0)
11     H(y) = cdf(distx, y) * pdf(disty, y-a)
12     quadgk(H, extrema(disty + a)...)[1]
13 end
14
15 """
16 tieshift(distx::UnivariateDistribution, disty::UnivariateDistribution;
17     p = 0.5)
18
19     この函数は，連続分布 `distx`，`disty` と実数 `p` について，
20     `distx` と `disty` に従って生成される乱数をそれぞれ X, Y と書くとき，
21      $X \leq Y + a$  が成立する確率が `p` に等しくなるような実数 a を返す。
22 """
23 function tieshift(distx::UnivariateDistribution, disty::UnivariateDistribution;
24     p=0.5)
25     find_zero(a → prob_x_le_y(distx, disty, a) - p, 0.0)
26 end
27
28 @show tieshift(Normal(0, 1), Normal(2, 2))
29 @show tieshift(Normal(0, 1), Laplace(2, 2))
30 @show tieshift(Normal(0, 1), Uniform(0, 1));
```

```
tieshift(Normal(0, 1), Normal(2, 2)) = -1.999999999999923
tieshift(Normal(0, 1), Laplace(2, 2)) = -1.999999999999945
tieshift(Normal(0, 1), Uniform(0, 1)) = -0.499999999999999
```

## **2 Brunner-Munzel検定**

### **2.1 Brunner-Munzel検定の実装**

In [6]:

```

1 """
2     h_brunner_munzel(x, y)
3
4 この函数は, x < y のとき 1.0 を, x = y のとき 0.5 を, それら以外のとき 0.0 返す.
5 """
6 h_brunner_munzel(x, y) = (x < y) + (x == y)/2
7
8 @doc raw"""
9     phat_brunner_munzel(X, Y)
10
11 まず以下のようにおく:
12
13 ````math
14 \begin{aligned}
15 &
16 H(x, y) = \begin{cases} 1 & (x < y) \\ 1/2 & (x = y) \\ 0 & (x > y) \end{cases}, \end{aligned}
17 ````\\
18 m = \mathrm{length}(X), \quad
19 n = \mathrm{length}(Y), \quad
20 x_i = X[i], \quad
21 y_j = Y[j]
22 \end{aligned}
23 ````\\
24
25 この函数は次の  $\hat{p}$  を返す:
26
27 ````math
28 \hat{p} = \frac{1}{mn} \sum_{i=1}^m \sum_{j=1}^n H(x_i, y_j).
29 ````\\
30
31 phat_brunner_munzel(X, Y) = mean(h_brunner_munzel(x, y) for x in X, y in Y)
32
33 @doc raw"""
34     statistics_brunner_munzel(X, Y,
35         Hx = similar(X, Float64),
36         Hy = similar(Y, Float64);
37         p = 1/2
38     )
39
40 この函数はデータ 'X', 'Y' について, Brunner-Munzel検定関係の統計量達を計算する. 詳細は以
41
42 函数  $H(x, y)$  と  $\hat{p}$ ,  $H^x_i$ ,  $H^y_j$ ,  $\bar{H}^x$ ,  $\bar{H}^y$  を次のように定め:
43
44 ````math
45 \begin{aligned}
46 &
47 m = \mathrm{length}(X), \quad
48 n = \mathrm{length}(Y), \quad
49 x_i = X[i], \quad
50 y_j = Y[j],
51 ````\\
52 \hat{p} = \frac{1}{mn} \sum_{i=1}^m \sum_{j=1}^n H(x_i, y_j),
53 ````\\
54 H(x, y) = \begin{cases} 1 & (x < y) \\ 1/2 & (x = y) \\ 0 & (x > y) \end{cases},
55 ````\\
56 H^x_i = \sum_{j=1}^n H(y_j, x_i), \quad
57 H^y_j = \sum_{i=1}^m H(x_i, y_j),
58 ````\\
59 \bar{H}^x = \frac{1}{m} \sum_{i=1}^m H^x_i = n - n\hat{p},
60 ````\\
61 \bar{H}^y = \frac{1}{n} \sum_{j=1}^n H^y_j = m\hat{p}.
62 \end{aligned}
63 ````\\
64
65 この函数は以下達の named tuple で返す:
66
67 ````math
68 \begin{aligned}
69 &
70 \mathrm{phat} =
71 \hat{p} = \frac{\bar{H}^x - \bar{H}^y + n\{m + n\}}{m + n},
72 ````\\
73 \mathrm{sx2} =
74 \hat{\sigma}_x^2 = \frac{1}{n^2} \frac{1}{m-1} \sum_{i=1}^m (H^x_i - \bar{H}^x)^2,
75 ````\\
76 \mathrm{sy2} =
77 \hat{\sigma}_y^2 = \frac{1}{m^2} \frac{1}{n-1} \sum_{j=1}^n (H^y_j - \bar{H}^y)^2,
78 ````\\
79 \mathrm{sehat} =
80 \widehat{\mathrm{se}} = \sqrt{\frac{\hat{\sigma}_x^2}{m} + \frac{\hat{\sigma}_y^2}{n}},
81 \end{aligned}

```

```

81 \\ &
82 \mathrm{tvalue} = t = \frac{\hat{p} - p}{\widehat{\mathsf{se}}}, \\
83 \\ &
84 \mathrm{df} =
85 \nu =
86 \frac
87 {\left(\hat{\sigma}_x^2/m + \hat{\sigma}_y^2/n\right)^2}
88 \\
89 \frac{\left(\hat{\sigma}_x^2/m\right)^2}{m-1} +
90 \frac{\left(\hat{\sigma}_y^2/n\right)^2}{n-1}
91 },
92 \\ &
93 \mathrm{pvalue} =
94 2\mathrm{ccdf}(\mathrm{TDist}(\nu), |t|).
95 \end{aligned}
96 \\
97 """
98 function statistics_brunner_munzel(X, Y,
99     Hx = similar(X, Float64),
100    Hy = similar(Y, Float64);
101    p = 1/2
102 )
103 m, n = length(X), length(Y)
104 for (i, x) in pairs(X)
105     Hx[i] = sum(h_brunner_munzel(y, x) for y in Y)
106 end
107 for (j, y) in pairs(Y)
108     Hy[j] = sum(h_brunner_munzel(x, y) for x in X)
109 end
110 phat = (mean(Hy) - mean(Hx) + n)/(m + n)
111 sx2, sy2 = var(Hx)/n^2, var(Hy)/m^2
112 sehat = sqrt(sx2/m + sy2/n)
113 tvalue = (phat - p)/sehat
114 df = safediv((sx2/m + sy2/n)^2, (sx2/m)^2/(m-1) + (sy2/n)^2/(n-1))
115 pvalue = (df != 0 && isfinite(df)) ? 2ccdf(TDist(df), abs(tvalue)) : zero(df)
116 (; phat, sx2, sy2, sehat, tvalue, df, pvalue)
117 end
118 @doc raw"""
119     pvalue_brunner_munzel(X, Y,
120         Hx = similar(X, Float64),
121         Hy = similar(Y, Float64);
122         p = 1/2
123     )
124 """
125 この函数はBrunner-Munzel検定のP値 `pvalue` を返す.
126 """
127 function pvalue_brunner_munzel(X, Y,
128     Hx = similar(X, Float64),
129     Hy = similar(Y, Float64);
130     p = 1/2
131 )
132     statistics_brunner_munzel(X, Y, Hx, Hy; p).pvalue
133 end
134 """
135 """
136 tieshift(X::AbstractVector, Y::AbstractVector; p = 1/2)
137 """
138 この函数は `phat_brunner_munzel(X, Y .+ a)` の値が `p` に等しくなる `a` を返す.
139 """
140 function tieshift(X::AbstractVector, Y::AbstractVector; p = 1/2)
141     shiftmin = minimum(X) - maximum(Y) - 0.1
142     shiftmax = maximum(X) - minimum(Y) + 0.1
143     find_zero(a → phat_brunner_munzel(X, Y .+ a) - p, (shiftmin, shiftmax))
144 end
145 """
146 @doc raw"""
147     brunner_munzel(X, Y,
148         Hx = similar(X, Float64),
149         Hy = similar(Y, Float64),
150         Ytmp = similar(Y, Float64);
151         p = 1/2,
152         α = 0.05,
153         maxsplit = 30
154     )
155 """
156 この函数はBrunner-Munzel検定を実行する. 詳細は以下の通り.
157 """
158 """
159 この函数は `phat`, `sehat`, `tvalue`, `df`, `p`, `pvalue`, `α` および \
160 以下達の named tuple を返す.

```

```

161 ````math
162 \begin{aligned}
163 & \\
164 \mathit{confint\_p} = (\text{$p$ の信頼度 $1-\alpha$ の信頼区間}), \\
165 \\ &
166 \mathit{confint\_shift} = (\text{2つの集団が互角になるようなシフトの信頼度 $1-\alpha$ の信頼区間}), \\
167 \\ &
168 \mathit{pvalue\_shift} = (\text{$\mathit{confint\_shift}$ の計算で使われたP値函数}), \\
169 \\ &
170 \mathit{shiftthat} = (\text{2つの集団が互角になるようなシフトの点推定値}). \\
171 \end{aligned}
172 ````\\
173 \\
174 さらに, $\mathit{shiftmin}$, $\mathit{shiftmax}$ はデータから推定されるシフトの下限と上限
175 \\
176 """
177 function brunner_munzel(X, Y,
178     Hx = similar(X, Float64),
179     Hy = similar(Y, Float64),
180     Ytmp = similar(Y, Float64);
181     p = 1/2,
182     α = 0.05,
183     maxsplit = 30
184 )
185     (; phat, sehat, tvalue, df, pvalue) = statistics_brunner_munzel(X, Y, Hx, Hy; p)
186
187     c = df == 0 ? Inf : quantile(TDist(df), 1 - α/2)
188     confint_p = [max(0, phat - c*sehat), min(1, phat + c*sehat)]
189
190     function pvalue_shift(a)
191         @. Ytmp = Y + a
192         pvalue_brunner_munzel(X, Ytmp, Hx, Hy; p)
193     end
194     shiftmin = minimum(X) - maximum(Y) - 0.1
195     shiftmax = maximum(X) - minimum(Y) + 0.1
196     shiftthat = tieshift(X, Y; p)
197     confint_shift = [
198         find_zero(a → pvalue_shift(a) - α, (shiftmin, shiftthat))
199         find_zero(a → pvalue_shift(a) - α, (shiftthat, shiftmax))
200     ]
201
202     (; phat, sehat, tvalue, df, p, pvalue, α, confint_p,
203      confint_shift, pvalue_shift, shiftthat, shiftmin, shiftmax)
204 end
205
206
207 function show_plot_brunner_munzel(X, Y,
208     Hx = similar(X, Float64),
209     Hy = similar(Y, Float64),
210     Ytmp = similar(Y, Float64);
211     p = 1/2,
212     α = 0.05,
213     showXY = false,
214     kwargs...
215 )
216     showXY && (@show X Y)
217     (; phat, sehat, tvalue, df, p, pvalue, α, confint_p,
218      confint_shift, pvalue_shift, shiftthat, shiftmin, shiftmax) =
219         brunner_munzel(X, Y, Hx, Hy, Ytmp; p, α)
220     pprint(; phat, sehat, tvalue, df, p, pvalue, α, confint_p,
221         confint_shift, shiftthat))
222     println()
223     @show median(X) median(Y)
224     plot(pvalue_shift, shiftmin, shiftmax; label="")
225     vline!([tieshift(X, Y)]; label="", ls=:dash)
226     title!("P-value function of shift")
227     plot!(ytick=0:0.05:1)
228     plot!(; kwargs...)
229 end

```

Out[6]: show\_plot\_brunner\_munzel (generic function with 4 methods)

In [7]: 1 @doc h\_brunner\_munzel

Out[7]: \begin{verbatim}
h\_brunner\_munzel(x,y)
\end{verbatim}

In [8]: 1 @doc phat\_brunner\_munzel

Out[8]:

```
\begin{verbatim}
phat_brunner_munzel(X, Y)
\end{verbatim}
```

まず以下のようにおく:

$$H(x, y) = \begin{cases} 1 & (x < y) \\ 1/2 & (x = y) \\ 0 & (x > y), \end{cases}$$
$$m = \text{length}(X), \quad n = \text{length}(Y), \quad x_i = X[i], \quad y_j = Y[j]$$

この函数は次の  $\hat{p}$  を返す:

$$\hat{p} = \frac{1}{mn} \sum_{i=1}^m \sum_{j=1}^n H(x_i, y_j).$$

In [9]: 1 @doc statistics\_brunner\_munzel

Out[9]: 

```
\begin{verbatim} statistics_brunner_munzel(X, Y, Hx = similar(X, Float64), Hy = similar(Y, Float64); p = 1/2 ) \end{verbatim}
```

 この函数はデータ `X`, `Y` について, Brunner-Munzel検定関係の統計量達を計算する. 詳細は以下の通り. フункци  $H(x, y)$  と  $\hat{p}$ ,  $H^x_i$ ,  $H^y_j$ ,  $\bar{H}^x$ ,  $\bar{H}^y$  を次のように定める:  $\hat{p} = \frac{1}{mn} \sum_{i=1}^m \sum_{j=1}^n H(x_i, y_j)$ ,  $H(x, y) = \begin{cases} 1 & (x < y) \\ 1/2 & (x = y) \\ 0 & (x > y), \end{cases}$ ,  $H^x_i = \sum_{j=1}^n H(x_i, y_j)$ ,  $H^y_j = \sum_{i=1}^m H(x_i, y_j)$ ,  $\bar{H}^x = \frac{1}{m} \sum_{i=1}^m H^x_i$ ,  $\bar{H}^y = \frac{1}{n} \sum_{j=1}^n H^y_j$ . この函数は以下達の named tuple で返す:  $\hat{p}, \hat{\sigma}_x, \hat{\sigma}_y, \hat{\sigma}_{xy}, \text{sehat}, \text{tvalue}, \text{df}, \text{pvalue}$ .

In [10]: 1 @doc brunner\_munzel

Out[10]: 

```
brunner_munzel(X, Y,
    Hx = similar(X, Float64),
    Hy = similar(Y, Float64),
    Ytmp = similar(Y, Float64);
    p = 1/2,
    α = 0.05,
    maxsplit = 30
)
```

この函数はBrunner-Munzel検定を実行する. 詳細は以下の通り.

この函数は `phat`, `sehat`, `tvalue`, `df`, `p`, `pvalue`, `α` および  
以下達の named tuple を返す.

```
\begin{aligned}
& \mathbf{\hat{p}} = (\text{p 的信頼度 } 1 - \alpha \text{ の信頼区間}), \\
& \mathbf{\hat{\sigma}_x} = (\text{2つの集団が互角になるようなシフトの信頼度 } 1 - \alpha \text{ の信頼区間}), \\
& \mathbf{\hat{\sigma}_y} = (\text{2つの集団が互角になるようなシフトの信頼度 } 1 - \alpha \text{ の信頼区間}), \\
& \mathbf{\hat{\sigma}_{xy}} = (\text{2つの集団が互角になるようなシフトの信頼度 } 1 - \alpha \text{ の信頼区間}), \\
& \mathbf{\text{sehat}} = (\text{2つの集団が互角になるようなシフトの標準誤差}), \\
& \mathbf{\text{tvalue}} = (\text{2つの集団が互角になるようなシフトのt値}), \\
& \mathbf{\text{df}} = (\text{自由度}), \\
& \mathbf{\text{pvalue}} = (\text{P値})
\end{aligned}
```

さらに, `shiftmin`, `shiftmax`  
はデータから推定されるシフトの下限と上限.

In [11]:

```
1 X = randn(10)
2 Y = randn(10)
3 @show shiftmin = minimum(X) - maximum(Y) - 1
4 @show shiftmax = maximum(X) - minimum(Y) + 1
5 pvalue_brunner_munzel(X, Y)
```

```
shiftmin = (minimum(X) - maximum(Y)) - 1 = -4.810896555167297
shiftmax = (maximum(X) - minimum(Y)) + 1 = 3.8662791521107494
```

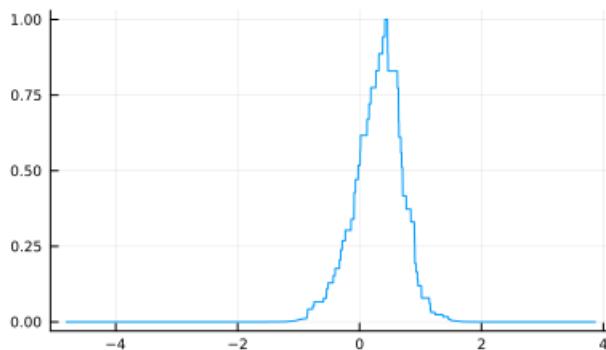
Out[11]: 0.5181701995353223

In [12]:

```
1 plot(a → pvalue_brunner_munzel(X, Y .+ a), shiftmin, shiftmax;
2     label="", title="P-value function of shift")
```

Out[12]:

P-value function of shift



## 2.2 よく使われているっぽいテストデータで正しく実装されているかを確認

<https://okumuralab.org/~okumura/stat/brunner-munzel.html> (<https://okumuralab.org/~okumura/stat/brunner-munzel.html>)

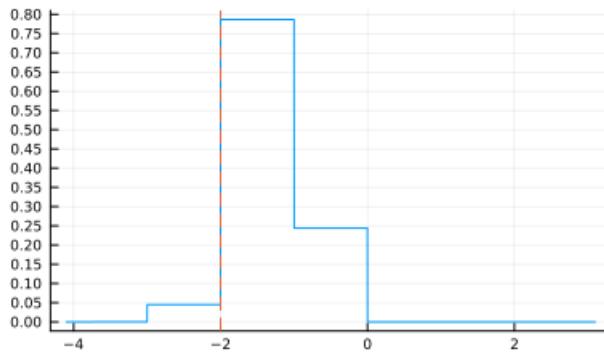
```
x = c(1,2,1,1,1,1,1,1,1,2,4,1,1)
y = c(3,3,4,3,1,2,3,1,1,5,4)
brunnermunzel.test(x, y)

data: x and y
Brunner-Munzel Test Statistic = 3.1375, df = 17.683, p-value = 0.005786
95 percent confidence interval:
 0.5952169 0.9827052
sample estimates:
P(X<Y)+.5*P(X=Y)
 0.788961
```

```
In [13]: X = [1,2,1,1,1,1,1,1,1,2,4,1,1]
          Y = [3,3,4,3,1,2,3,1,1,5,4]
          show_plot_brunner_munzel(X, Y)
```

```
(phat = 0.788961038961039,
 sehat = 0.09210009046816862,
 tvalue = 3.1374674823029505,
 df = 17.682841979481545,
 p = 0.5,
 pvalue = 0.005786208666151463,
 α = 0.05,
 confint_p = [0.5952168642537363, 0.9827052136683416],
 confint_shift = [-2.0000000000000004, -5.551115123125783e-17],
 shifthat = -1.9999999999999998)
median(X) = 1.0
median(Y) = 3.0
```

Out[13]: P-value function of shift



```
In [14]: X = [1,2,1,1,1,1,1,1,1,2,4,1,1]
          Y = [3,3,4,3,1,2,3,1,1,5,4]
          @rput X Y
          R"""
          library(lawstat)
          brunner.munzel.test(X, Y)
          """
```

[ Warning: RCall.jl: Warning: package 'lawstat' was built under R version 4.4.1  
@ RCall D:\.julia\packages\RCall\dDAVd\src\io.jl:172

Out[14]: RObject{VecSxp}

```
Brunner-Munzel Test

data: X and Y
Brunner-Munzel Test Statistic = 3.1375, df = 17.683, p-value = 0.005786
95 percent confidence interval:
 0.5952169 0.9827052
sample estimates:
P(X<Y)+.5*P(X=Y)
 0.788961
```

このように Brunner-Munzel 検定は R では [lawstat](https://cran.r-project.org/package=lawstat) (<https://cran.r-project.org/package=lawstat>) パッケージの [brunner.munzel.test](https://rdrr.io/cran/lawstat/man/brunner.munzel.test.html) (<https://rdrr.io/cran/lawstat/man/brunner.munzel.test.html>) で使える。

## 2.3 組み合わせの生成子

In [15]:

```
1 """
2     complementcomb!(complcomb::AbstractVector, comb::AbstractVector)
3
4     'comb' が {1,2,...,N} から重複無しに m 個を選ぶ組み合わせを表す配列であり,
5     'comb' の中で数は小さな順に並んでいるとし, 'complcomb' は長さ N - m の配列であると仮定する
6
7     このとき, この函数は配列 'complcomb' に配列 'comb' の補集合を格納し, 'complcomb' を返す.
8
9     この函数はメモリ割り当てゼロで実行される.
10 """
11 function complementcomb!(complcomb::AbstractVector, comb::AbstractVector)
12     N = length(comb) + length(complcomb)
13     k = 0
14     a = 0
15     @inbounds for b in comb
16         for i in a+1:b-1
17             k += 1
18             complcomb[k] = i
19         end
20         a = b
21     end
22     @inbounds for i in a+1:N
23         k += 1
24         complcomb[k] = i
25     end
26     complcomb
27 end
28 """
29     complementcomb(N, comb::AbstractVector)
30
31     'comb' が {1,2,...,N} から重複無しに m 個を選ぶ組み合わせを表す配列であり,
32     'comb' の中で数は小さな順に並んでいると仮定する.
33
34     この函数は 'comb' の補集合の配列を返す.
35
36     この函数は返り値の配列の分だけのメモリ割り当てを行う.
37 """
38 complementcomb(N, comb::AbstractVector) =
39     complementcomb!(similar(comb, N - length(comb)), comb)
```

Out[15]: complementcomb

In [16]:

```
1 @doc complementcomb!
```

Out[16]: \begin{verbatim}complementcomb!(complcomb::AbstractVector, comb::AbstractVector)\end{verbatim} \texttt{comb} が \{1,2,...,N\} から重複無しに m 個を選ぶ組み合わせを表す配列であり, \texttt{comb} の中で数は小さな順に並んでいるとし, \texttt{complcomb} は長さ N - m の配列であると仮定する. このとき, この函数は配列 \texttt{complcomb} に配列 \texttt{comb} の補集合を格納し, \texttt{complcomb} を返す. この函数はメモリ割り当てゼロで実行される.

In [17]:

```
1 @doc complementcomb
```

Out[17]: \begin{verbatim}complementcomb(N, comb::AbstractVector)\end{verbatim} \texttt{comb} が \{1,2,...,N\} から重複無しに m 個を選ぶ組み合わせを表す配列であり, \texttt{comb} の中で数は小さな順に並んでいると仮定する. この函数は \texttt{comb} の補集合の配列を返す. この函数は返り値の配列の分だけのメモリ割り当てを行う.

In [18]:

```
1 N = 10
2 comb = [2, 4, 5, 8]
3 ccomb = similar(comb, N - length(comb))
4 @btime complementcomb!($ccomb, $comb);
```

14.629 ns (0 allocations: 0 bytes)

```
In [19]: N
1 N, m = 5, 3
2 ccomb = Vector{Int}(undef, N-m)
3 [(copy(comb), copy(complementcomb!(ccomb, comb))) for comb in mycombinations(1:N, m)]
```

```
Out[19]: 10-element Vector{Tuple{Vector{Int64}, Vector{Int64}}}:
([1, 2, 3], [4, 5])
([1, 2, 4], [3, 5])
([1, 2, 5], [3, 4])
([1, 3, 4], [2, 5])
([1, 3, 5], [2, 4])
([1, 4, 5], [2, 3])
([2, 3, 4], [1, 5])
([2, 3, 5], [1, 4])
([2, 4, 5], [1, 3])
([3, 4, 5], [1, 2])
```

```
In [20]: N
1 N, m = 5, 3
2 ccomb = Vector{Int}(undef, N-m)
3 [(copy(comb), complementcomb(N, comb)) for comb in mycombinations(1:N, m)]
```

```
Out[20]: 10-element Vector{Tuple{Vector{Int64}, Vector{Int64}}}:
([1, 2, 3], [4, 5])
([1, 2, 4], [3, 5])
([1, 2, 5], [3, 4])
([1, 3, 4], [2, 5])
([1, 3, 5], [2, 4])
([1, 4, 5], [2, 3])
([2, 3, 4], [1, 5])
([2, 3, 5], [1, 4])
([2, 4, 5], [1, 3])
([3, 4, 5], [1, 2])
```

## 2.4 Brunner-Munzel検定のpermutation版の実装

In [21]:

```
1 """
2     permutation_tvalues_brunner_munzel(X, Y,
3         XandY = Vector{Float64}(undef, length(X)+length(Y)),
4         Tval = Vector{Float64}(undef, binomial(length(X)+length(Y), length(X))),
5         Hx = similar(X, Float64),
6         Hy = similar(Y, Float64)
7     )
8
9 Brunner-Munzel 検定のt値を `'[X; Y]'` から \
10 インデックスの重複無しに `length(X)` 個取る組み合わせと \
11 その補集合への分割のすべてについて計算して, 'Tval' に格納して返す.
12 """
13 function permutation_tvalues_brunner_munzel(X, Y,
14     XandY = Vector{Float64}(undef, length(X)+length(Y)),
15     Tval = Vector{Float64}(undef, binomial(length(X)+length(Y), length(X))),
16     Hx = similar(X, Float64),
17     Hy = similar(Y, Float64),
18     ccomb = Vector{Int}(undef, length(Y))
19     )
20     m, n = length(X), length(Y)
21     N = m + n
22     @views XandY[1:m] .= X
23     @views XandY[m+1:N] .= Y
24     for (k, comb) in enumerate(mycombinations(1:N, m))
25         complementcomb!(ccomb, comb)
26         Tval[k] = statistics_brunner_munzel(
27             view(XandY, comb), view(XandY, ccomb), Hx, Hy).tvalue
28     end
29     Tval
30 end
31 """
32 """
33 pvalue_brunner_munzel_perm(X, Y,
34     Tval = permutation_tvalues_brunner_munzel(X, Y),
35     tval = statistics_brunner_munzel(X, Y).tvalue;
36     le = ≈
37     )
38
39 Brunner-Munzel 検定のpermutation版のP値を返す.
40 """
41 function pvalue_brunner_munzel_perm(X, Y,
42     Tval = permutation_tvalues_brunner_munzel(X, Y),
43     tval = statistics_brunner_munzel(X, Y).tvalue;
44     le = ≈
45     )
46     pvalue_perm = mean(T → le(abs(tval), abs(T)), Tval)
47 end
```

Out[21]: pvalue\_brunner\_munzel\_perm

In [22]:

```
1 @doc permutation_tvalues_brunner_munzel
```

Out[22]: \begin{verbatim} permutation\_tvalues\_brunner\_munzel(X, Y, XandY = Vector{Float64}(undef, length(X)+length(Y)), Tval = Vector{Float64}(undef, binomial(length(X)+length(Y), length(X))), Hx = similar(X, Float64), Hy = similar(Y, Float64) ) \end{verbatim} Brunner-Munzel検定のt値を \texttt{[X; Y]} からインデックスの重複無しに \texttt{length(X)} 個取る組み合わせとその補集合への分割のすべてについて計算して, \texttt{Tval} に格納して返す.

In [23]:

```
1 @doc pvalue_brunner_munzel_perm
```

Out[23]: \begin{verbatim} pvalue\_brunner\_munzel\_perm(X, Y, Tval = permutation\_tvalues\_brunner\_munzel(X, Y), tval = statistics\_brunner\_munzel(X, Y).tvalue; le = ≈ ) \end{verbatim} \end{verbatim} Brunner-Munzel検定のpermutation版のP値を返す.

<https://okumuralab.org/~okumura/stat/brunner-munzel.html> (<https://okumuralab.org/~okumura/stat/brunner-munzel.html>)

```

bm = brunner.munzel.test(x, y)$statistic
n1 = length(x)
n2 = length(y)
N = n1 + n2
xandy = c(x, y)
foo = function(x) {

```

In [24]:

```

1 X = [1,2,1,1,1,1,1,1,1,2,4,1,1]
2 Y = [3,3,4,3,1,2,3,1,1,5,4]
3 @show X Y
4 @show m, n = length(X), length(Y)
5
6 Tval = @time permutation_tvalues_brunner_munzel(X, Y)
7 @show pvalue_brunner_munzel_perm(X, Y, Tval)
8 stephist(Tval; norm=true, bin=101, label="", title="permutation t-values")

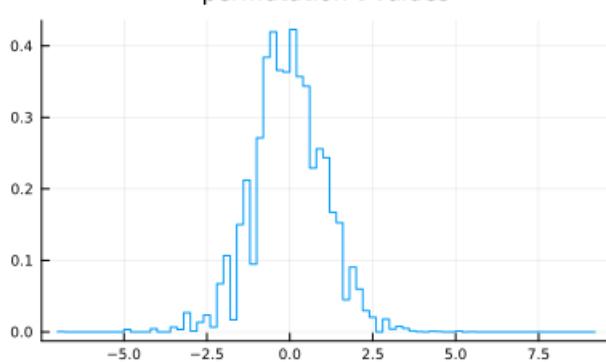
```

```

X = [1, 2, 1, 1, 1, 1, 1, 1, 1, 2, 4, 1, 1]
Y = [3, 3, 4, 3, 1, 2, 3, 1, 1, 5, 4]
(m, n) = (length(X), length(Y)) = (14, 11)
4.369441 seconds (2.91 M allocations: 798.552 MiB, 2.66% gc time, 6.21% compilation time)
pvalue_brunner_munzel_perm(X, Y, Tval) = 0.008037645264055279

```

Out[24]:



## 2.5 permutation版が正しく実装されているかの確認

- <https://github.com/toshi-ara/brunnermunzel/issues/14> (<https://github.com/toshi-ara/brunnermunzel/issues/14>)
- <https://github.com/toshi-ara/brunnermunzel/files/4395032/mwe.R.zip> (<https://github.com/toshi-ara/brunnermunzel/files/4395032/mwe.R.zip>)

追記 2022-08-06: <https://twitter.com/TA25140989/status/1555825941451923457>  
(<https://twitter.com/TA25140989/status/1555825941451923457>) を参考せよ.

<https://github.com/toshi-ara/brunnermunzel/tree/development> (<https://github.com/toshi-ara/brunnermunzel/tree/development>) の修正版の brunnermunzel パッケージをインストールし直した. 以下のセルの実行結果が変わらはずなので, その記録を残しておく.

以下の記録を見なくとも,

- <https://github.com/genkuroki/public/blob/e4faafc52721b63876b3b705f9450eade3c902f5/0034/Brunner-Munzel.ipynb> (<https://github.com/genkuroki/public/blob/e4faafc52721b63876b3b705f9450eade3c902f5/0034/Brunner-Munzel.ipynb>)

で閲覧できるが, わざわざ見に行くのも面倒なのでこのファイルにも記録を残しておく.

以前の実行結果:

```
@show pval_J - pval_J_le;
```

```
idx = @show findall(pval_J .!= pval_J_le)  
length(idx)
```

```
findall(pval_J .!= pval_J_le) = [3, 4, 5, 6, 9, 10, 12, 13, 15, 19, 21, 22, 23, 25, 27, 3  
2, 34, 36, 40, 42, 44, 47, 48, 49, 50, 51, 53, 55, 68, 69, 71, 78, 79, 80, 82, 86, 89, 9  
0, 92, 94]
```

40

```
@show pval_R - pval_J_le;
```

```
idx = @show findall(pval_R .!= pval_J_le)
length(idx)
```

```
findall(pval_R .!= pval_J_le) = [3, 4, 5, 6, 9, 10, 12, 13, 15, 19, 21, 22, 23, 25, 27, 3  
2, 34, 36, 40, 42, 44, 47, 48, 49, 50, 51, 53, 55, 68, 69, 78, 79, 80, 82, 86, 89, 90, 9  
2, 94]
```

39

```
@show pval_J - pval_R;
```

```
idx = @show findall(.!(pval_J .≈ pval_R))  
length(idx)
```

```
findall(.!(pyal_J, ≈ pyal_R)) = [22, 34, 71, 78]
```

4

```
all(pval >= pval_B & pval <= pval_L[e])
```

true

なるほど！

<https://github.com/toshi-ara/brunnermunzel/issues/14>

に書いてあるように, 22, 34, 71 and 78 の4つで, 値が一致していない.

○○以下または○○以上の判定を `$x \approx y$` のときも `true` にする必要があるのだが、その部分で違いが生じているものと思われる。

現時点では <https://CRAN.R-project.org/package=brunnermunzel> にアクセスすると、

```
>Package 'brunnermunzel' was removed from the CRAN repository.  
>  
>Formerly available versions can be obtained from the [archive](https://cran.r-project.org
```

```
In [25]: R"""
1 library(brunnermunzel)
2 set.seed(1290)
3 reps = 100
4 xx = c()
5 yy = c()
6 pval_R = numeric(reps)
7 for (i in seq_len(reps)){
8     x = rnorm(5)
9     y = rnorm(5)
10    xx = c(xx, x)
11    yy = c(yy, y)
12}
13 res_bm_perm <- brunnermunzel.permutation.test(x,y)
14 pval_R[i] <- res_bm_perm$p.value
15 }
16 """
17
18 @rget xx yy pval_R
19 XX = reshape(xx, 5, 100)
20 YY = reshape(yy, 5, 100)
21
22 pval_J = zeros(100)
23 pval_J_le = zeros(100)
24 for i in 1:100
25     pval_J[i] = pvalue_brunner_munzel_perm(XX[:,i], YY[:,i]; le = ≈)
26     pval_J_le[i] = pvalue_brunner_munzel_perm(XX[:,i], YY[:,i]; le = ≤)
27 end
```

[ Warning: RCall.jl: Warning: package 'brunnermunzel' was built under R version 4.4.1  
@ RCall D:\.julia\packages\RCall\dDAVD\src\io.jl:172

```
In [26]: 1 @show pval_J - pval_J_le;
```

pval\_J - pval\_J\_le = [0.0, 0.0, 0.007936507936507936, 0.023809523809523808, 0.023809523809523808, 0.0793650793650793, 0.0, 0.0, 0.007936507936507936, 0.0357142857142857, 0.0, 0.015873015873015872, 0.015873015873015872, 0.0, 0.015873015873015872, 0.0, 0.0, 0.0, 0.0, 0.007936507936507936, 0.0, 0.023809523809523836, 0.015873015873015928, 0.011904761904761918, 0.0, 0.015873015873015928, 0.0, 0.0, 0.003968253968253968, 0.0, 0.0, 0.0, 0.0, 0.007936507936507936, 0.0, 0.015873015873015928, 0.0, 0.015873015873015928, 0.0, 0.0, 0.0, 0.0, 0.007936507936507908, 0.0, 0.007936507936507908, 0.0, 0.007936507936507908, 0.0, 0.0, 0.0, 0.0, 0.007936507936507936, 0.007936507936507936, 0.011904761904761918, 0.007936507936507936, 0.03968253968253965, 0.0, 0.007936507936507936, 0.0, 0.0, 0.05555555555555547, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.015873015873015872, 0.015873015873015872, 0.0, 0.015873015873015928, 0.0, 0.0, 0.0, 0.0, 0.015873015873015928, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.015873015873015928, 0.007936507936507964, 0.007936507936507964, 0.0, 0.003968253968253954, 0.0, 0.0, 0.0, 0.015873015873015872, 0.0, 0.0, 0.0, 0.0357142857142857, 0.015873015873015872, 0.0, 0.023809523809523725, 0.0, 0.023809523809523808, 0.0, 0.0, 0.0, 0.0, 0.0]

```
In [27]: 1 idx = @show findall(pval_J .!= pval_J_le)
          2 length(idx)

findall(pval_J .!= pval_J_le) = [3, 4, 5, 6, 9, 10, 12, 13, 15, 19, 21, 22, 23, 25, 27, 32,
          34, 36, 40, 42, 44, 47, 48, 49, 50, 51, 53, 55, 68, 69, 71, 78, 79, 80, 82, 86, 89, 90, 92,
```

Out[87]: 48

```
In [28]: ┌ 1 @show pval_R - pval_J_le;
```

```
pval_R - pval_J_le = [0.0, 0.0, 0.007936507936507936, 0.023809523809523808, 0.023809523809523808, 0.0793650793650793, 0.0, 0.0, 0.007936507936507936, 0.0357142857142857, 0.0, 0.015873015873015828, 0.015873015873015872, 0.0, 0.015873015873015872, 0.0, 0.0, 0.0, 0.0, 0.007936507936507936, 0.0, 0.0, 0.023809523809523836, 0.015873015873015928, 0.011904761904761918, 0.0, 0.015873015873015928, 0.0, 0.0, 0.003968253968253968, 0.0, 0.0, 0.0, 0.0, 0.0, 0.007936507936507936, 0.0, 0.015873015873015928, 0.0, 0.015873015873015928, 0.0, 0.0, 0.0, 0.0, 0.007936507936507908, 0.0, 0.007936507936507908, 0.0, 0.007936507936507908, 0.0, 0.0, 0.007936507936507936, 0.007936507936507936, 0.007936507936507936, 0.011904761904761918, 0.007936507936507936, 0.03968253968253965, 0.0, 0.007936507936507936, 0.0, 0.05555555555555547, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.015873015873015872, 0.015873015873015872, 0.0, 0.015873015873015928, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.015873015873015928, 0.007936507936507964, 0.007936507936507964, 0.0, 0.003968253954, 0.0, 0.0, 0.0, 0.015873015873015872, 0.0, 0.0, 0.0, 0.0357142857142857, 0.015873015873015872, 0.0, 0.023809523809523725, 0.0, 0.023809523809523808, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0]
```

```
In [29]: 1 idx = @show findall(pval_R .!= pval_J_le)
          2 length(idx)
```

```
findall(pval_R .!= pval_J_le) = [3, 4, 5, 6, 9, 10, 12, 13, 15, 19, 21, 22, 23, 25, 27, 32, 34, 36, 40, 42, 44, 47, 48, 49, 50, 51, 53, 55, 68, 69, 71, 78, 79, 80, 82, 86, 89, 90, 92, 94]
```

Out[29]: 40

```
In [30]: 1 @show pval_J - pval_R;
```

```
In [31]: 1 idx = @show findall(.!(pval_J .≈ pval_R))
          2 length(idx)
```

```
findall(.!(pval_J .≈ pval_R)) = Int64[]
```

Out[31]: 0

```
In [32]: 1 all(pval_J .≥ pval_R .≥ pval_J_le)
```

Out[32]: true

2022-08-06: やった! 値が完全に一致した! permutation版Brunner-Munzel検定について、

- <https://github.com/toshi-ara/brunnermunzel/tree/development> (<https://github.com/toshi-ara/brunnermunzel/tree/development>)

の実装と私による実装の計算結果は以上の場合において完全に一致している。

3 計算例

```
In [33]: 1 m, n = 10, 10
          2 X, Y = rand(Normal(0, 1), m), rand(Normal(0, 2), n)
          3 @show pval_brmu = pvalue_brunner_munzel(X, Y)
          4 @show pval_perm = pvalue_brunner_munzel_perm(X, Y);
```

```
pval_brmu = pvalue_brunner_munzel(X, Y) = 1.0  
pval_perm = pvalue_brunner_munzel_perm(X, Y) = 1.0
```

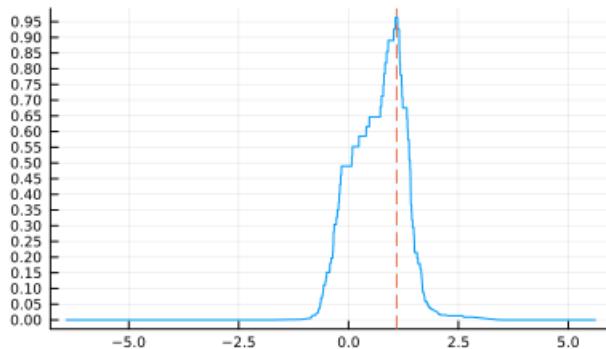
In [34]:

```
1 Random.seed!(4)
2
3 m, n = 10, 20
4 X, Y = rand(Normal(0, 1), m), rand(Normal(0, 2), n)
5 show_plot_brunner_munzel(X, Y)
```

```
(phat = 0.425,
sehat = 0.10709480961098924,
tvalue = -0.7003140513758763,
df = 27.239689302515885,
p = 0.5,
pvalue = 0.4896699803996247,
α = 0.05,
confint_p = [0.20535003891236225, 0.6446499610876377],
confint_shift = [-0.642106099966724, 1.7940553106227974],
shiftthat = 1.09533347422796)
median(X) = 0.4366097489932352
median(Y) = -0.8104755143666504
```

Out[34]:

P-value function of shift



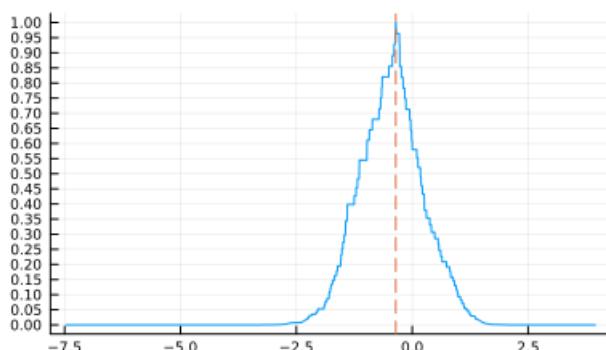
In [35]:

```
1 m, n = 10, 20
2 X, Y = rand(Normal(0, 1), m), rand(Normal(0, 2), n)
3 show_plot_brunner_munzel(X, Y)
```

```
(phat = 0.5549999999999999,
sehat = 0.10651441985678231,
tvalue = 0.5163620106456207,
df = 27.01715719647146,
p = 0.5,
pvalue = 0.6098024812257896,
α = 0.05,
confint_p = [0.33645695680473703, 0.7735430431952628],
confint_shift = [-2.032882441018063, 1.1693659401747338],
shiftthat = -0.3550706804781761)
median(X) = -0.17035935201789232
median(Y) = -0.16549665940751157
```

Out[35]:

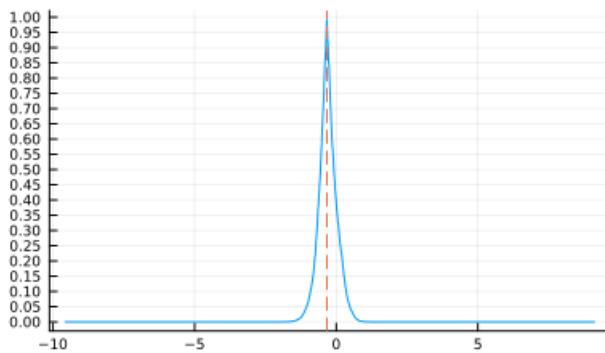
P-value function of shift



```
In [36]: 1 m, n = 100, 50
          2 X, Y = rand(Normal(0, 1), m), rand(Normal(0, 2), n)
          3 show_plot_brunner_munzel(X, Y)
```

```
(phat = 0.5518,
sehat = 0.059995235294905844,
tvalue = 0.8634018975903285,
df = 55.799136185276105,
p = 0.5,
pvalue = 0.39161265935808287,
α = 0.05,
confint_p = [0.431605553791199, 0.6719944462088009],
confint_shift = [-1.0466768697894908, 0.480114930555916],
shiftthat = -0.32905087954292944)
median(X) = -0.044389033553791826
median(Y) = 0.35316051685566774
```

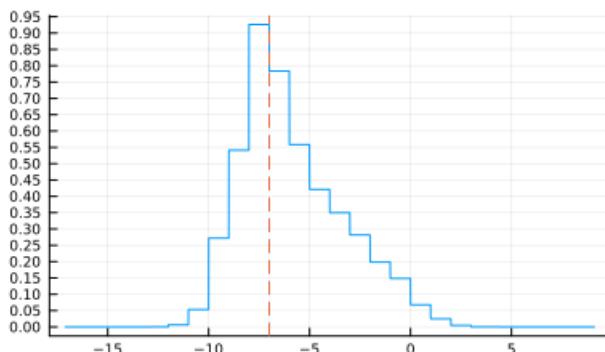
Out[36]: P-value function of shift



```
In [37]: 1 Random.seed!(4)
          2
          3 m, n = 10, 20
          4 X, Y = rand(1:m, m), rand(1:n, n)
          5 show_plot_brunner_munzel(X, Y)
```

```
(phat = 0.6725,
sehat = 0.1005725567999867,
tvalue = 1.715179622439735,
df = 20.71835365424782,
p = 0.5,
pvalue = 0.1012319072053917,
α = 0.05,
confint_p = [0.4631746597119115, 0.8818253402880885],
confint_shift = [-10.999999999999998, 0.9999999999999997],
shiftthat = -7.000000000000001)
median(X) = 6.5
median(Y) = 14.5
```

Out[37]: P-value function of shift



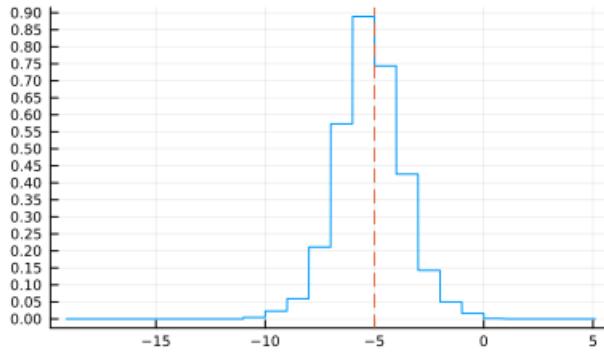
In [38]:

```
1 m, n = 10, 20
2 X, Y = rand(1:m, m), rand(1:n, n)
3 show_plot_brunner_munzel(X, Y)

(phat = 0.765,
sehat = 0.08626255311649104,
tvalue = 3.0720166564295583,
df = 27.60349766038036,
p = 0.5,
pvalue = 0.0047415464916824275,
α = 0.05,
confint_p = [0.5881847509519754, 0.9418152490480246],
confint_shift = [-8.999999999999998, -1.999999999999993],
shiftthat = -5.0)
median(X) = 5.5
median(Y) = 12.0
```

Out[38]:

P-value function of shift



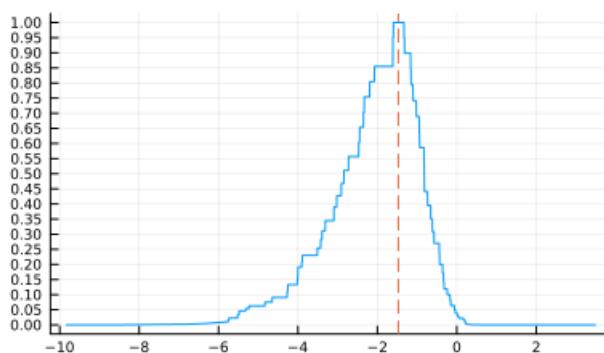
In [39]:

```
1 distx, disty = LogNormal(), LogNormal(1)
2 m, n = 10, 10
3 X, Y = rand(distx, m), rand(disty, n)
4 show_plot_brunner_munzel(X, Y)

(phat = 0.75,
sehat = 0.11205157542647741,
tvalue = 2.231115439907736,
df = 18.0,
p = 0.5,
pvalue = 0.03863103202434312,
α = 0.05,
confint_p = [0.5145883755427825, 0.9854116244572175],
confint_shift = [-5.318179432610124, -0.05841508479073754],
shiftthat = -1.4733410643825495)
median(X) = 1.3089930416807416
median(Y) = 2.9790059563220423
```

Out[39]:

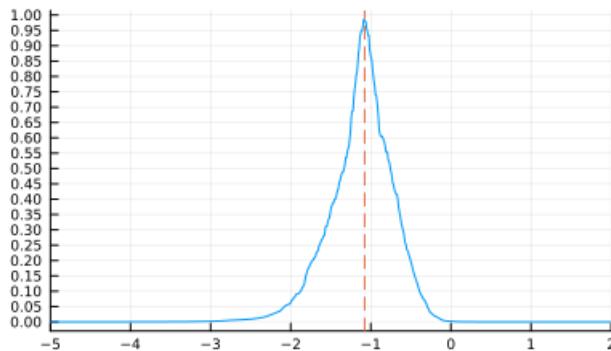
P-value function of shift



```
In [40]: M 1 distx, disty = LogNormal(), LogNormal(1)
2 m, n = 40, 40
3 X, Y = rand(distx, m), rand(disty, n)
4 show_plot_brunner_munzel(X, Y; xlim=(-5, 2))
```

```
(phat = 0.6912499999999999,
sehat = 0.05960002903974512,
tvalue = 3.208891053936604,
df = 75.84080882915966,
p = 0.5,
pvalue = 0.001953574321225764,
α = 0.05,
confint_p = [0.5725422250913458, 0.8099577749086541],
confint_shift = [-2.07997163873439, -0.29464656357396524],
shiftthat = -1.0768936945378669)
median(X) = 1.044779302249417
median(Y) = 2.2527593566191237
```

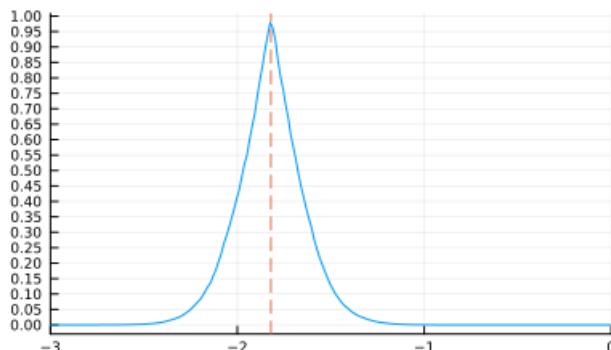
Out[40]: P-value function of shift



```
In [41]: M 1 distx, disty = LogNormal(), LogNormal(1)
2 m, n = 160, 160
3 X, Y = rand(distx, m), rand(disty, n)
4 show_plot_brunner_munzel(X, Y; xlim=(-3, 0))
```

```
(phat = 0.79234375,
sehat = 0.025077139980214335,
tvalue = 11.657778767062629,
df = 315.6587283148897,
p = 0.5,
pvalue = 2.3287816342564484e-26,
α = 0.05,
confint_p = [0.7430042840809774, 0.8416832159190226],
confint_shift = [-2.250313266911184, -1.404197630202607],
shiftthat = -1.8203983580938916)
median(X) = 0.8417596982690712
median(Y) = 2.974128627007357
```

Out[41]: P-value function of shift



## 4 Brunner-Munzel検定とWelchのt検定の比較

### 4.1 第一種の過誤の確率

In [42]:

```

1  function sim_brunner_mumzel();
2      distx = Normal(0, 1), disty = Normal(0, 2), m = 10, n = 10,
3      L = 10^6)
4      pval_bm = Vector{Float64}(undef, L)
5      tmpX = [Vector{Float64}(undef, m) for _ in 1:nthreads()]
6      tmpY = [Vector{Float64}(undef, n) for _ in 1:nthreads()]
7      tmpHx = [Vector{Float64}(undef, m) for _ in 1:nthreads()]
8      tmpHy = [Vector{Float64}(undef, n) for _ in 1:nthreads()]
9      @threads for i in 1:L
10         X = rand!(distx, tmpX[threadid()])
11         Y = rand!(disty, tmpY[threadid()])
12         pval_bm[i] = pvalue_brunner_munzel(X, Y, tmpHx[threadid()], tmpHy[threadid()])
13     end
14     ecdf(pval_bm)
15 end
16
17 function sim_welch();
18     distx = Normal(0, 1), disty = Normal(0, 2), m = 10, n = 10,
19     L = 10^6)
20     pval_w = Vector{Float64}(undef, L)
21     tmpX = [Vector{Float64}(undef, m) for _ in 1:nthreads()]
22     tmpY = [Vector{Float64}(undef, n) for _ in 1:nthreads()]
23     @threads for i in 1:L
24         X = rand!(distx, tmpX[threadid()])
25         Y = rand!(disty, tmpY[threadid()])
26         pval_w[i] = pvalue_welch(X, Y)
27     end
28     ecdf(pval_w)
29 end
30
31 function printcompact(io, xs...)
32     print(IOContext(io, :compact => true), xs...)
33 end
34
35 function distname(dist)
36     replace(sprint(printcompact, dist), r"\{\[^{}]\}*{}"=>"")
37 end
38
39 function plot_ecdf(ecdf_pval, distx, disty, m, n, a;
40     testname = "", kwargs...)
41     plot(p → ecdf_pval(p), 0, 0.1; label="ecdf of P-values")
42     plot!([0, 0.1], [0, 0.1]; label="", ls=:dot, c=:black)
43     plot!(legend=:topleft)
44     plot!(xtick=0:0.01:0.1, ytick=0:0.01:1)
45     plot!(xguide="nominal significance level  $\alpha$ ",
46           yguide="probability of P-value <  $\alpha$ ")
47     s = (a < 0 ? "-" : "+") * string(round(a; digits=4))
48     title!("$(testname)X: $(distname(distx)), m=$m\n\
49             Y: $(distname(disty))$s, n=$n")
50     plot!(size=(400, 450))
51     plot!(; kwargs...)
52 end
53
54 function plot_pvals();
55     distx = Normal(0, 1), disty = Normal(0, 2), m = 10, n = 10,
56     L = 10^6, a = nothing, Δμ = nothing, kwargs...)
57     @show (mean(distx), std(distx))
58     @show (mean(disty), std(disty))
59
60     if isnothing(a)
61         @show a = tieshift(distx, disty)
62         @show prob_x_le_y(distx, disty + a)
63     else
64         @show a
65         @show median(distx) - median(disty)
66     end
67     if isnothing(Δμ)
68         @show Δμ = mean(distx) - mean(disty)
69         @show mean(distx), mean(disty + Δμ)
70     else
71         @show Δμ
72         @show mean(distx), mean(disty + Δμ)
73     end
74
75     ecdf_bm = @time sim_brunner_mumzel();
76     distx = distx,
77     disty = disty + a,
78     m, n, L, kwargs...
79
80     ecdf_w = @time sim_welch();
81     distx = distx,

```

```

81     disty = disty + Δμ,
82     m, n, L, kwargs...)
83 ymax = max(ecdf_bm(0.1), ecdf_w(0.1))
84 P1 = plot_ecdf(ecdf_bm, distx, disty, m, n, a;
85   testname="Brunner-Munzel test\n",
86   ylim=(-0.002, 1.02*ymax), kwargs...)
87 P2 = plot_ecdf(ecdf_w, distx, disty, m, n, Δμ;
88   testname="Welch t-test\n",
89   ylim=(-0.002, 1.02*ymax), kwargs...)
90 plot(P1, P2; size=(800, 450), topmargin=3.5Plots.mm)
91 end

```

Out[42]: `plot_pvals` (generic function with 1 method)

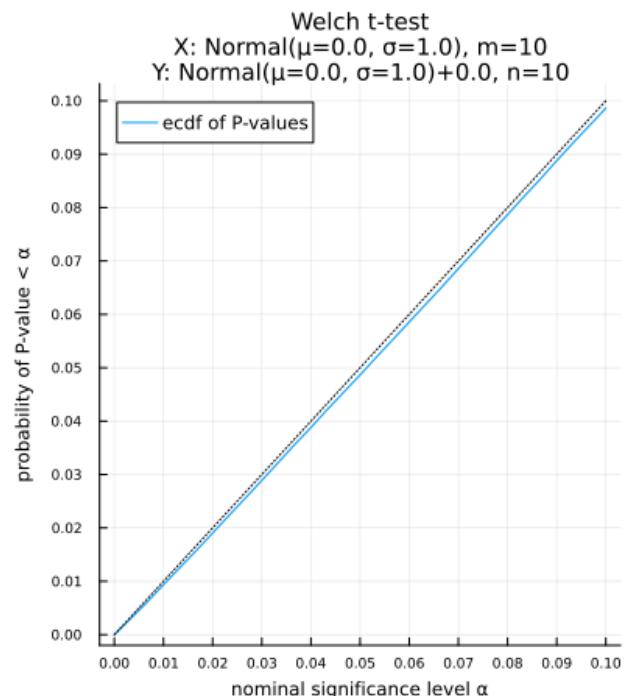
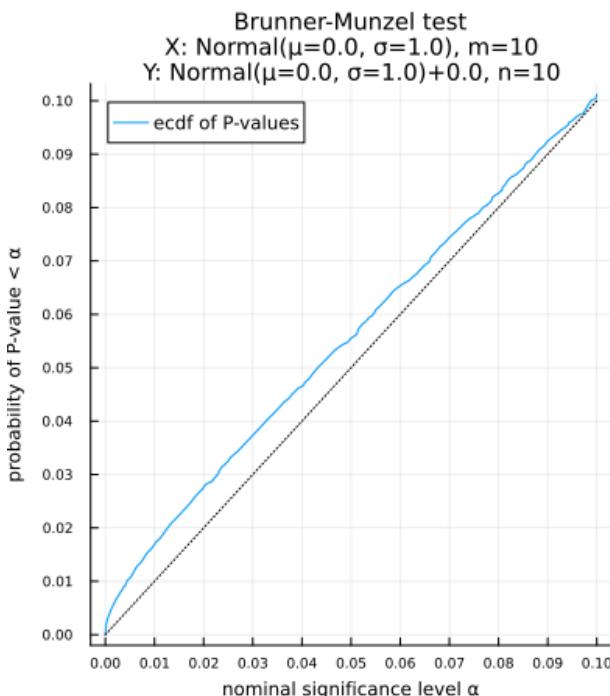
In [43]: `plot_pvals(; distx = Normal(0, 1), disty = Normal(0, 1), m = 10, n = 10)`

```

(mean(distx), std(distx)) = (0.0, 1.0)
(mean(disty), std(disty)) = (0.0, 1.0)
a = tieshift(distx, disty) = 0.0
prob_x_le_y(distx, disty + a) = 0.5
Δμ = mean(distx) - mean(disty) = 0.0
(mean(distx), mean(disty + Δμ)) = (0.0, 0.0)
0.323238 seconds (640.44 k allocations: 196.441 MiB, 3.10% gc time, 212.22% compilation time: 4% of which was recompilation)
0.311052 seconds (629.88 k allocations: 208.505 MiB, 10.85% gc time, 170.78% compilation time)

```

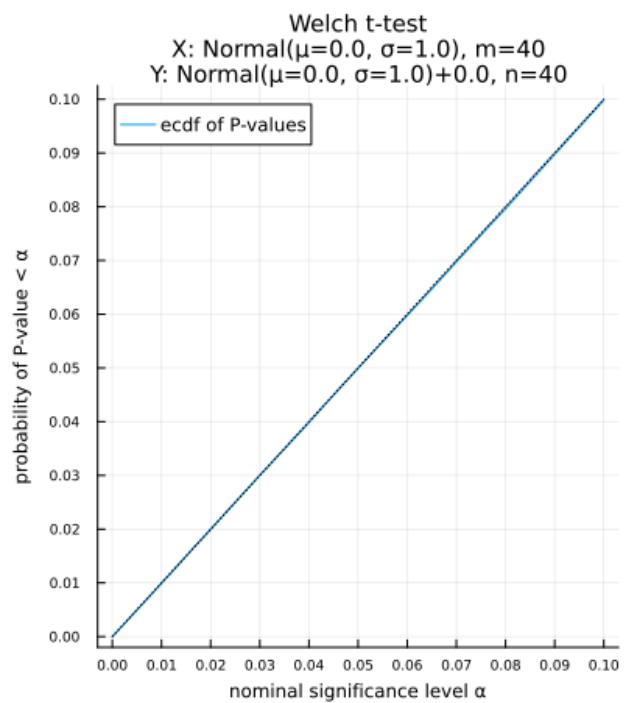
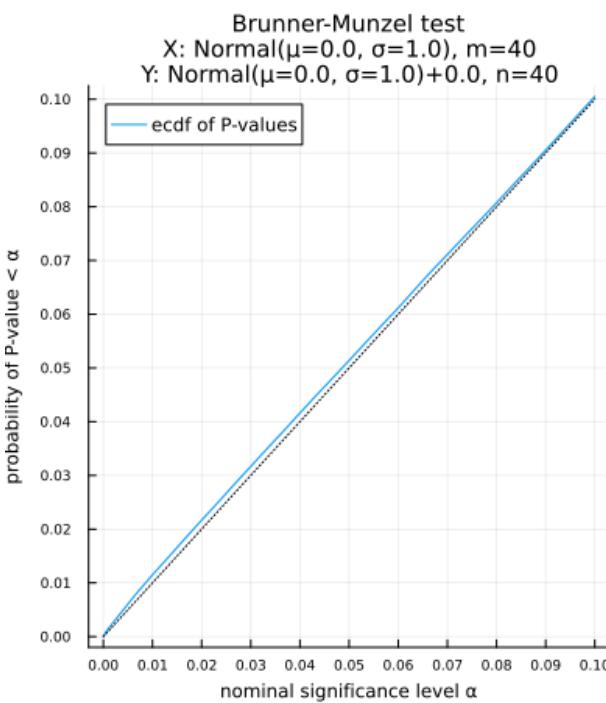
Out[43]:



In [44]: 1 plot\_pvals(; distx = Normal(0, 1), disty = Normal(0, 1), m = 40, n = 40)

```
(mean(distx), std(distx)) = (0.0, 1.0)
(mean(disty), std(disty)) = (0.0, 1.0)
a = tieshift(distx, disty) = 0.0
prob_x_le_y(distx, disty + a) = 0.5
Δμ = mean(distx) - mean(disty) = 0.0
(mean(distx), mean(disty + Δμ)) = (0.0, 0.0)
0.697588 seconds (636.69 k allocations: 215.096 MiB)
0.290773 seconds (645.87 k allocations: 217.731 MiB, 13.63% gc time)
```

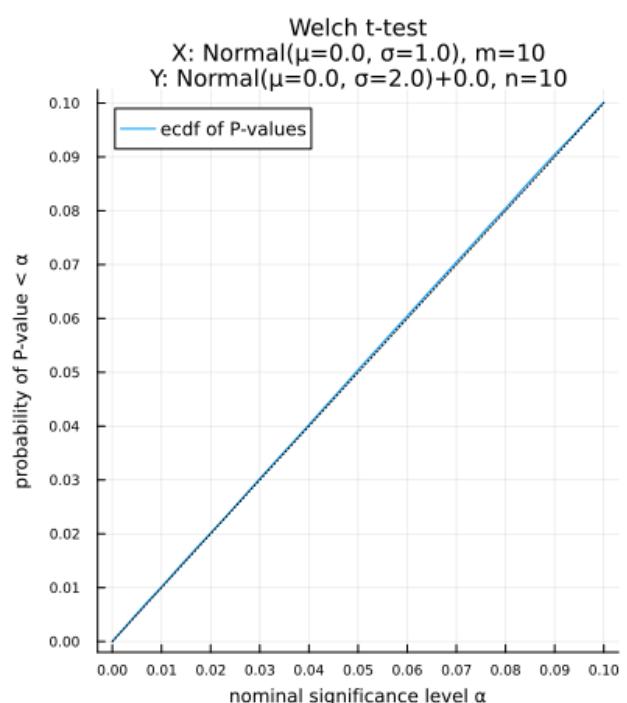
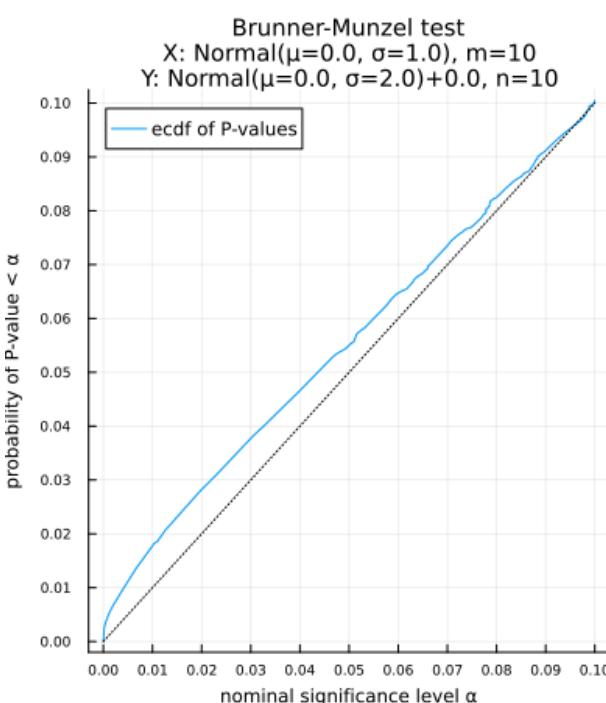
Out[44]:



In [45]: 1 plot\_pvals(; distx = Normal(0, 1), disty = Normal(0, 2), m = 10, n = 10)

```
(mean(distx), std(distx)) = (0.0, 1.0)
(mean(disty), std(disty)) = (0.0, 2.0)
a = tieshift(distx, disty) = 7.685641860444171e-14
prob_x_le_y(distx, disty + a) = 0.5
Δμ = mean(distx) - mean(disty) = 0.0
(mean(distx), mean(disty + Δμ)) = (0.0, 0.0)
0.324353 seconds (524.30 k allocations: 182.475 MiB, 9.79% gc time)
0.294450 seconds (585.29 k allocations: 200.184 MiB, 20.81% gc time)
```

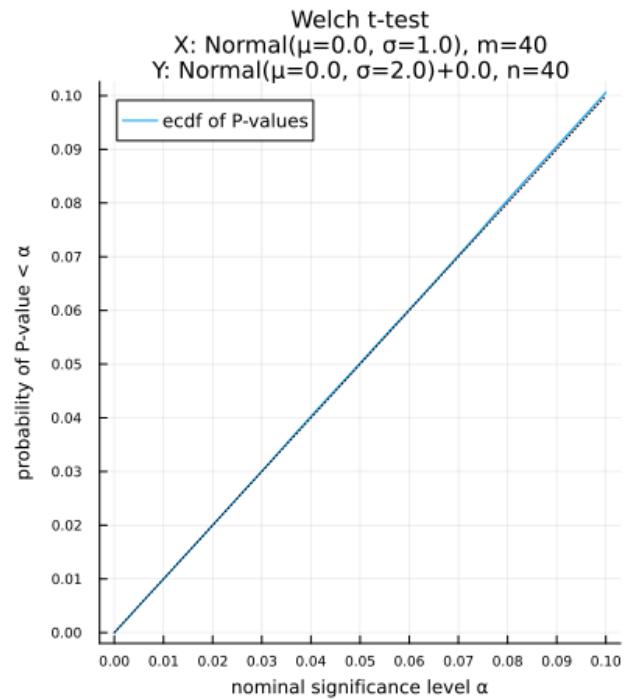
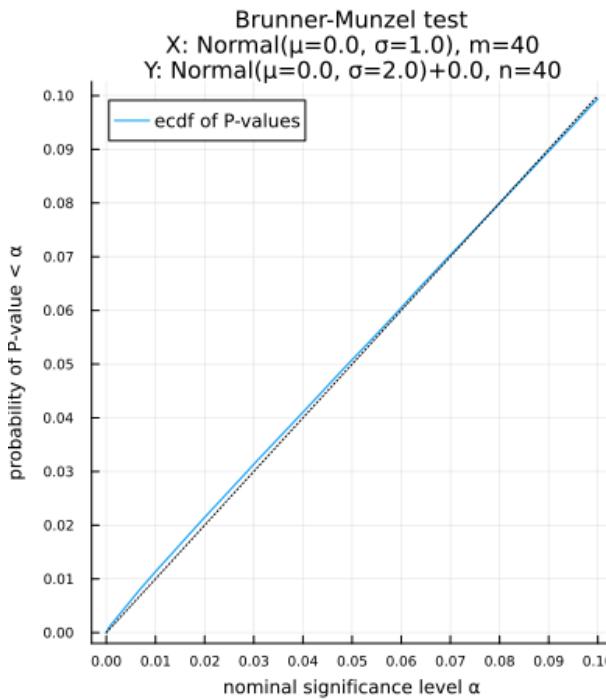
Out[45]:



In [46]: 1 plot\_pvals(; distx = Normal(0, 1), disty = Normal(0, 2), m = 40, n = 40)

```
(mean(distx), std(distx)) = (0.0, 1.0)
(mean(disty), std(disty)) = (0.0, 2.0)
a = tieshift(distx, disty) = 7.685641860444171e-14
prob_x_le_y(distx, disty + a) = 0.5
Δμ = mean(distx) - mean(disty) = 0.0
(mean(distx), mean(disty + Δμ)) = (0.0, 0.0)
0.782463 seconds (633.98 k allocations: 214.285 MiB, 4.49% gc time)
0.254228 seconds (647.02 k allocations: 218.062 MiB)
```

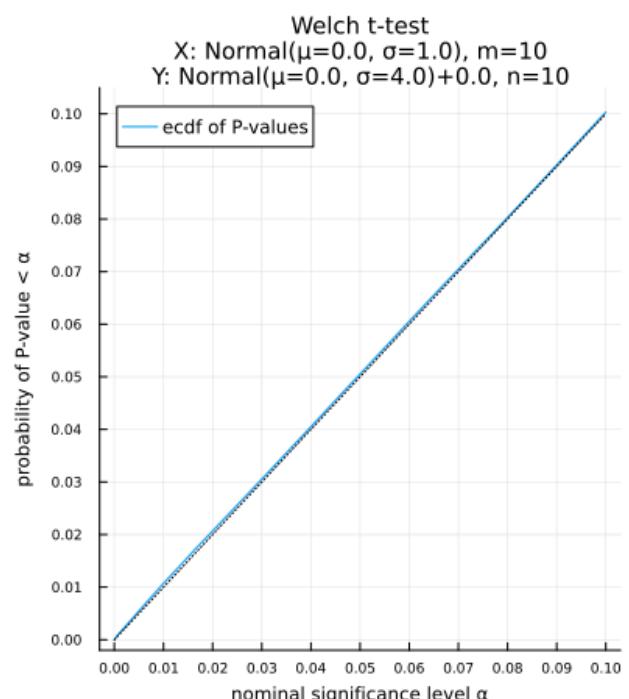
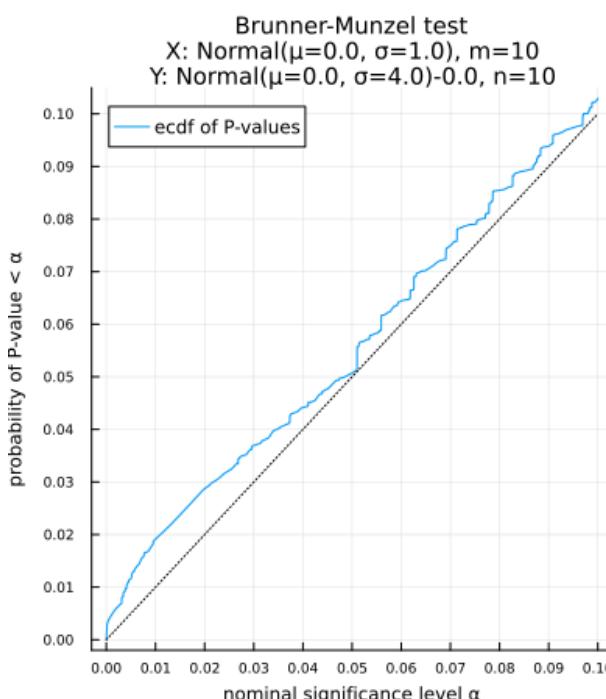
Out[46]:



In [47]: 1 plot\_pvals(; distx = Normal(0, 1), disty = Normal(0, 4), m = 10, n = 10)

```
(mean(distx), std(distx)) = (0.0, 1.0)
(mean(disty), std(disty)) = (0.0, 4.0)
a = tieshift(distx, disty) = -2.715980146908965e-13
prob_x_le_y(distx, disty + a) = 0.5
Δμ = mean(distx) - mean(disty) = 0.0
(mean(distx), mean(disty + Δμ)) = (0.0, 0.0)
0.323551 seconds (486.84 k allocations: 171.615 MiB, 11.33% gc time)
0.263402 seconds (534.49 k allocations: 185.457 MiB, 15.19% gc time)
```

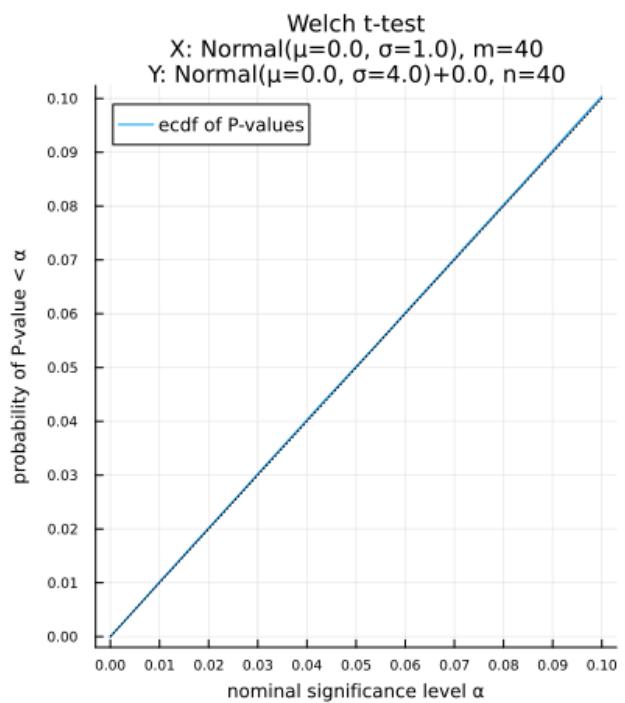
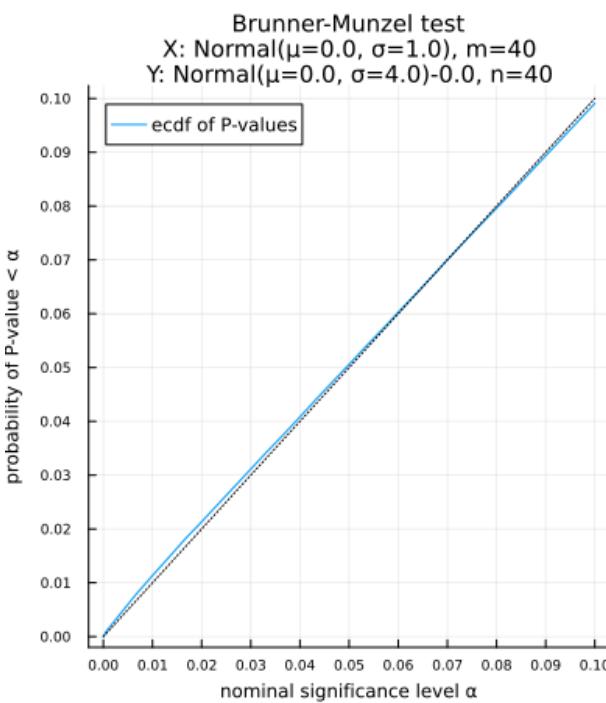
Out[47]:



In [48]: 1 plot\_pvals(; distx = Normal(0, 1), disty = Normal(0, 4), m = 40, n = 40)

```
(mean(distx), std(distx)) = (0.0, 1.0)
(mean(disty), std(disty)) = (0.0, 4.0)
a = tieshift(distx, disty) = -2.715980146908965e-13
prob_x_le_y(distx, disty + a) = 0.5
Δμ = mean(distx) - mean(disty) = 0.0
(mean(distx), mean(disty + Δμ)) = (0.0, 0.0)
0.783298 seconds (633.66 k allocations: 214.193 MiB, 4.72% gc time)
0.244073 seconds (645.09 k allocations: 217.503 MiB)
```

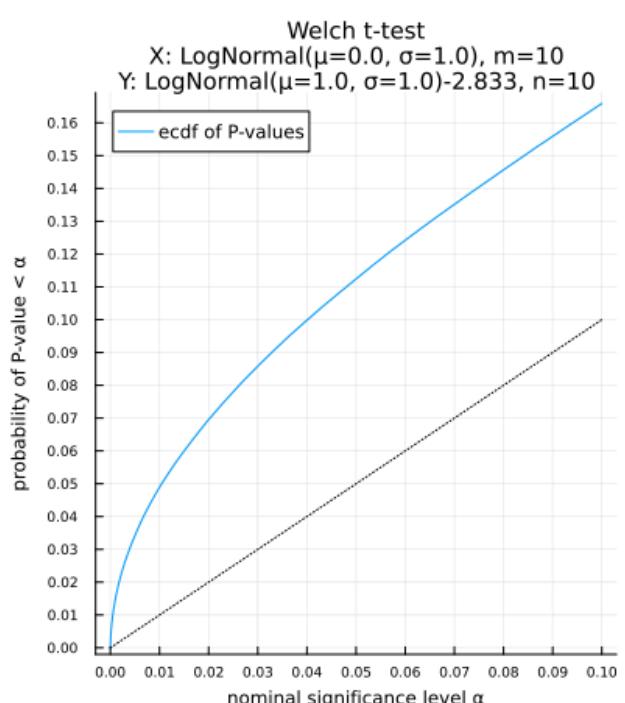
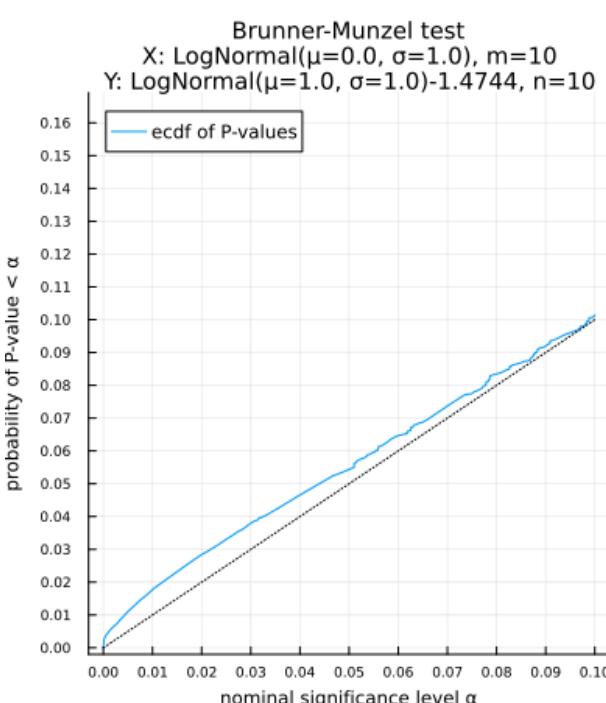
Out[48]:



In [49]: 1 plot\_pvals(; distx = LogNormal(), disty = LogNormal(1), m = 10, n = 10)

```
(mean(distx), std(distx)) = (1.6487212707001282, 2.1611974158950877)
(mean(disty), std(disty)) = (4.4816890703380645, 5.874743663340262)
a = tieshift(distx, disty) = -1.4744426128871542
prob_x_le_y(distx, disty + a) = 0.5
Δμ = mean(distx) - mean(disty) = -2.8329677996379363
(mean(distx), mean(disty + Δμ)) = (1.6487212707001282, 1.6487212707001282)
0.333222 seconds (528.03 k allocations: 179.550 MiB, 8.37% gc time, 120.31% compilation time)
0.259252 seconds (623.33 k allocations: 208.906 MiB, 124.62% compilation time)
```

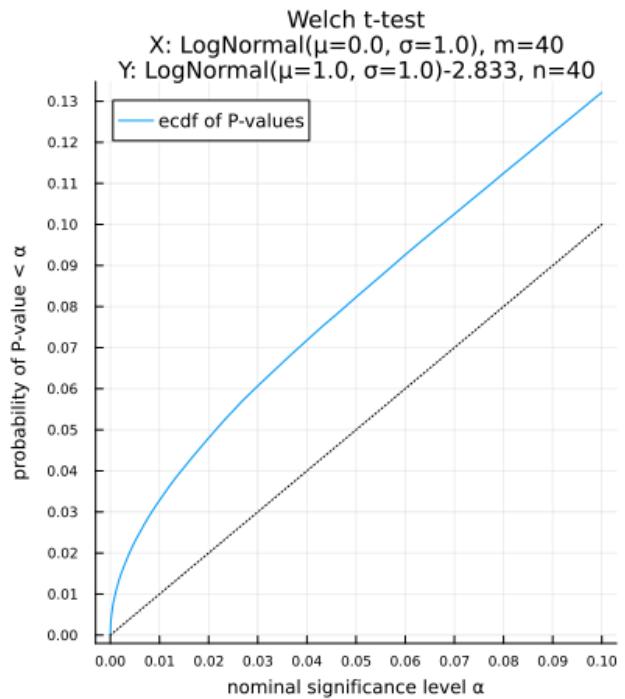
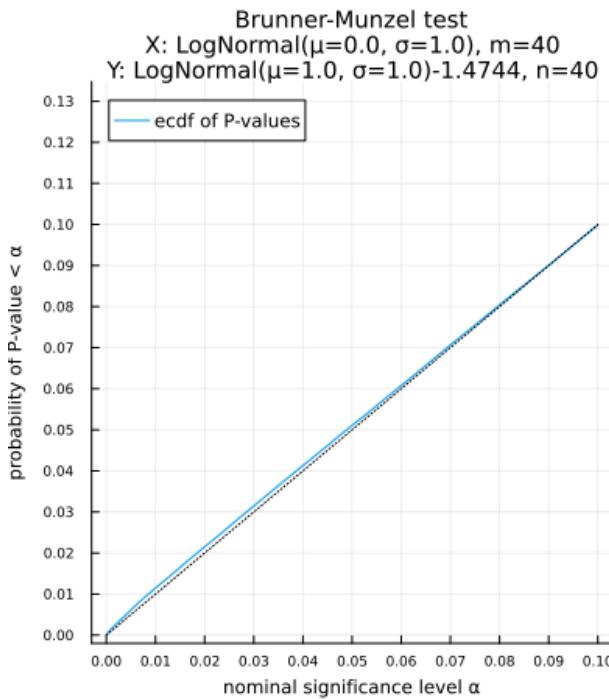
Out[49]:



In [50]: 1 plot\_pvals(; distx = LogNormal(), disty = LogNormal(1), m = 40, n = 40)

```
(mean(distx), std(distx)) = (1.6487212707001282, 2.1611974158950877)
(mean(disty), std(disty)) = (4.4816890703380645, 5.874743663340262)
a = tieshift(distx, disty) = -1.4744426128871542
prob_x_le_y(distx, disty + a) = 0.5
Δμ = mean(distx) - mean(disty) = -2.8329677996379363
(mean(distx), mean(disty + Δμ)) = (1.6487212707001282, 1.6487212707001282)
0.883382 seconds (636.07 k allocations: 214.890 MiB, 4.82% gc time)
0.419585 seconds (709.10 k allocations: 236.063 MiB, 15.21% gc time)
```

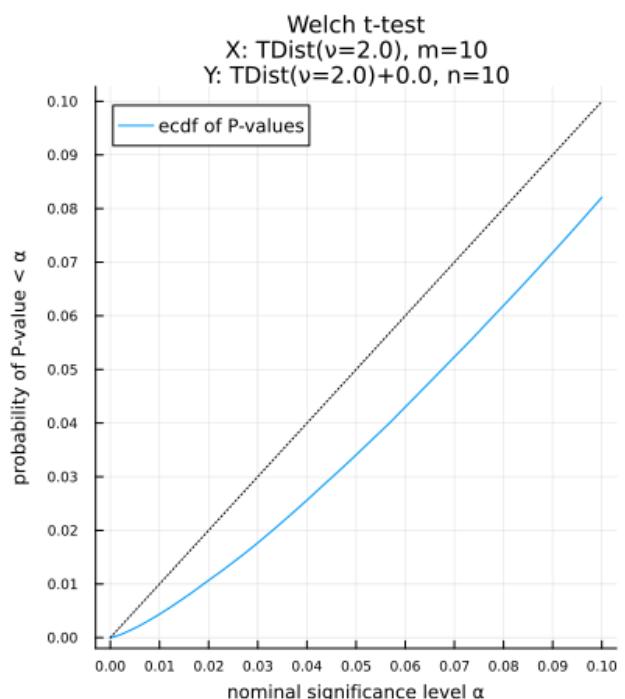
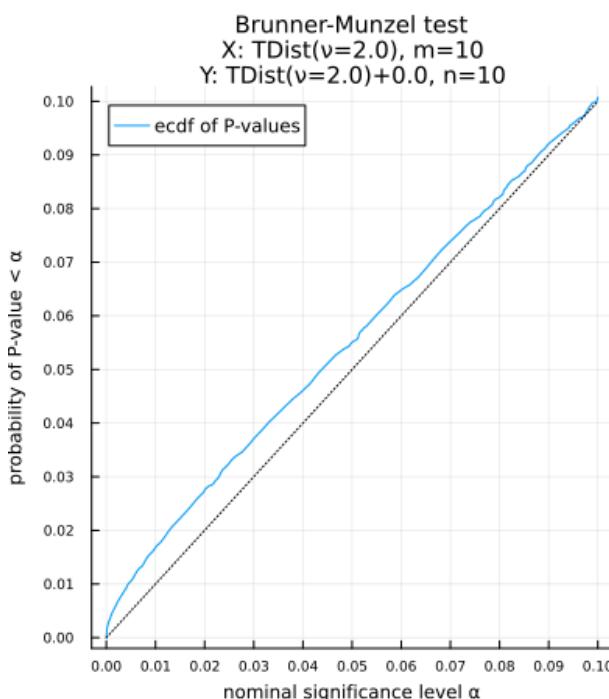
Out[50]:



In [51]: 1 plot\_pvals(; distx = TDist(2), disty = TDist(2), m = 10, n = 10, Δμ = 0.0)

```
(mean(distx), std(distx)) = (0.0, Inf)
(mean(disty), std(disty)) = (0.0, Inf)
a = tieshift(distx, disty) = 0.0
prob_x_le_y(distx, disty + a) = 0.5000000000000001
Δμ = 0.0
(mean(distx), mean(disty + Δμ)) = (0.0, 0.0)
0.434859 seconds (688.11 k allocations: 199.328 MiB, 5.49% gc time, 369.47% compilation time)
0.343614 seconds (674.43 k allocations: 223.719 MiB, 17.65% gc time, 98.98% compilation time)
```

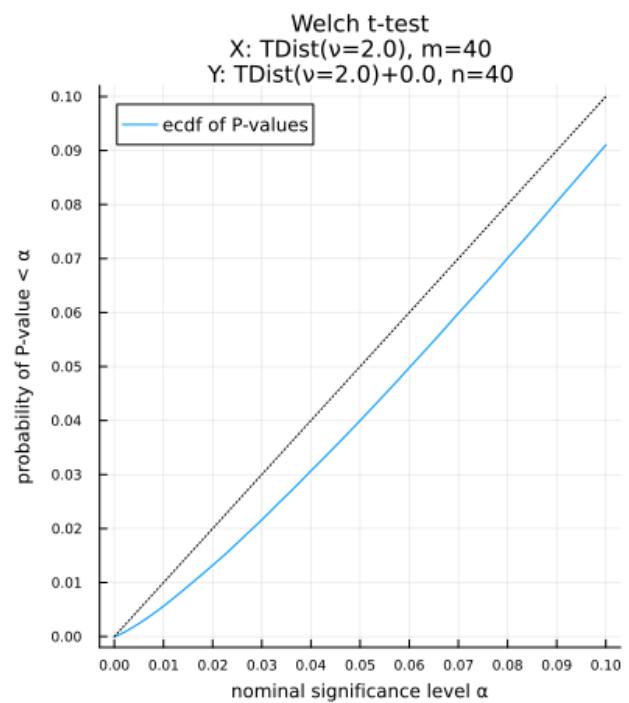
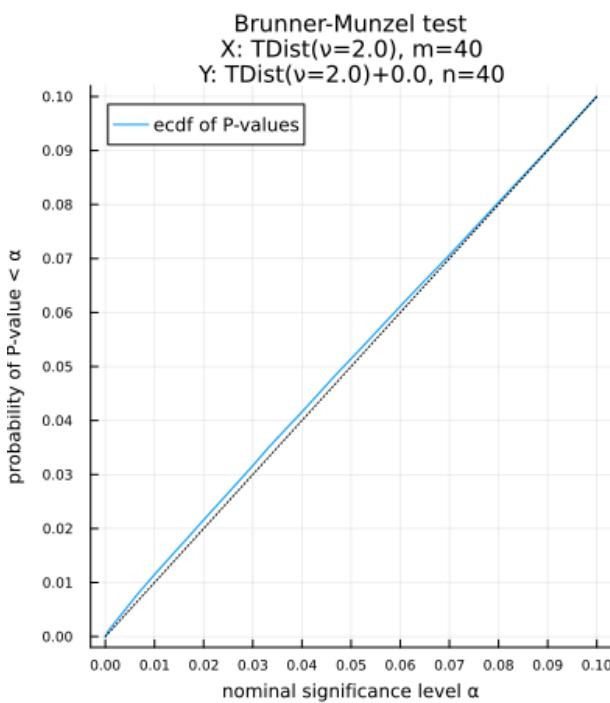
Out[51]:



In [52]: 1 plot\_pvals(; distx = TDist(2), disty = TDist(2), m = 40, n = 40, Δμ = 0.0)

```
(mean(distx), std(distx)) = (0.0, Inf)
(mean(disty), std(disty)) = (0.0, Inf)
a = tieshift(distx, disty) = 0.0
prob_x_le_y(distx, disty + a) = 0.5000000000000001
Δμ = 0.0
(mean(distx), mean(disty + Δμ)) = (0.0, 0.0)
0.896450 seconds (636.62 k allocations: 215.052 MiB, 3.57% gc time)
0.404594 seconds (690.55 k allocations: 230.683 MiB)
```

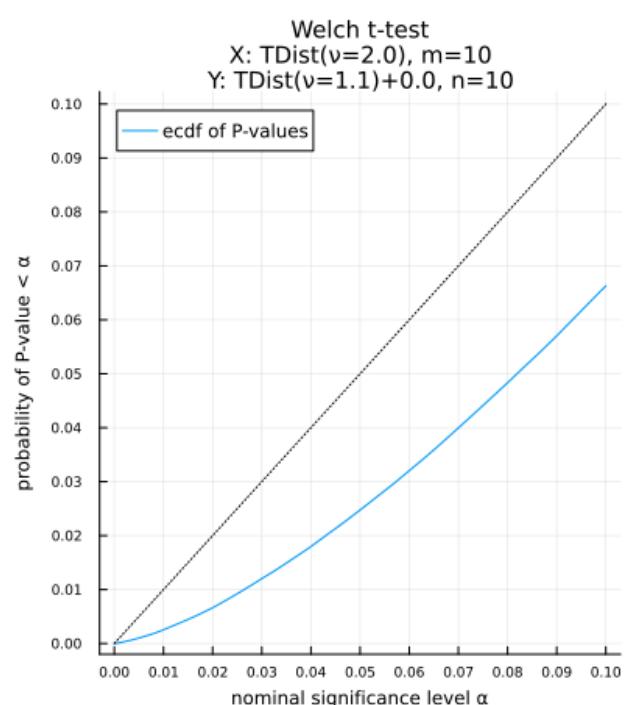
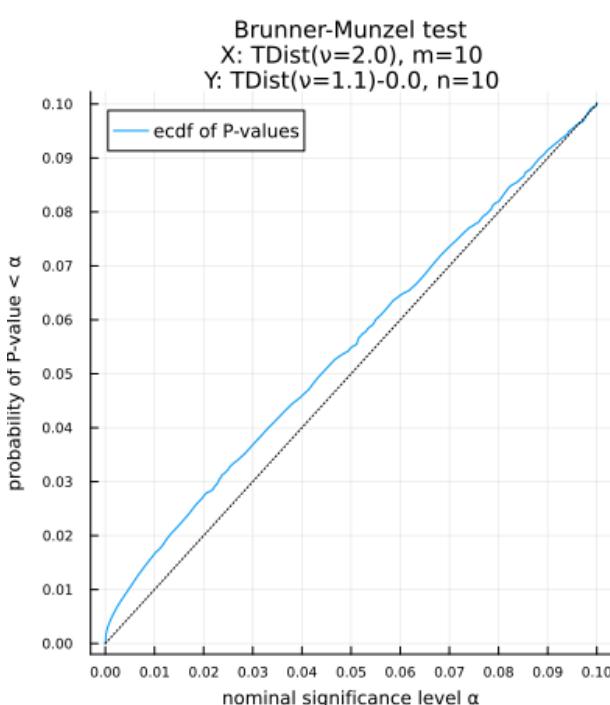
Out[52]:



In [53]: 1 plot\_pvals(; distx = TDist(2), disty = TDist(1.1), m = 10, n = 10, Δμ = 0.0)

```
(mean(distx), std(distx)) = (0.0, Inf)
(mean(disty), std(disty)) = (0.0, Inf)
a = tieshift(distx, disty) = -3.4064502213362996e-9
prob_x_le_y(distx, disty + a) = 0.5
Δμ = 0.0
(mean(distx), mean(disty + Δμ)) = (0.0, 0.0)
0.425950 seconds (547.89 k allocations: 189.315 MiB, 14.85% gc time)
0.381495 seconds (714.35 k allocations: 237.577 MiB, 11.94% gc time)
```

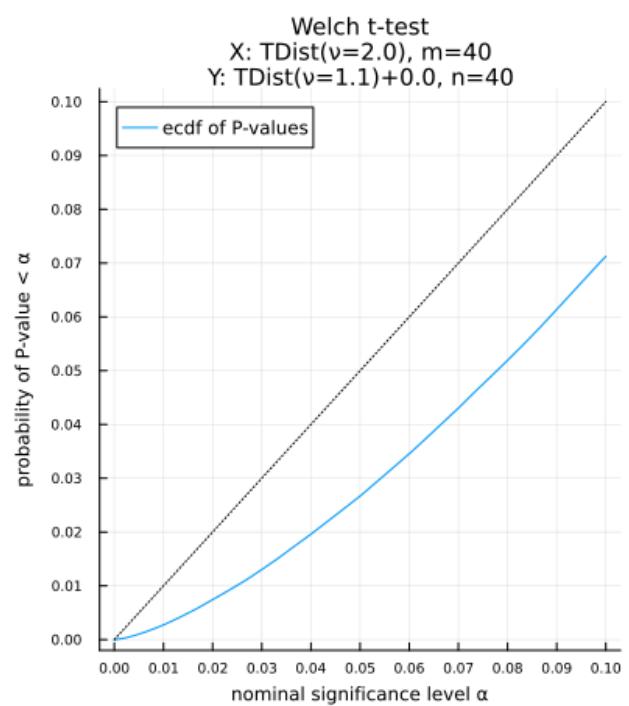
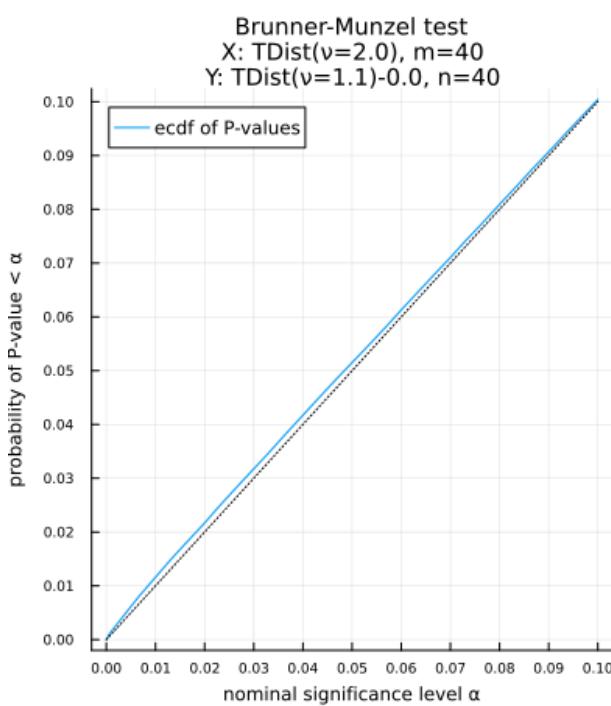
Out[53]:



In [54]: 1 plot\_pvals(; distx = TDist(2), disty = TDist(1.1), m = 40, n = 40, Δμ = 0.0)

```
(mean(distx), std(distx)) = (0.0, Inf)
(mean(disty), std(disty)) = (0.0, Inf)
a = tieshift(distx, disty) = -3.4064502213362996e-9
prob_x_le_y(distx, disty + a) = 0.5
Δμ = 0.0
(mean(distx), mean(disty + Δμ)) = (0.0, 0.0)
1.052793 seconds (636.12 k allocations: 214.907 MiB)
0.625337 seconds (757.21 k allocations: 250.009 MiB, 4.88% gc time)
```

Out[54]:

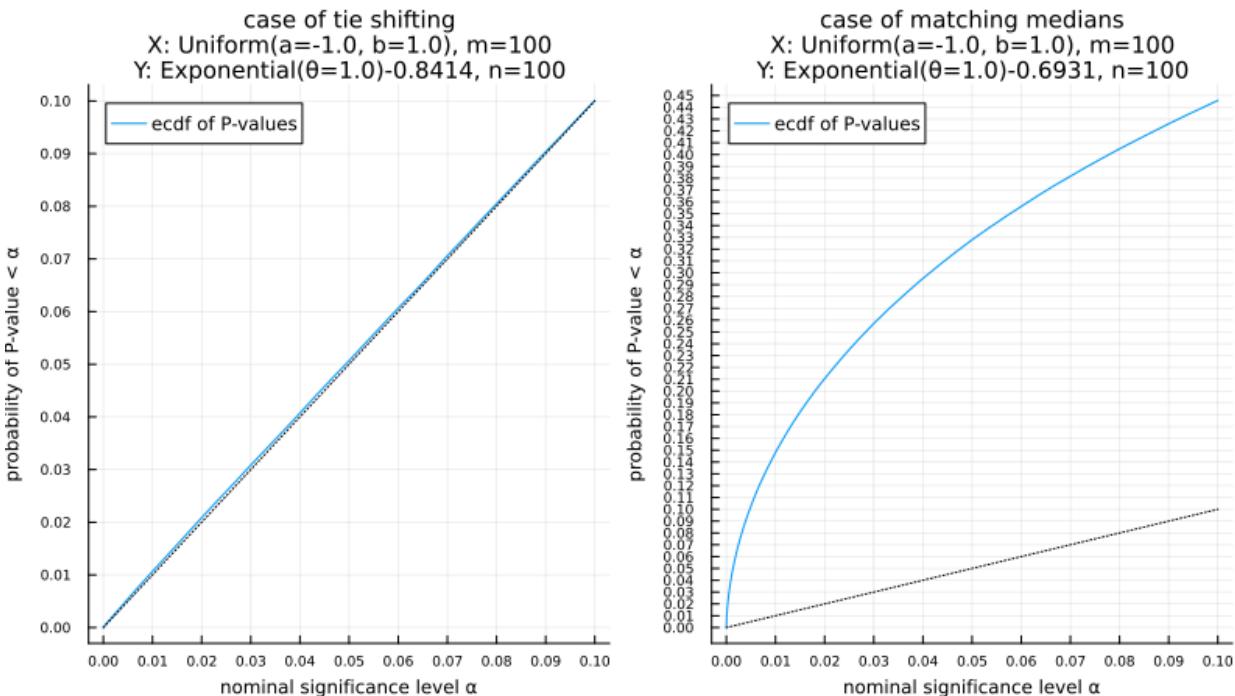


## 4.2 Brunner-Munzel検定は中央値に関する検定ではないことの証拠

```
In [55]: M
1 distx, disty = Uniform(-1, 1), Exponential()
2 m, n, = 100, 100
3
4 @show distx, std(distx)
5 @show disty, std(disty)
6
7 @show a = tieshift(distx, disty)
8 ecdf_pval1 = @time sim_brunner_mumzel();
9     distx = distx, disty = disty + a, m, n)
10 P1 = plot_ecdf(ecdf_pval1, distx, disty, m, n, a;
11     testname="case of tie shifting\n")
12
13 @show a = median(distx) - median(disty)
14 ecdf_pval2 = @time sim_brunner_mumzel();
15     distx = distx, disty = disty + a, m, n)
16 P2 = plot_ecdf(ecdf_pval2, distx, disty, m, n, a;
17     testname="case of matching medians\n")
18
19 plot(P1, P2; size=(800, 450), topmargin=4Plots.mm)
```

(distx, std(distx)) = (Uniform{Float64}(a=-1.0, b=1.0), 0.5773502691896257)  
 (disty, std(disty)) = (Exponential{Float64}( $\theta=1.0$ ), 1.0)  
 a = tieshift(distx, disty) = -0.8414056600399943  
 3.157133 seconds (798.40 k allocations: 227.154 MiB, 2.15% gc time, 28.78% compilation time)  
 a = median(distx) - median(disty) = -0.6931471805599453  
 2.948860 seconds (1.40 M allocations: 437.444 MiB, 1.55% gc time)

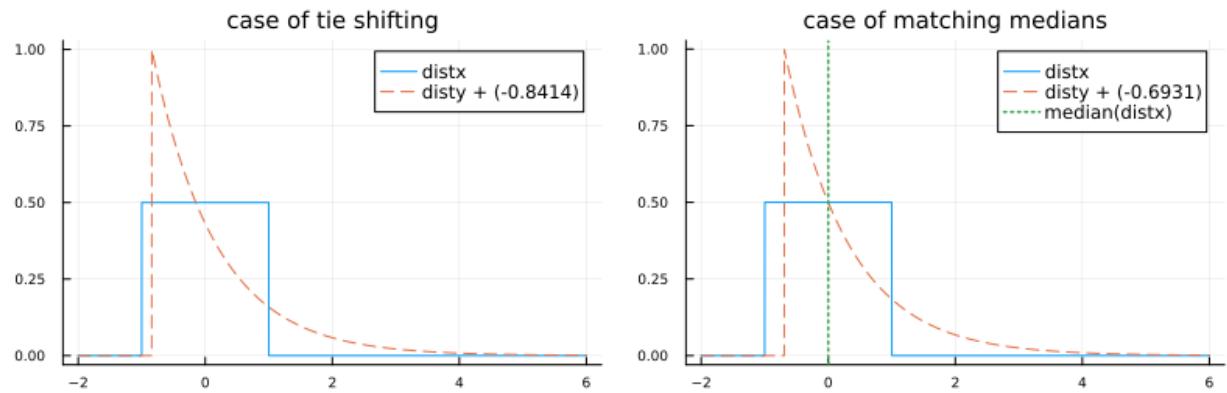
Out[55]:



```
In [56]: 1 distx, disty = Uniform(-1, 1), Exponential()
2 @show distx, std(distx)
3 @show disty, std(disty)
4
5 a = @show tieshift(distx, disty)
6 P1 = plot(distx, -2, 6; label="distx")
7 plot!(disty + a, -2, 6; label="disty + $(round(a; digits=4))", ls=:dash)
8 title!("case of tie shifting")
9
10 a = @show median(distx) - median(disty)
11 P2 = plot(distx, -2, 6; label="distx")
12 plot!(disty + a, -2, 6; label="disty + $(round(a; digits=4))", ls=:dash)
13 vline!([median(distx)]; label="median(distx)", ls=:dot, lw=1.5)
14 title!("case of matching medians")
15
16 plot(P1, P2; size=(800, 250))
```

(distx, std(distx)) = (Uniform{Float64}(a=-1.0, b=1.0), 0.5773502691896257)  
 (disty, std(disty)) = (Exponential{Float64}( $\theta=1.0$ ), 1.0)  
 tieshift(distx, disty) = -0.8414056600399943  
 median(distx) - median(disty) = -0.6931471805599453

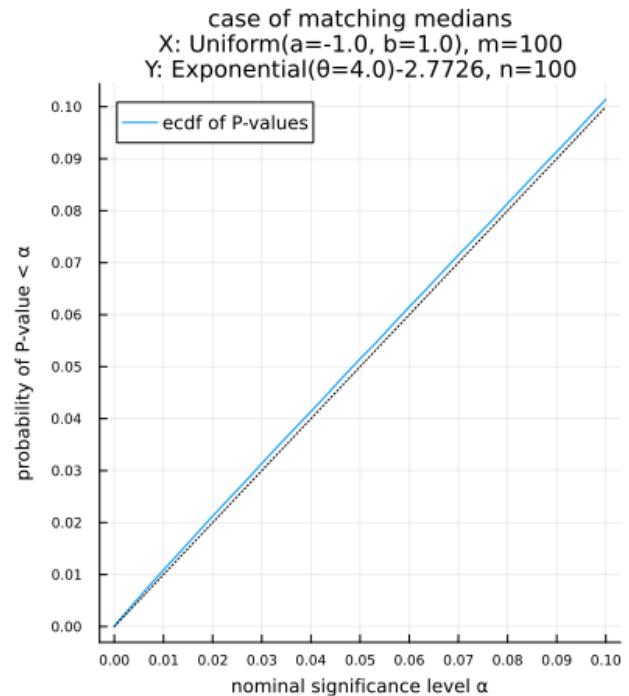
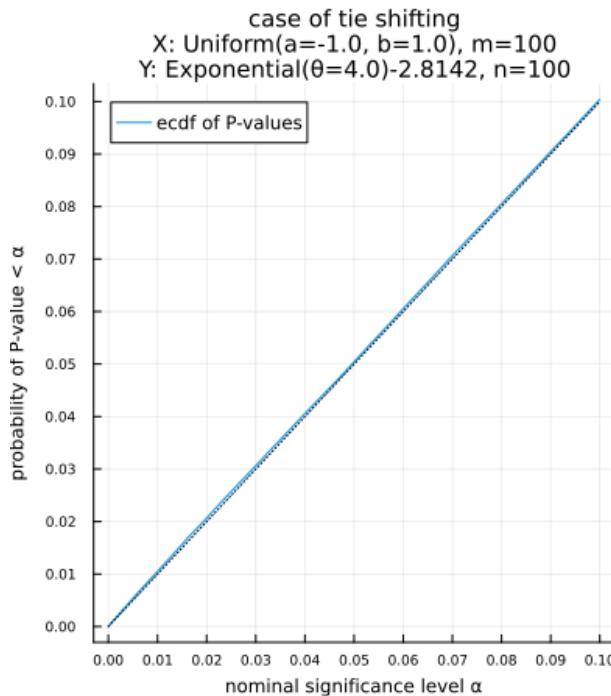
Out[56]:



```
In [57]: 1 distx, disty = Uniform(-1, 1), Exponential(4)
2 m, n, = 100, 100
3
4 @show distx, std(distx)
5 @show disty, std(disty)
6
7 @show a = tieshift(distx, disty)
8 ecdf_pval1 = @time sim_brunner_mumzel();
9 distx = distx, disty = disty + a, m, n)
10 P1 = plot_ecdf(ecdf_pval1, distx, disty, m, n, a;
11 testname="case of tie shifting\n")
12
13 @show a = median(distx) - median(disty)
14 ecdf_pval2 = @time sim_brunner_mumzel();
15 distx = distx, disty = disty + a, m, n)
16 P2 = plot_ecdf(ecdf_pval2, distx, disty, m, n, a;
17 testname="case of matching medians\n")
18
19 plot(P1, P2; size=(800, 450), topmargin=4Plots.mm)
```

```
(distx, std(distx)) = (Uniform{Float64}(a=-1.0, b=1.0), 0.5773502691896257)
(disty, std(disty)) = (Exponential{Float64}(\theta=4.0), 4.0)
a = tieshift(distx, disty) = -2.814168911097314
2.906576 seconds (639.61 k allocations: 215.939 MiB)
a = median(distx) - median(disty) = -2.772588722239781
2.929456 seconds (645.53 k allocations: 217.653 MiB, 1.19% gc time)
```

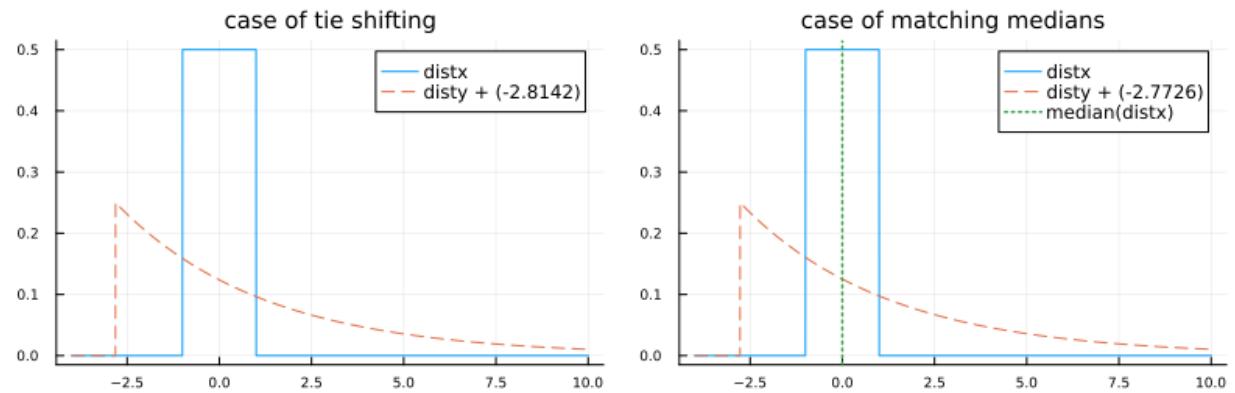
Out[57]:



```
In [58]: 1 distx, disty = Uniform(-1, 1), Exponential(4)
2 @show distx, std(distx)
3 @show disty, std(disty)
4
5 a = @show tieshift(distx, disty)
6 P1 = plot(distx, -4, 10; label="distx")
7 plot!(disty + a, -4, 10; label="disty + $(round(a; digits=4))", ls=:dash)
8 title!("case of tie shifting")
9
10 a = @show median(distx) - median(disty)
11 P2 = plot(distx, -4, 10; label="distx")
12 plot!(disty + a, -4, 10; label="disty + $(round(a; digits=4))", ls=:dash)
13 vline!([median(distx)]; label="median(distx)", ls=:dot, lw=1.5)
14 title!("case of matching medians")
15
16 plot(P1, P2; size=(800, 250))
```

(distx, std(distx)) = (Uniform{Float64}(a=-1.0, b=1.0), 0.5773502691896257)  
 (disty, std(disty)) = (Exponential{Float64}( $\theta=4.0$ ), 4.0)  
 tieshift(distx, disty) = -2.814168911097314  
 median(distx) - median(disty) = -2.772588722239781

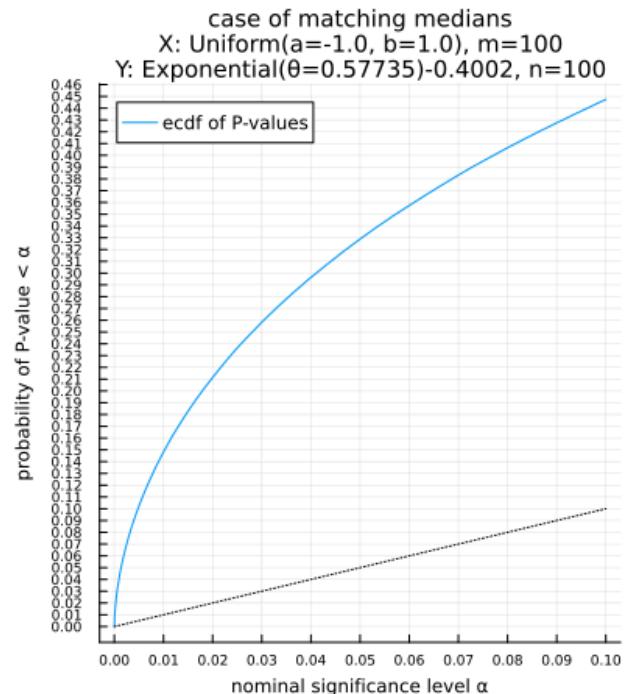
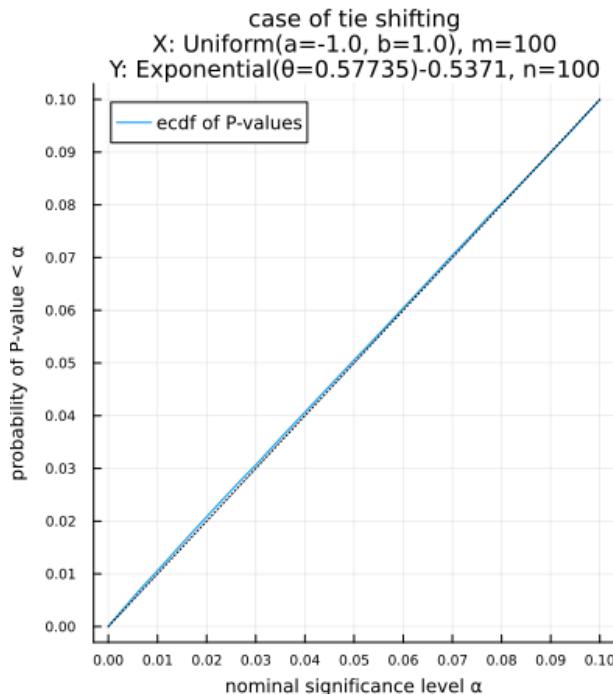
Out[58]:



```
In [59]: 1 distx, disty = Uniform(-1, 1), Exponential(0.5773502691896257)
2 m, n, = 100, 100
3
4 @show distx, std(distx)
5 @show disty, std(disty)
6
7 @show a = tieshift(distx, disty)
8 ecdf_pval1 = @time sim_brunner_mumzel();
9 distx = distx, disty = disty + a, m, n)
10 P1 = plot_ecdf(ecdf_pval1, distx, disty, m, n, a;
11 testname="case of tie shifting\n")
12
13 @show a = median(distx) - median(disty)
14 ecdf_pval2 = @time sim_brunner_mumzel();
15 distx = distx, disty = disty + a, m, n)
16 P2 = plot_ecdf(ecdf_pval2, distx, disty, m, n, a;
17 testname="case of matching medians\n")
18
19 plot(P1, P2; size=(800, 450), topmargin=4Plots.mm)
```

```
(distx, std(distx)) = (Uniform{Float64}(a=-1.0, b=1.0), 0.5773502691896257)
(disty, std(disty)) = (Exponential{Float64}(\theta=0.5773502691896257), 0.5773502691896257)
a = tieshift(distx, disty) = -0.5370568188698567
3.047695 seconds (641.75 k allocations: 216.560 MiB, 1.90% gc time)
a = median(distx) - median(disty) = -0.40018871128431455
3.046602 seconds (1.40 M allocations: 437.657 MiB, 2.13% gc time)
```

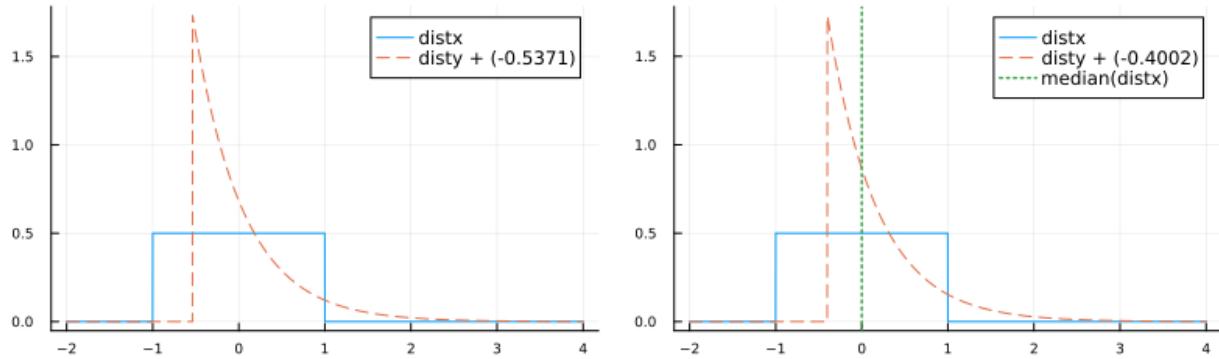
Out[59]:



```
In [60]: 1 distx, disty = Uniform(-1, 1), Exponential(0.5773502691896257)
2 @show distx, std(distx)
3 @show disty, std(disty)
4
5 a = @show tieshift(distx, disty)
6 P1 = plot(distx, -2, 4; label="distx")
7 plot!(disty + a, -2, 4; label="disty + $(round(a; digits=4))", ls=:dash)
8 title!("case of tie shifting")
9
10 a = @show median(distx) - median(disty)
11 P2 = plot(distx, -2, 4; label="distx")
12 plot!(disty + a, -2, 4; label="disty + $(round(a; digits=4))", ls=:dash)
13 vline!([median(distx)]; label="median(distx)", ls=:dot, lw=1.5)
14 title!("case of matching medians")
15
16 plot(P1, P2; size=(800, 250))

(distx, std(distx)) = (Uniform{Float64}(a=-1.0, b=1.0), 0.5773502691896257)
(disty, std(disty)) = (Exponential{Float64}(\theta=0.5773502691896257), 0.5773502691896257)
tieshift(distx, disty) = -0.5370568188698567
median(distx) - median(disty) = -0.40018871128431455
```

Out[60]:



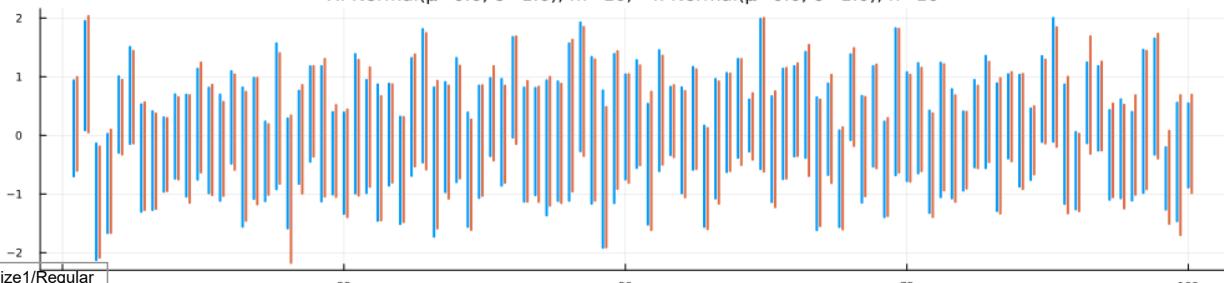
#### 4.3 BM検定による互角シフトの信頼区間とWelchのt検定による平均の差の信頼区間の比較

```
In [61]: 1 function plot_confints();
2     distx = Normal(0, 1), disty = Normal(0, 2), m = 10, n = 10,
3     L = 100, kwargs...
4     a = tieshift(distx, disty)
5     Δμ = mean(distx) - mean(disty)
6     BM = fill(zeros(2), 0)
7     W = fill(zeros(2), 0)
8     for _ in 1:L
9         X = rand(distx, m)
10        Y = rand(disty, n)
11        push!(BM, brunner_munzel(X, Y .+ a).confint_shift)
12        push!(W, confint_welch(X, Y .+ Δμ))
13    end
14    P = plot()
15    for i in 1:L
16        plot!(fill(i, 2), [first(BM[i]), last(BM[i])]; label="", c=1, lw=2)
17        plot!(fill(i+0.3, 2), [first(W[i]), last(W[i])]; label="", c=2, lw=2)
18    end
19    title!("X: $(distname(distx)), m=$m, Y: $(distname(disty)), n=$n")
20    plot!(size=(1000, 250))
21 end
```

Out[61]: plot\_confints (generic function with 1 method)

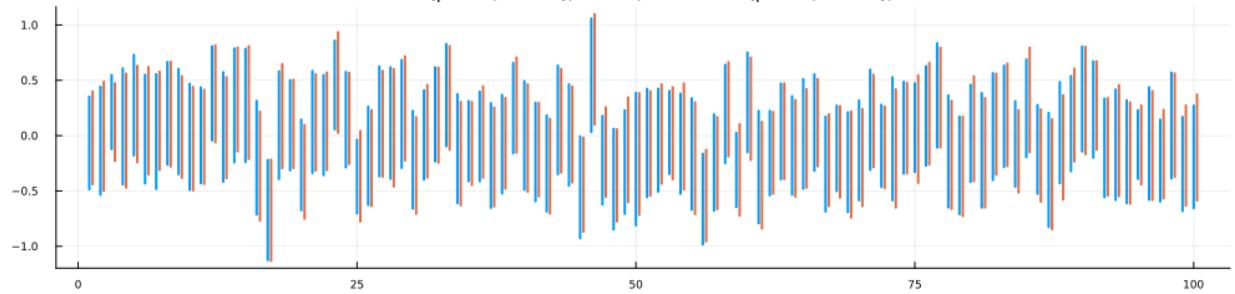
In [62]: 1 plot\_confints(distx = Normal(0, 1), disty = Normal(0, 1), m = 10, n = 10)

Out[62]:



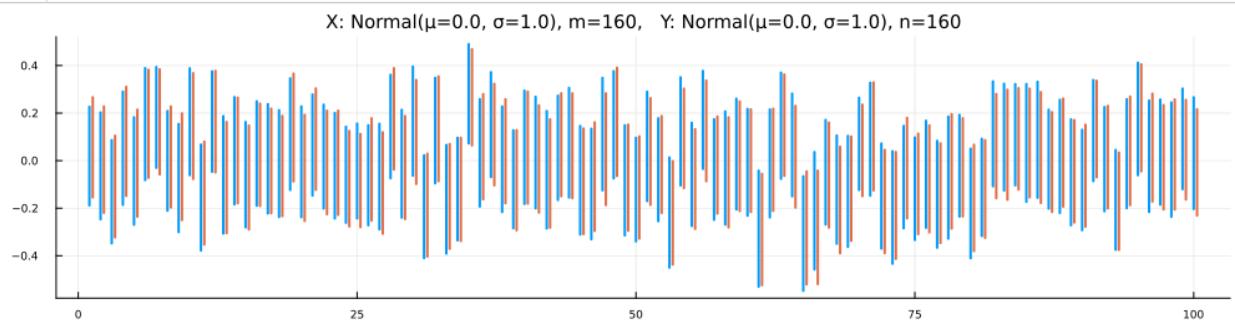
```
In [63]: 1 plot_confints(distx = Normal(0, 1), disty = Normal(0, 1), m = 40, n = 40)
```

Out[63]:



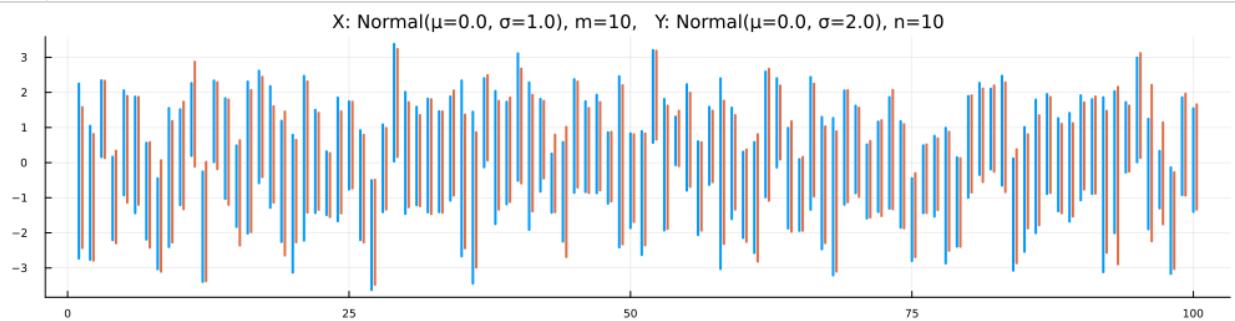
```
In [64]: 1 plot_confints(distx = Normal(0, 1), disty = Normal(0, 1), m = 160, n = 160)
```

Out[64]:



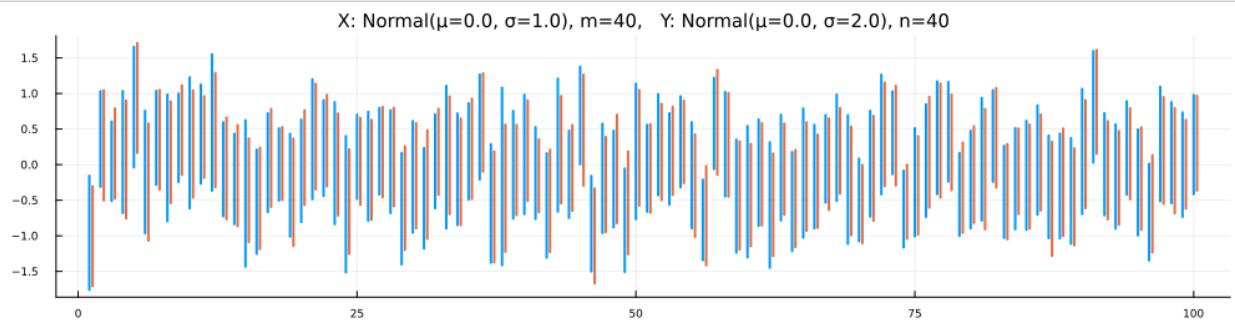
```
In [65]: 1 plot_confints(distx = Normal(0, 1), disty = Normal(0, 2), m = 10, n = 10)
```

Out[65]:



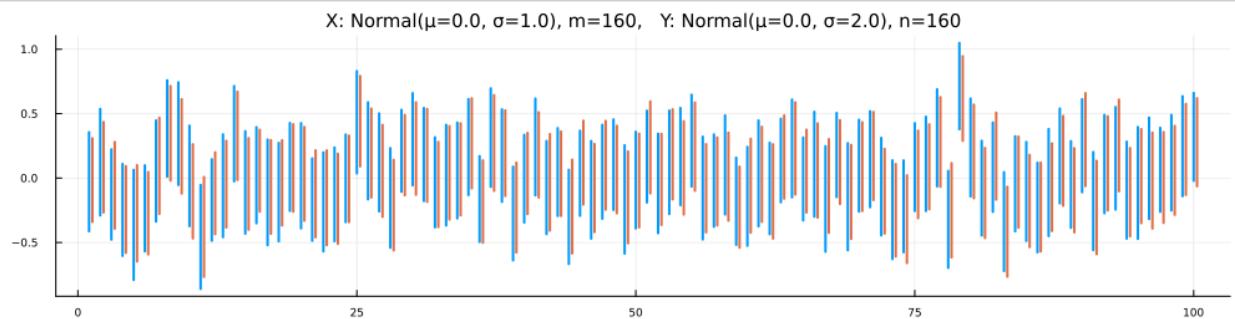
```
In [66]: 1 plot_confints(distx = Normal(0, 1), disty = Normal(0, 2), m = 40, n = 40)
```

Out[66]:



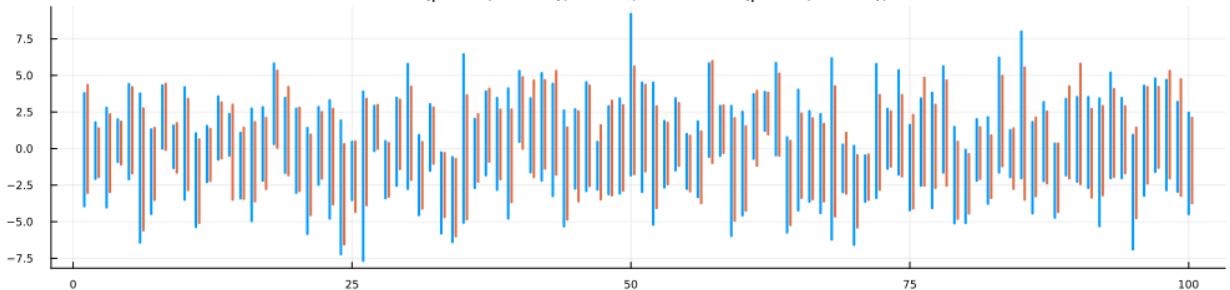
```
In [67]: 1 plot_confints(distx = Normal(0, 1), disty = Normal(0, 2), m = 160, n = 160)
```

Out[67]:



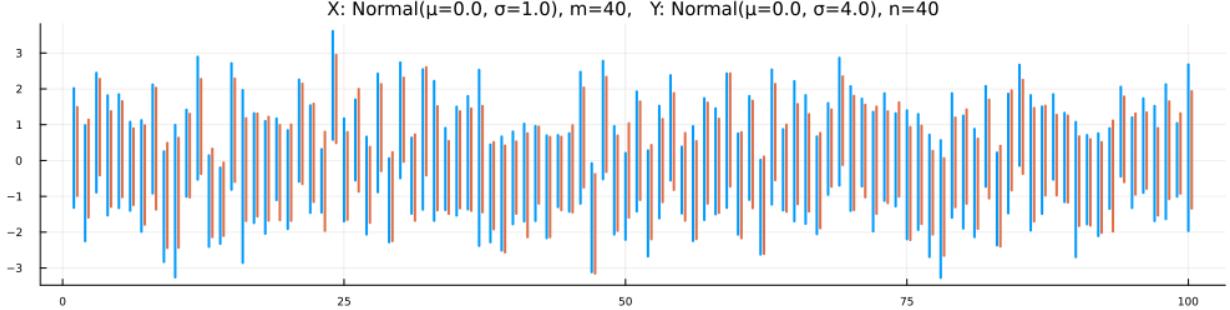
```
In [68]: 1 plot_confints(distx = Normal(0, 1), disty = Normal(0, 4), m = 10, n = 10)
```

Out[68]:



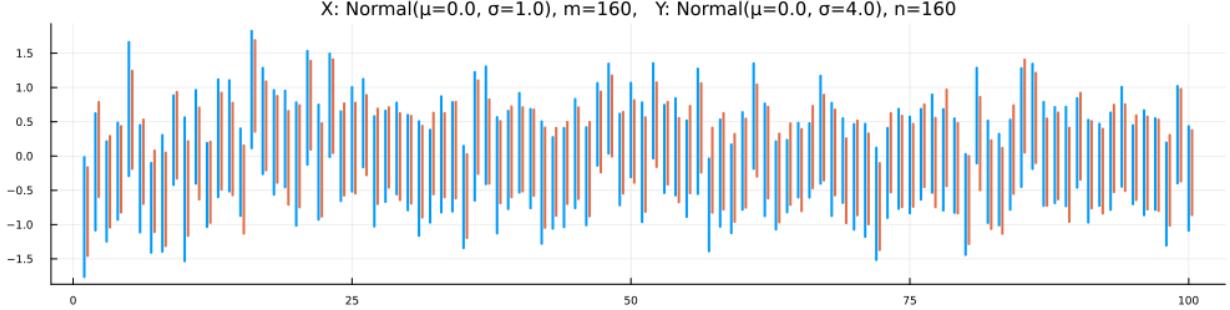
```
In [69]: 1 plot_confints(distx = Normal(0, 1), disty = Normal(0, 4), m = 40, n = 40)
```

Out[69]:



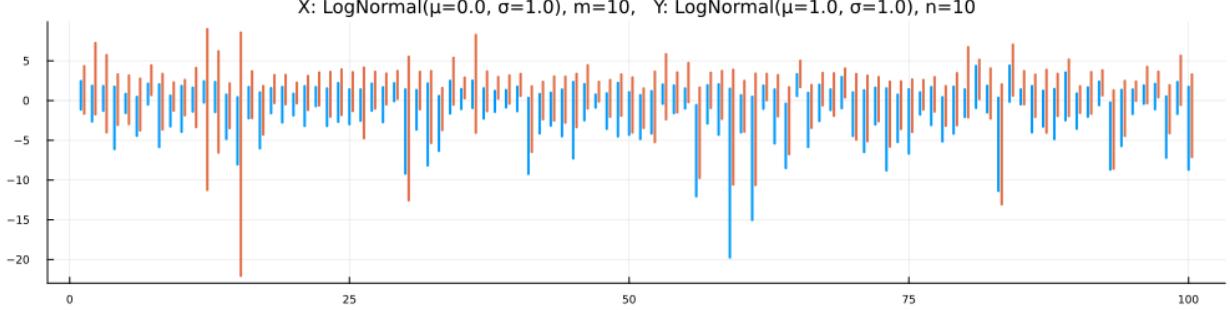
```
In [70]: 1 plot_confints(distx = Normal(0, 1), disty = Normal(0, 4), m = 160, n = 160)
```

Out[70]:



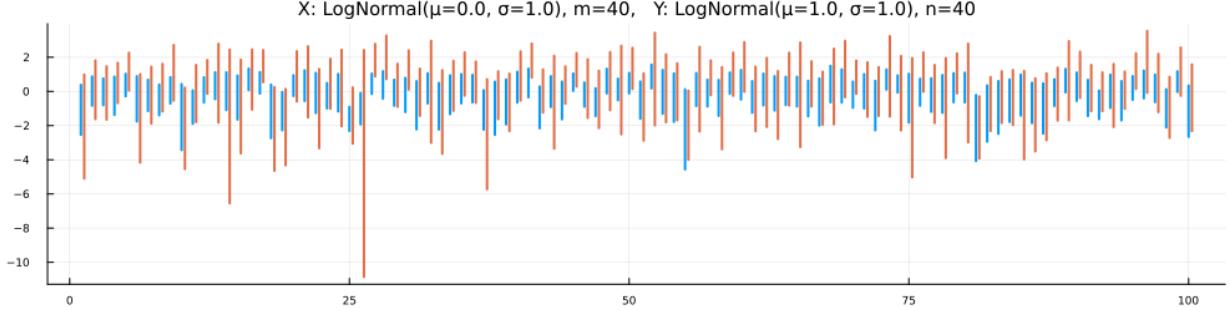
```
In [71]: 1 plot_confints(distx = LogNormal(0), disty = LogNormal(1), m = 10, n = 10)
```

Out[71]:



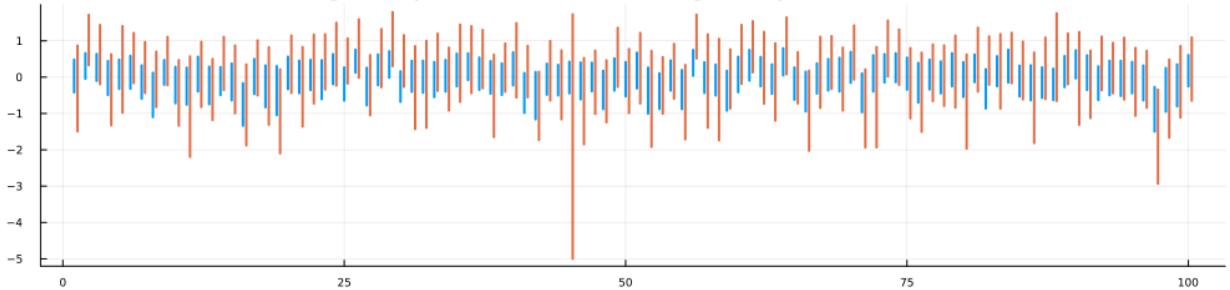
```
In [72]: 1 plot_confints(distx = LogNormal(0), disty = LogNormal(1), m = 40, n = 40)
```

Out[72]:



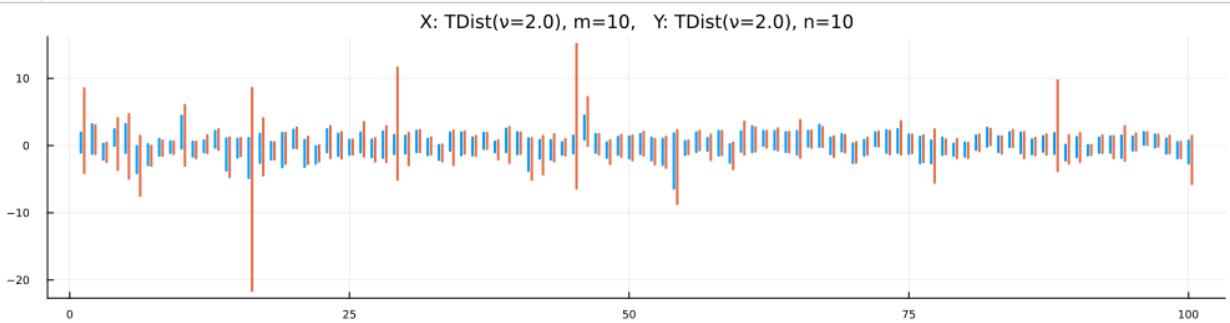
```
In [73]: 1 plot_confints(distx = LogNormal(0), disty = LogNormal(1), m = 160, n = 160)
```

Out[73]:



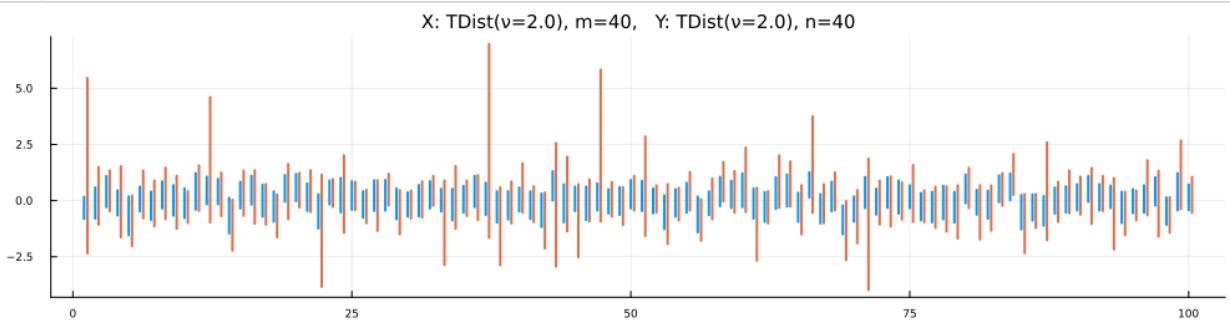
```
In [74]: 1 plot_confints(distx = TDist(2), disty = TDist(2), m = 10, n = 10)
```

Out[74]:



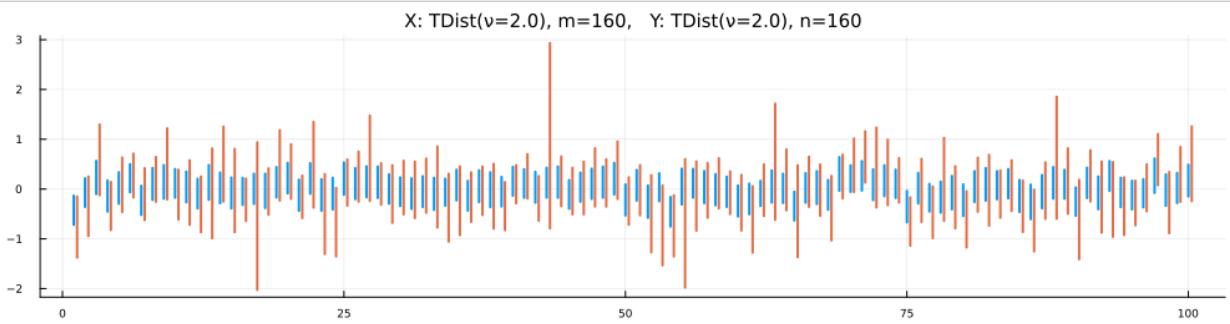
```
In [75]: 1 plot_confints(distx = TDist(2), disty = TDist(2), m = 40, n = 40)
```

Out[75]:



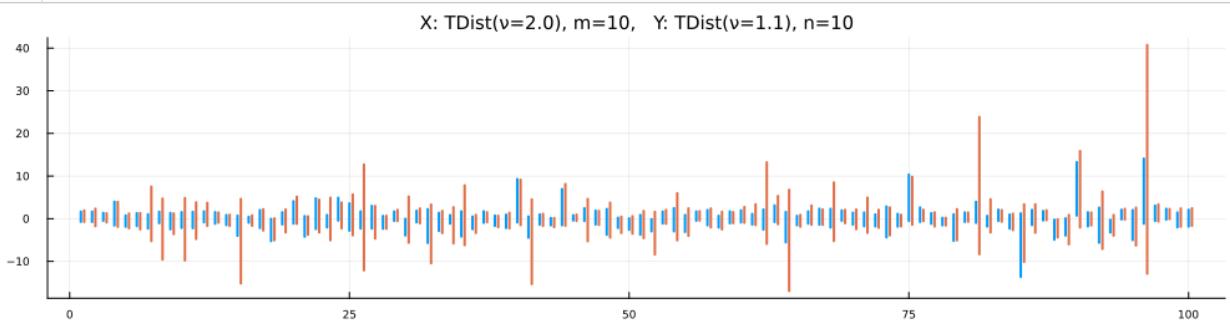
```
In [76]: 1 plot_confints(distx = TDist(2), disty = TDist(2), m = 160, n = 160)
```

Out[76]:



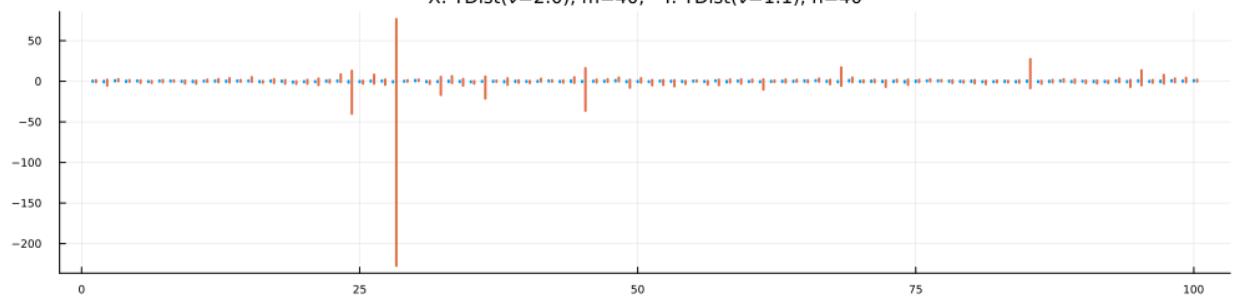
```
In [77]: 1 plot_confints(distx = TDist(2), disty = TDist(1.1), m = 10, n = 10)
```

Out[77]:



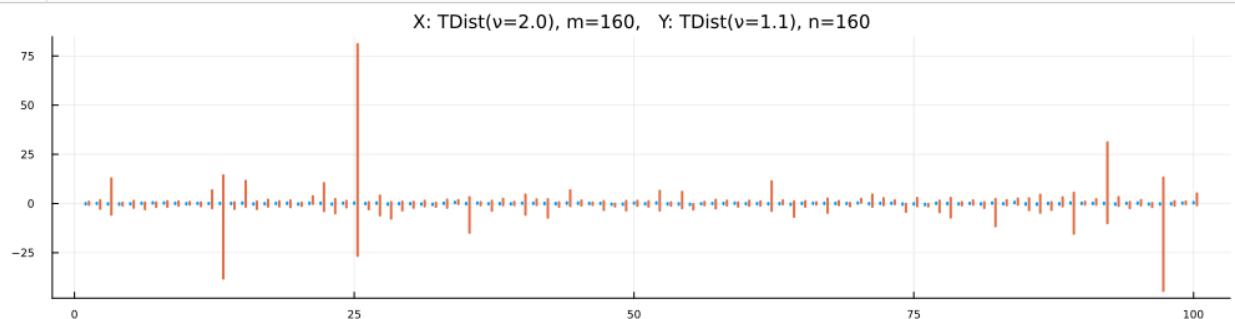
```
In [78]: 1 plot_confints(distx = TDist(2), disty = TDist(1.1), m = 40, n = 40)
```

Out[78]:



```
In [79]: 1 plot_confints(distx = TDist(2), disty = TDist(1.1), m = 160, n = 160)
```

Out[79]:



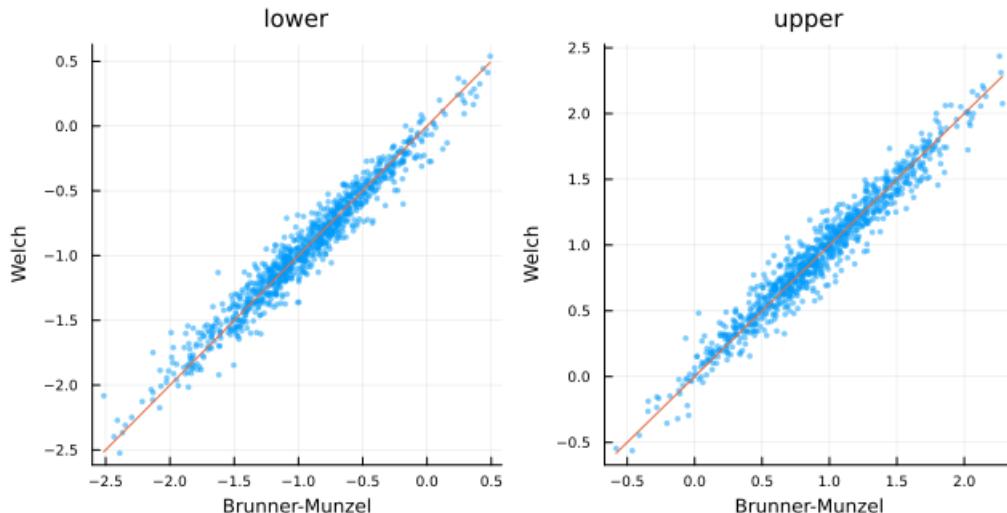
```
In [80]: 1 function plot_limits();
2     distx = Normal(0, 1), disty = Normal(0, 2), m = 10, n = 10,
3     L = 1000, kwargs...)
4
5     @show distx, m
6     @show disty, n
7
8     a = tieshift(distx, disty)
9     Δμ = mean(distx) - mean(disty)
10
11    BM = fill(zeros(2), 0)
12    W = fill(zeros(2), 0)
13    for _ in 1:L
14        X = rand(distx, m)
15        Y = rand(disty, n)
16        push!(BM, brunner_munzel(X, Y .+ a).confint_shift)
17        push!(W, confint_welch(X, Y .+ Δμ))
18    end
19
20    lower = [(first(BM[i]), first(W[i])) for i in 1:L]
21    upper = [(last(BM[i]), last(W[i])) for i in 1:L]
22
23    P1 = scatter(lower; label="", msc=:auto, ms=2, ma=0.5)
24    plot!(identity; label="")
25    plot!(xguide="Brunner-Munzel", yguide="Welch")
26    title!("lower")
27
28    P2 = scatter(upper; label="", msc=:auto, ms=2, ma=0.5)
29    plot!(identity; label="")
30    plot!(xguide="Brunner-Munzel", yguide="Welch")
31    title!("upper")
32
33    plot(P1, P2; size=(640, 320))
34 end
```

Out[80]: plot\_limits (generic function with 1 method)

```
In [81]: 1 plot_limits(distx = Normal(0, 1), disty = Normal(0, 1), m = 10, n = 10)
```

```
(distx, m) = (Normal{Float64}(\mu=0.0, σ=1.0), 10)
(disty, n) = (Normal{Float64}(\mu=0.0, σ=1.0), 10)
```

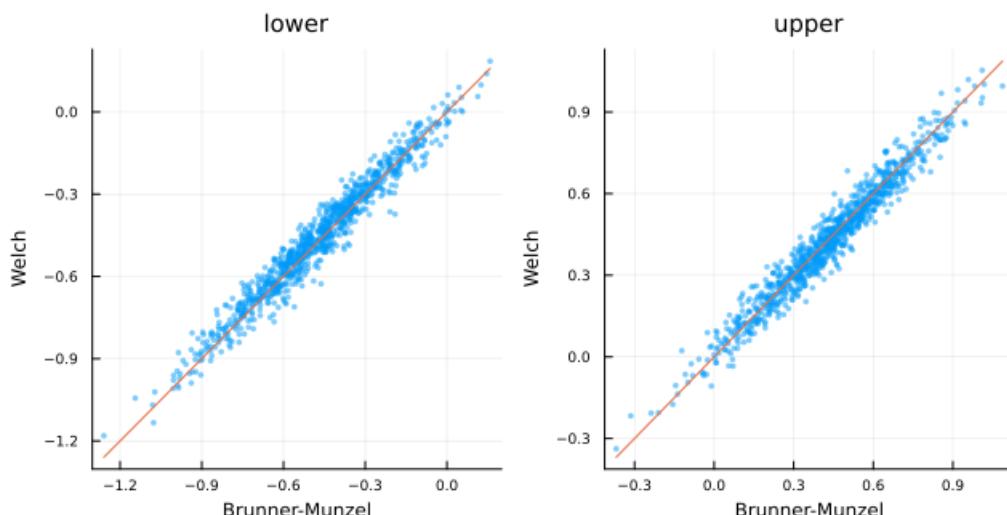
Out[81]:



```
In [82]: 1 plot_limits(distx = Normal(0, 1), disty = Normal(0, 1), m = 40, n = 40)
```

```
(distx, m) = (Normal{Float64}(\mu=0.0, σ=1.0), 40)
(disty, n) = (Normal{Float64}(\mu=0.0, σ=1.0), 40)
```

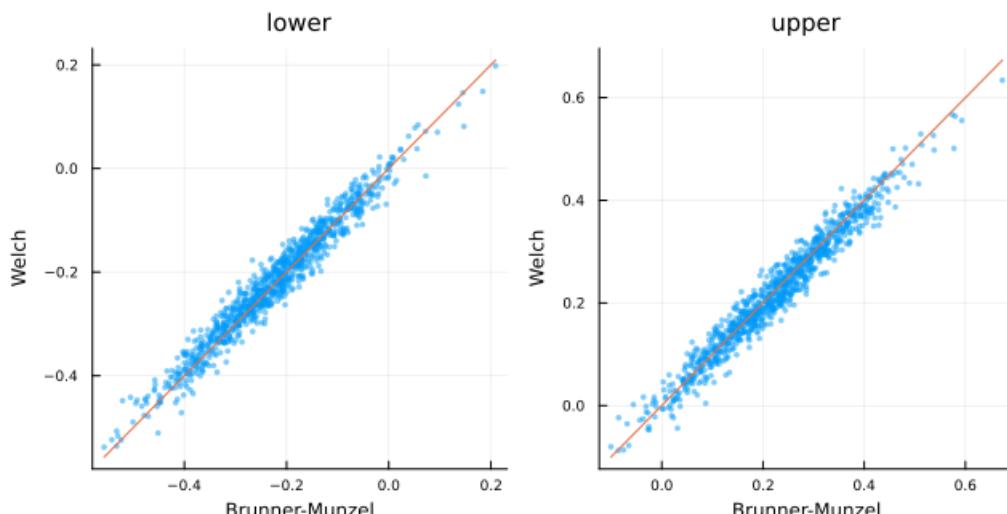
Out[82]:



```
In [83]: 1 plot_limits(distx = Normal(0, 1), disty = Normal(0, 1), m = 160, n = 160)
```

```
(distx, m) = (Normal{Float64}(\mu=0.0, σ=1.0), 160)
(disty, n) = (Normal{Float64}(\mu=0.0, σ=1.0), 160)
```

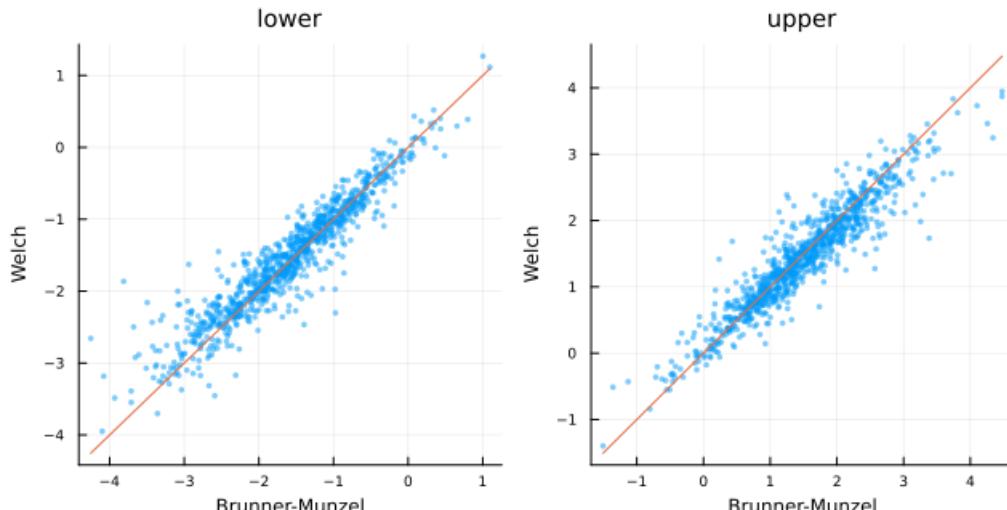
Out[83]:



```
In [84]: 1 plot_limits(distx = Normal(0, 1), disty = Normal(0, 2), m = 10, n = 10)
```

```
(distx, m) = (Normal{Float64}(\mu=0.0, σ=1.0), 10)
(disty, n) = (Normal{Float64}(\mu=0.0, σ=2.0), 10)
```

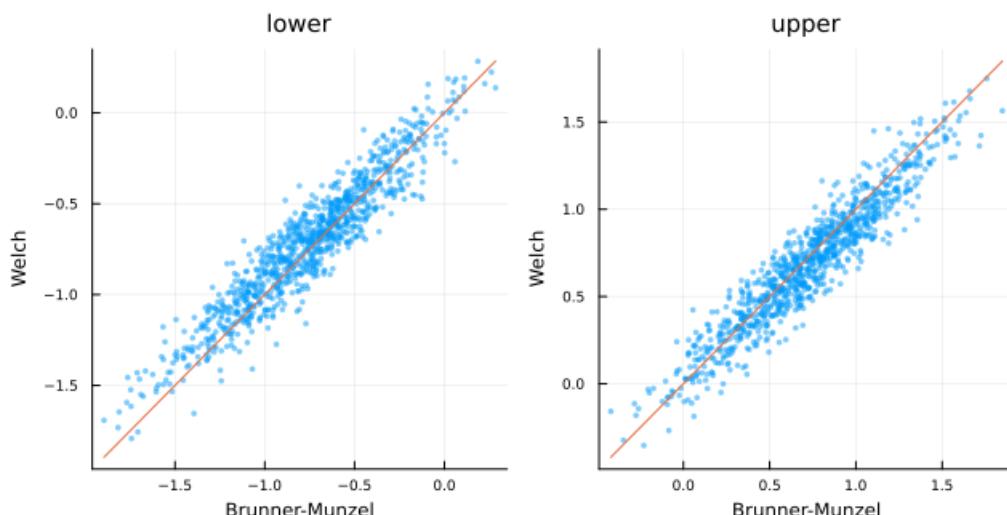
Out[84]:



```
In [85]: 1 plot_limits(distx = Normal(0, 1), disty = Normal(0, 2), m = 40, n = 40)
```

```
(distx, m) = (Normal{Float64}(\mu=0.0, σ=1.0), 40)
(disty, n) = (Normal{Float64}(\mu=0.0, σ=2.0), 40)
```

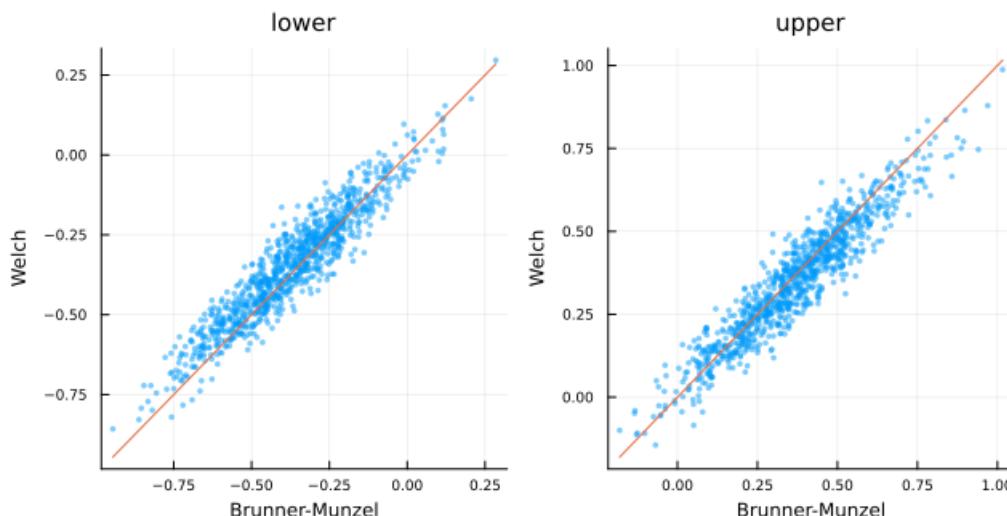
Out[85]:



```
In [86]: 1 plot_limits(distx = Normal(0, 1), disty = Normal(0, 2), m = 160, n = 160)
```

```
(distx, m) = (Normal{Float64}(\mu=0.0, σ=1.0), 160)
(disty, n) = (Normal{Float64}(\mu=0.0, σ=2.0), 160)
```

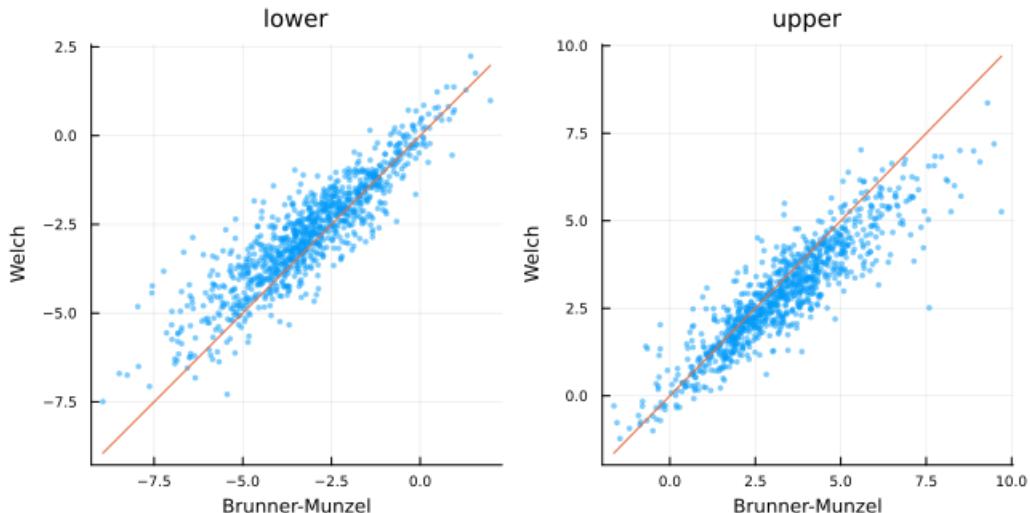
Out[86]:



```
In [87]: 1 plot_limits(distx = Normal(0, 1), disty = Normal(0, 4), m = 10, n = 10)
```

```
(distx, m) = (Normal{Float64}(\mu=0.0, σ=1.0), 10)
(disty, n) = (Normal{Float64}(\mu=0.0, σ=4.0), 10)
```

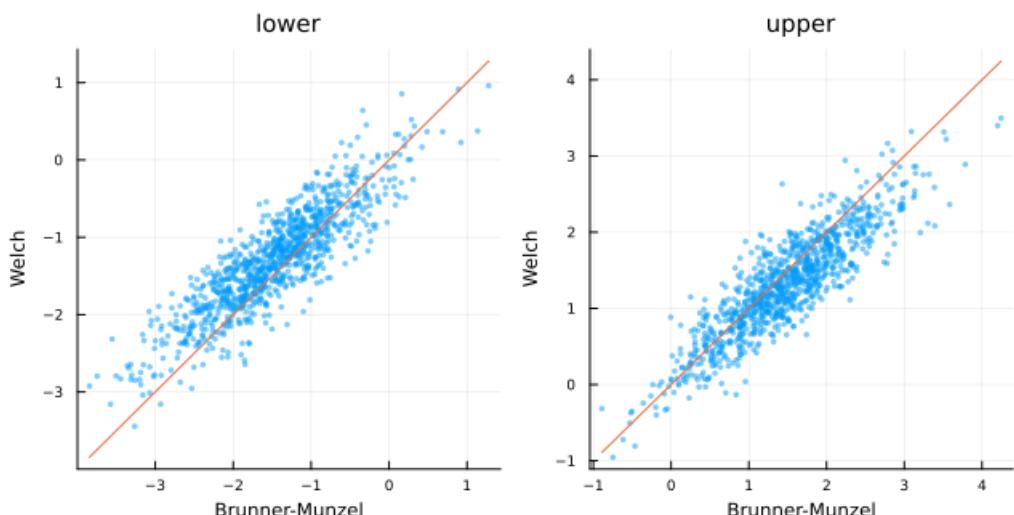
Out[87]:



```
In [88]: 1 plot_limits(distx = Normal(0, 1), disty = Normal(0, 4), m = 40, n = 40)
```

```
(distx, m) = (Normal{Float64}(\mu=0.0, σ=1.0), 40)
(disty, n) = (Normal{Float64}(\mu=0.0, σ=4.0), 40)
```

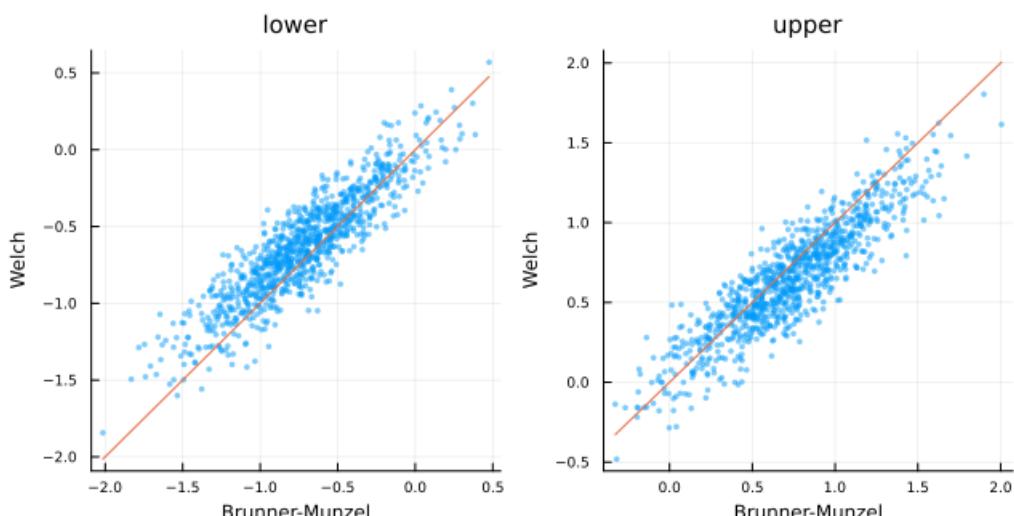
Out[88]:



```
In [89]: 1 plot_limits(distx = Normal(0, 1), disty = Normal(0, 4), m = 160, n = 160)
```

```
(distx, m) = (Normal{Float64}(\mu=0.0, σ=1.0), 160)
(disty, n) = (Normal{Float64}(\mu=0.0, σ=4.0), 160)
```

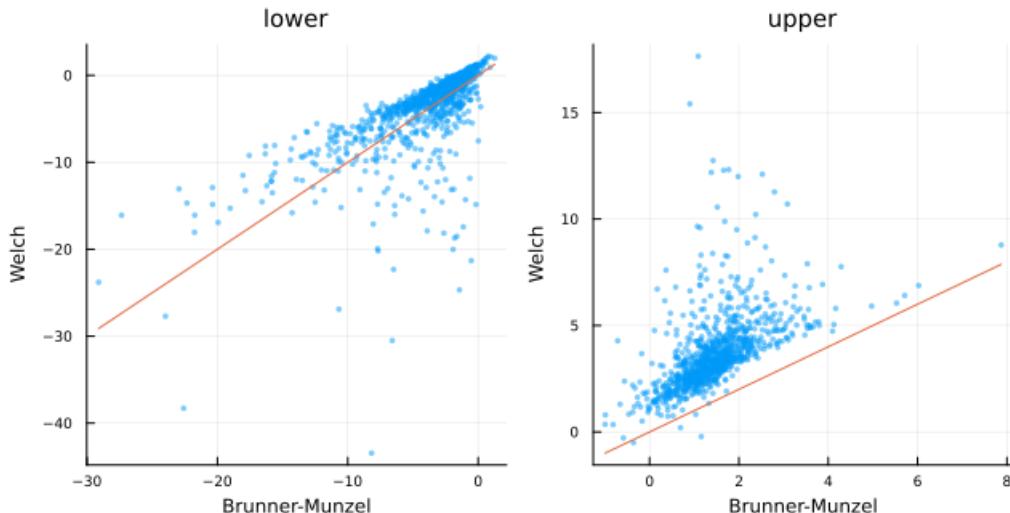
Out[89]:



```
In [90]: 1 plot_limits(distx = LogNormal(), disty = LogNormal(1), m = 10, n = 10)
```

```
(distx, m) = (LogNormal{Float64}(\mu=0.0, σ=1.0), 10)
(disty, n) = (LogNormal{Float64}(\mu=1.0, σ=1.0), 10)
```

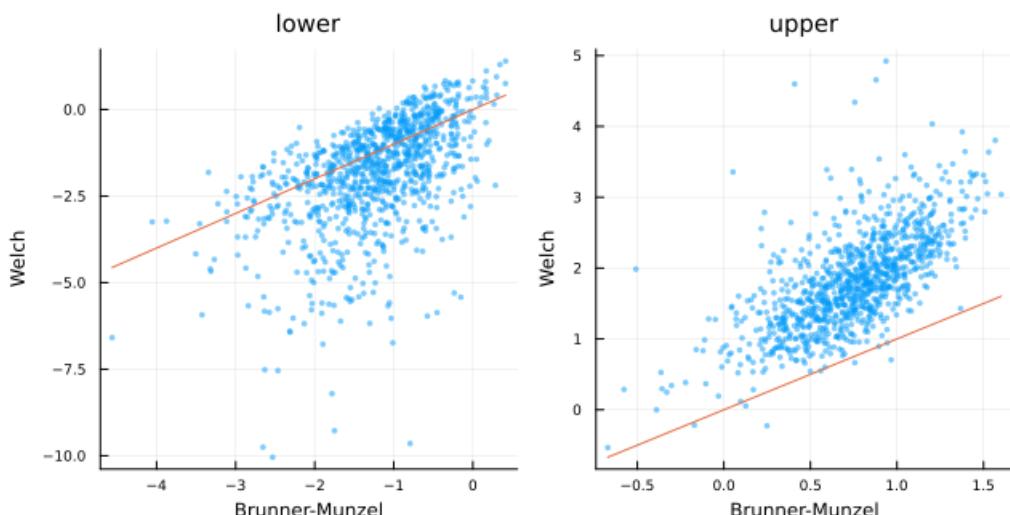
Out[90]:



```
In [91]: 1 plot_limits(distx = LogNormal(), disty = LogNormal(1), m = 40, n = 40)
```

```
(distx, m) = (LogNormal{Float64}(\mu=0.0, σ=1.0), 40)
(disty, n) = (LogNormal{Float64}(\mu=1.0, σ=1.0), 40)
```

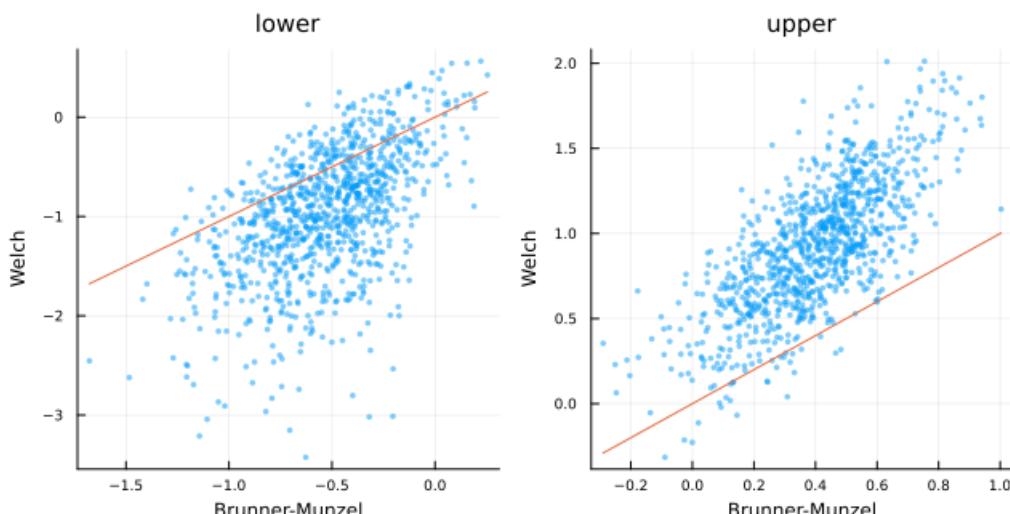
Out[91]:



```
In [92]: 1 @time plot_limits(distx = LogNormal(), disty = LogNormal(1), m = 160, n = 160)
```

```
(distx, m) = (LogNormal{Float64}(\mu=0.0, σ=1.0), 160)
(disty, n) = (LogNormal{Float64}(\mu=1.0, σ=1.0), 160)
7.017568 seconds (284.24 k allocations: 108.661 MiB, 0.26% gc time)
```

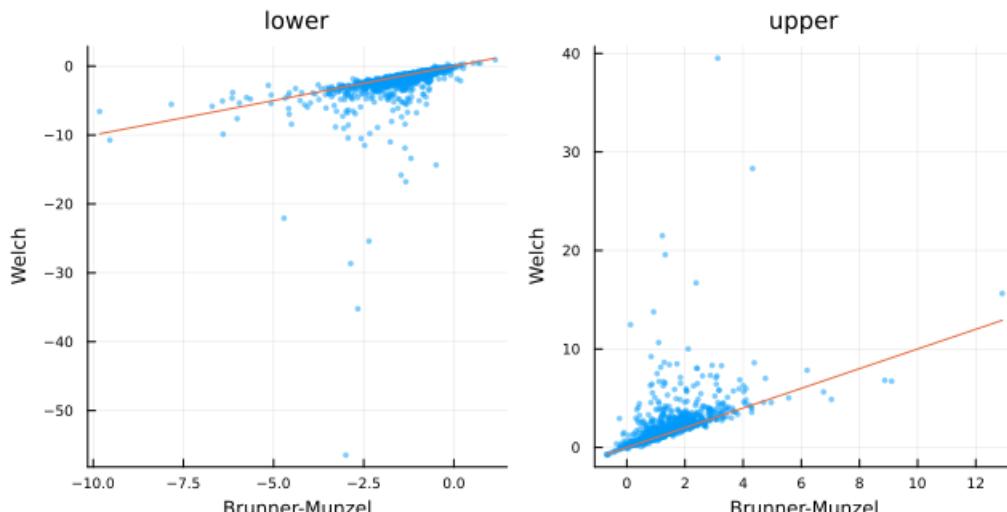
Out[92]:



In [93]: 1 plot\_limits(distx = TDist(2), disty = TDist(2), m = 10, n = 10)

```
(distx, m) = (TDist{Float64}(v=2.0), 10)
(disty, n) = (TDist{Float64}(v=2.0), 10)
```

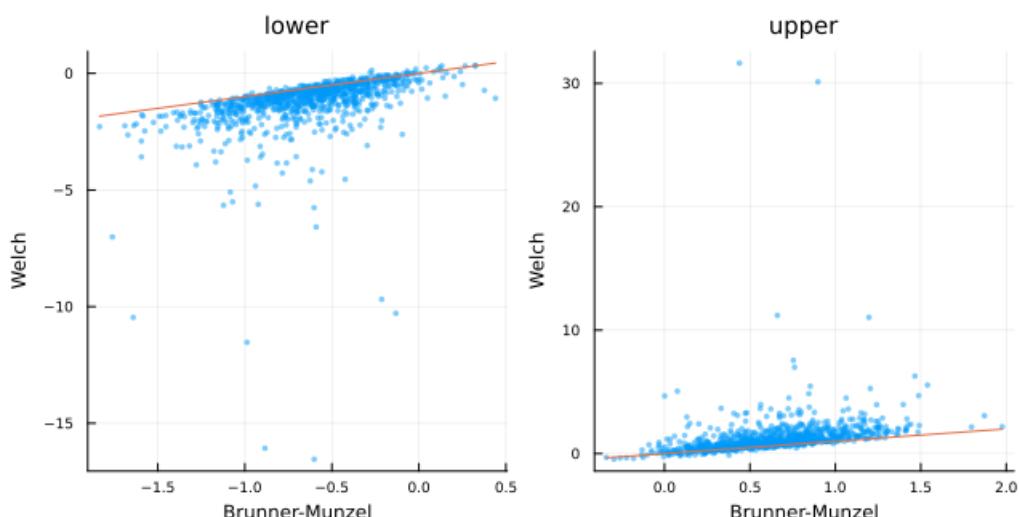
Out[93]:



In [94]: 1 plot\_limits(distx = TDist(2), disty = TDist(2), m = 40, n = 40)

```
(distx, m) = (TDist{Float64}(v=2.0), 40)
(disty, n) = (TDist{Float64}(v=2.0), 40)
```

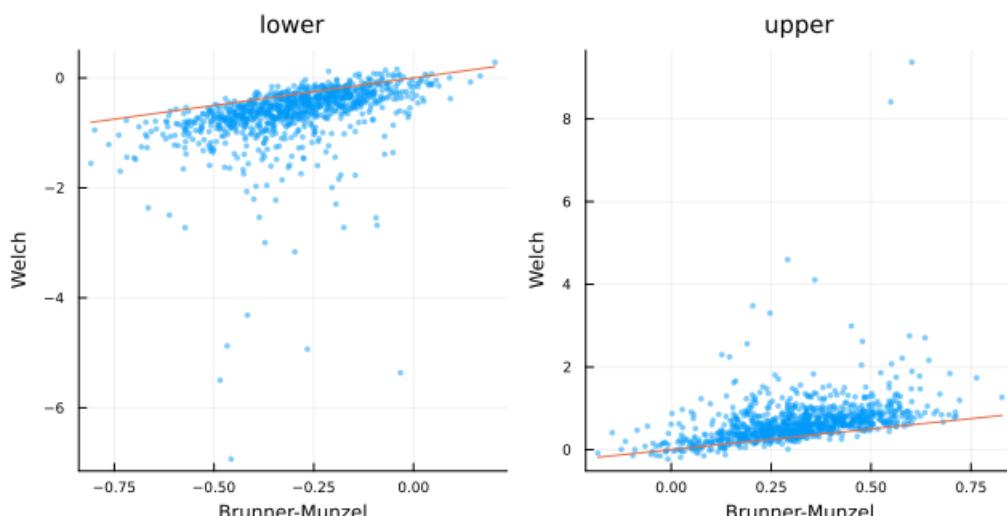
Out[94]:



In [95]: 1 plot\_limits(distx = TDist(2), disty = TDist(2), m = 160, n = 160)

```
(distx, m) = (TDist{Float64}(v=2.0), 160)
(disty, n) = (TDist{Float64}(v=2.0), 160)
```

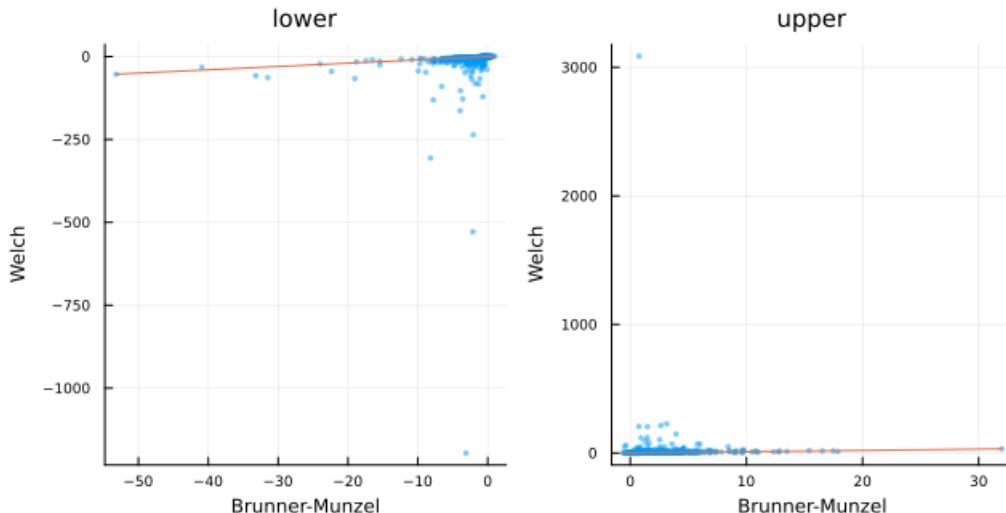
Out[95]:



```
In [96]: 1 plot_limits(distx = TDist(2), disty = TDist(1.1), m = 10, n = 10)
```

```
(distx, m) = (TDist{Float64}(v=2.0), 10)
(disty, n) = (TDist{Float64}(v=1.1), 10)
```

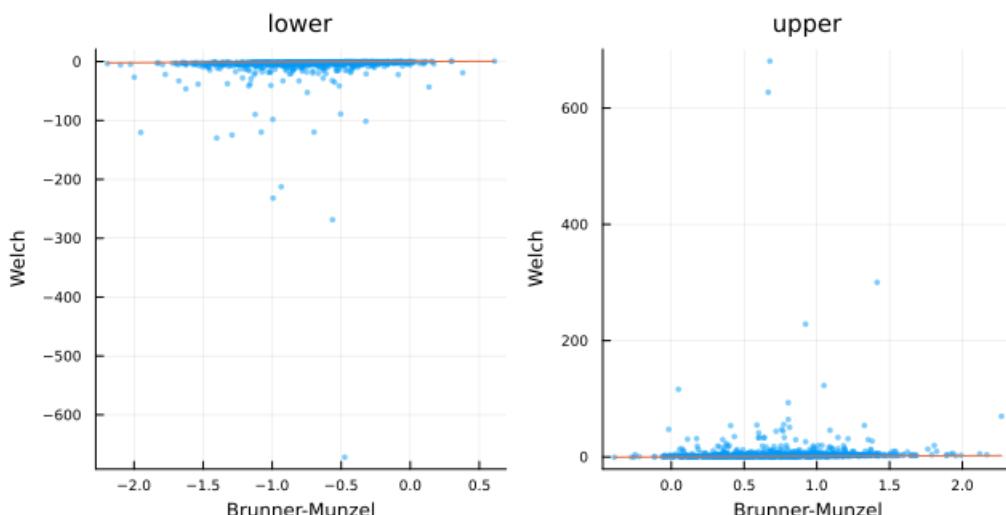
Out[96]:



```
In [97]: 1 plot_limits(distx = TDist(2), disty = TDist(1.1), m = 40, n = 40)
```

```
(distx, m) = (TDist{Float64}(v=2.0), 40)
(disty, n) = (TDist{Float64}(v=1.1), 40)
```

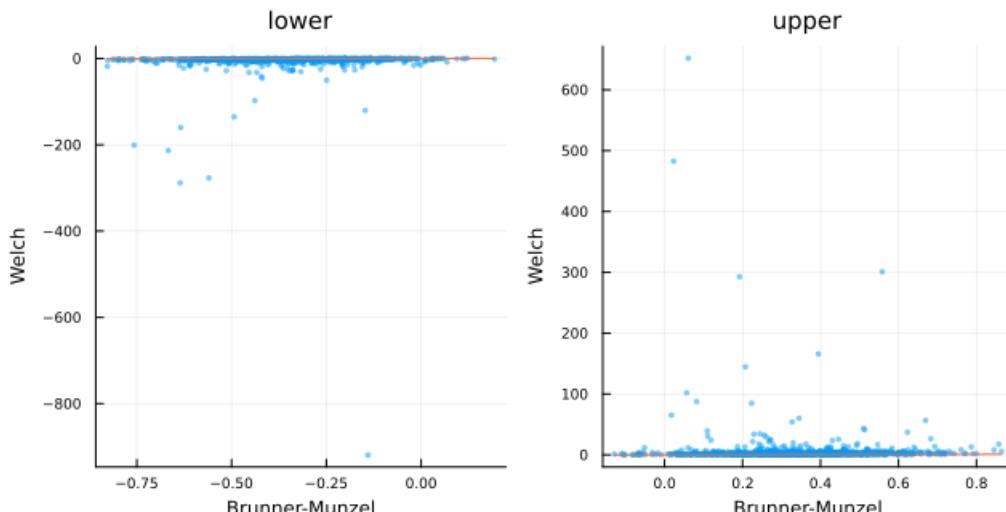
Out[97]:



```
In [98]: 1 plot_limits(distx = TDist(2), disty = TDist(1.1), m = 160, n = 160)
```

```
(distx, m) = (TDist{Float64}(v=2.0), 160)
(disty, n) = (TDist{Float64}(v=1.1), 160)
```

Out[98]:



## 5 小サンプルでのpermutation版の検定とBM検定とWelchのt検定の比較

```
In [99]: 1 function sim_brunner_mumzel_perm();
2     distx = Normal(0, 1), disty = Normal(0, 2), m = 5, n = 5,
3     L = 10^2)
4     pval_bm_perm = Vector{Float64}(undef, L)
5     tmpX = [Vector{Float64}(undef, m) for _ in 1:nthreads()]
6     tmpY = [Vector{Float64}(undef, n) for _ in 1:nthreads()]
7     tmpXandY = [Vector{Float64}(undef, m+n) for _ in 1:nthreads()]
8     tmpTval = [Vector{Float64}(undef, binomial(m+n, m)) for _ in 1:nthreads()]
9     tmpHx = [Vector{Float64}(undef, m) for _ in 1:nthreads()]
10    tmpHy = [Vector{Float64}(undef, n) for _ in 1:nthreads()]
11    tmpccomb = [Vector{Int}(undef, n) for _ in 1:nthreads()]
12    @threads for i in 1:L
13        tid = threadid()
14        X = rand!(distx, tmpX[tid])
15        Y = rand!(disty, tmpY[tid])
16        Tval = permutation_tvalues_brunner_munzel(X, Y,
17            tmpXandY[tid], tmpTval[tid], tmpHx[tid], tmpHy[tid], tmpccomb[tid])
18        tval = statistics_brunner_munzel(X, Y, tmpHx[tid], tmpHy[tid]).tvalue
19        pval_bm_perm[i] = pvalue_brunner_munzel_perm(X, Y, Tval, tval)
20    end
21    ecdf(pval_bm_perm)
22 end
```

Out[99]: sim\_brunner\_mumzel\_perm (generic function with 1 method)

```
In [100]: 1 @time ecdf_bm_perm = sim_brunner_mumzel_perm(
2     distx = Normal(0, 1), disty = Normal(0, 2), m = 7, n = 7, L = 10^4)
```

5.033259 seconds (16.27 M allocations: 4.602 GiB, 15.78% gc time, 9.61% compilation time)

```
Out[100]: ECDF{Vector{Float64}, Weights{Float64, Float64, Vector{Float64}}}([0.0005827505827505828, 0.0005827505827505828, 0.0005827505827505828, 0.0005827505827505828, 0.0005827505827505828, 0.0005827505827505828, 0.0005827505827505828, 0.0011655011655011655, 0.0011655011655011655, 0.0011655011655011655, 0.0011655011655011655 ... 1.0, 1.0, 1.0, 1.0, 1.0, 1.0, 1.0, 1.0, 1.0], Float64[])
```

```
In [101]: 1 function plot_pvals_with_perm();
2     distx = Normal(0, 1),
3     disty = Normal(0, 2),
4     m = 7,
5     n = 7,
6     L = 10^4,
7     kwargs...
8     )
9     a = tieshift(distx, disty)
10    @time ecdf_bm_perm = sim_brunner_mumzel_perm(; distx, disty = disty + a, m, n, L)
11    @time ecdf_bm = sim_brunner_mumzel(; distx, disty = disty + a, m, n, L)
12    Δμ = mean(distx) - mean(disty)
13    @time ecdf_w = sim_welch(; distx, disty = disty + Δμ, m, n, L)
14    @show a Δμ
15
16    plot(legend=:topleft)
17    plot!(α → ecdf_bm_perm(α), 0, 0.1; label="BM permutation")
18    plot!(α → ecdf_bm(α), 0, 0.1; label="Brunner-Munzel", ls=:dash)
19    plot!(α → ecdf_w(α), 0, 0.1; label="Welch", ls=:dashdot)
20    plot!(identity; label="", c=:black, ls=:dot)
21    plot!(xtick=0:0.01:0.1, ytick=0:0.01:1)
22    plot!(xguide="nominal significance level α",
23          yguide="probability of P-value < α")
24    a_ = string(round(a; digits=4))
25    Δμ_ = string(round(Δμ; digits=4))
26    title!("X: $(distname(distx)), m=$m\n"
27           "Y: $(distname(disty))+(a, Δμ), n=$n\n"
28           "a=$a_, Δμ=$Δμ_")
29    plot!(size=(400, 450), titlefontsize=9)
30    plot!(; kwargs...)
31 end
```

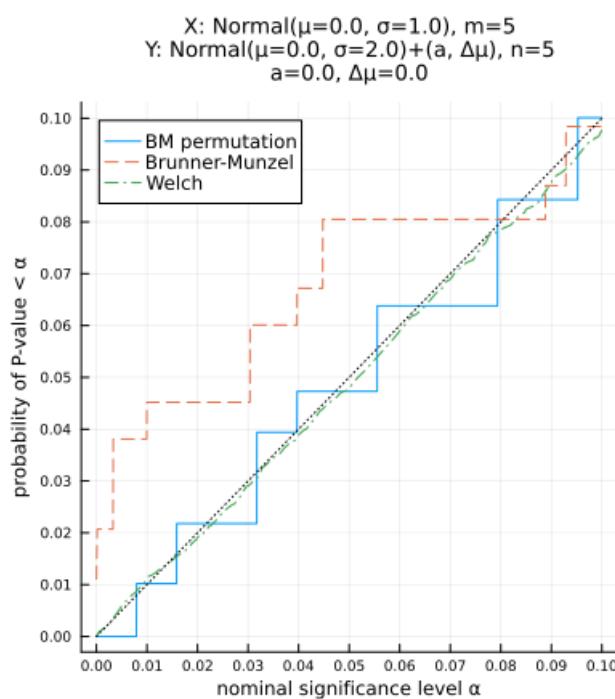
Out[101]: plot\_pvals\_with\_perm (generic function with 1 method)

In [102]:

```
1 plot_pvals_with_perm(
2     distx = Normal(0, 1), disty = Normal(0, 2), m = 5, n = 5, L = 10^4)
```

0.363926 seconds (1.22 M allocations: 350.587 MiB, 29.71% gc time)  
 0.024036 seconds (4.97 k allocations: 1.724 MiB)  
 0.001940 seconds (5.94 k allocations: 2.009 MiB)  
 $a = 7.685641860444171e-14$   
 $\Delta\mu = 0.0$

Out[102]:

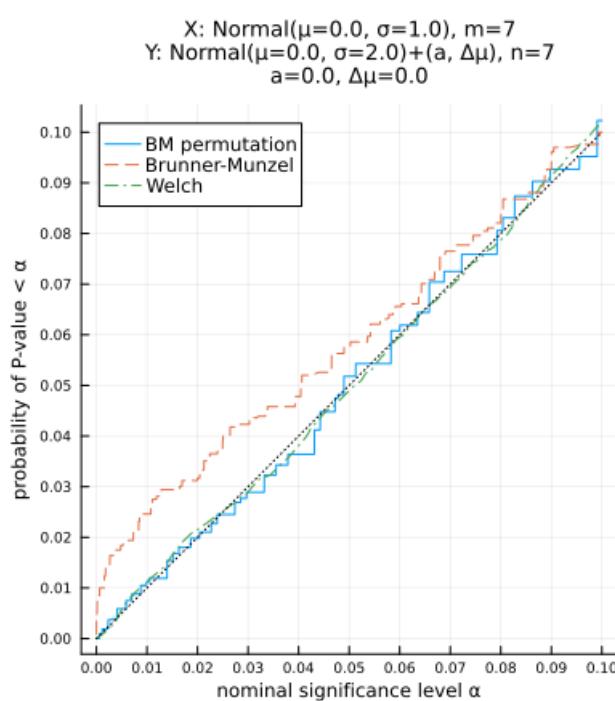


In [103]:

```
1 plot_pvals_with_perm(
2     distx = Normal(0, 1), disty = Normal(0, 2), m = 7, n = 7, L = 10^4)
```

5.505243 seconds (16.26 M allocations: 4.601 GiB, 16.43% gc time)  
 0.017051 seconds (5.07 k allocations: 1.753 MiB)  
 0.001551 seconds (5.75 k allocations: 1.955 MiB)  
 $a = 7.685641860444171e-14$   
 $\Delta\mu = 0.0$

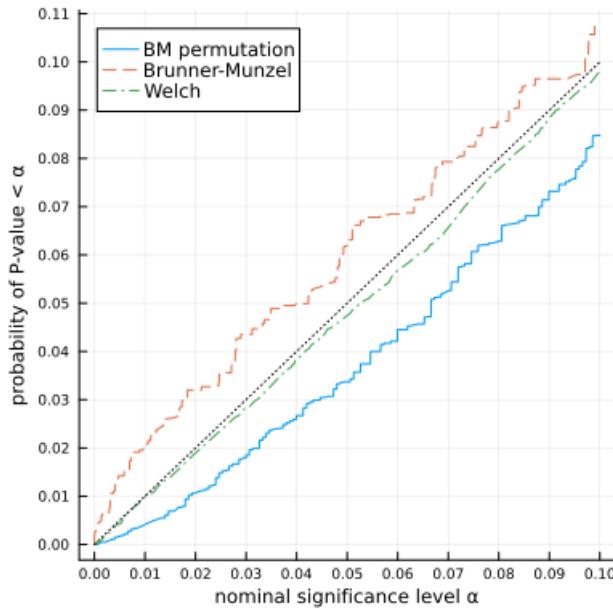
Out[103]:



```
In [104]: 1 plot_pvals_with_perm(
2     distx = Normal(0, 1), disty = Normal(0, 2), m = 5, n = 10, L = 10^4)

4.904059 seconds (15.70 M allocations: 4.442 GiB, 15.94% gc time)
0.002604 seconds (5.22 k allocations: 1.797 MiB)
0.011408 seconds (5.65 k allocations: 1.925 MiB)
a = 7.685641860444171e-14
Δμ = 0.0
```

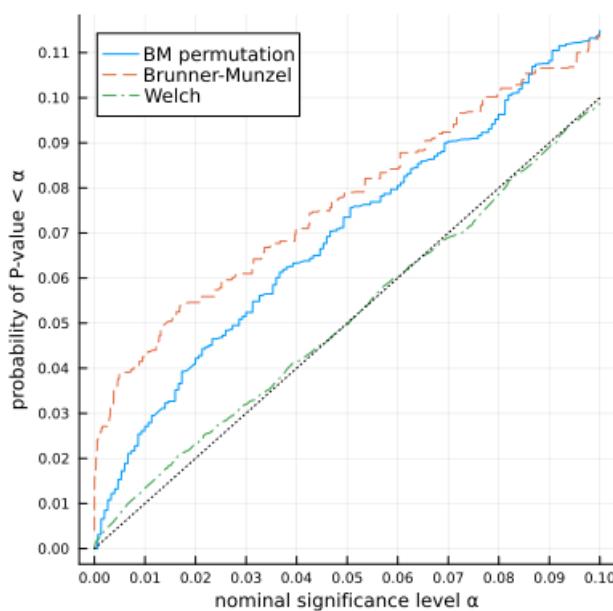
Out[104]:  
X:  $\text{Normal}(\mu=0.0, \sigma=1.0)$ , m=5  
Y:  $\text{Normal}(\mu=0.0, \sigma=2.0)+(a, \Delta\mu)$ , n=10  
a=0.0, Δμ=0.0



```
In [105]: 1 plot_pvals_with_perm(
2     distx = Normal(0, 1), disty = Normal(0, 2), m = 10, n = 5, L = 10^4)

4.785355 seconds (15.70 M allocations: 4.443 GiB, 15.10% gc time)
0.001891 seconds (5.21 k allocations: 1.794 MiB)
0.002406 seconds (6.05 k allocations: 2.042 MiB)
a = 7.685641860444171e-14
Δμ = 0.0
```

Out[105]:  
X:  $\text{Normal}(\mu=0.0, \sigma=1.0)$ , m=10  
Y:  $\text{Normal}(\mu=0.0, \sigma=2.0)+(a, \Delta\mu)$ , n=5  
a=0.0, Δμ=0.0



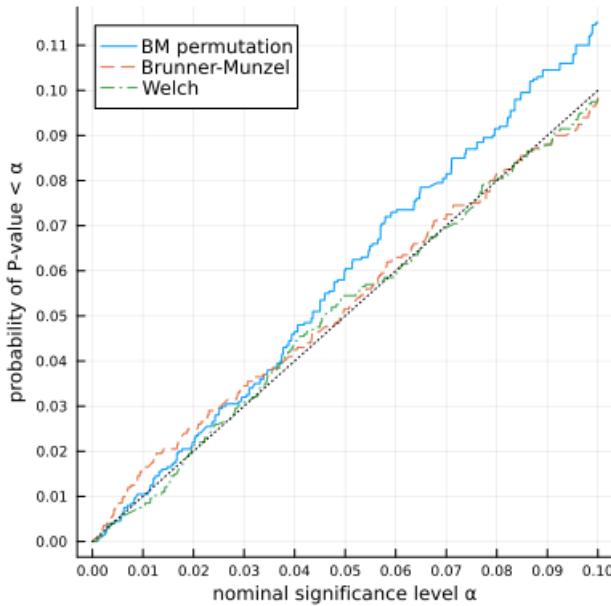
In [106]:

```
1 plot_pvals_with_perm(
2     distx = Normal(0, 1), disty = Normal(0, 2), m = 10, n = 10, L = 2000)
```

67.435797 seconds (203.34 M allocations: 57.586 GiB, 16.83% gc time)  
 0.021784 seconds (1.18 k allocations: 391.758 KiB)  
 0.015124 seconds (1.31 k allocations: 432.234 KiB)  
 $a = 7.685641860444171e-14$   
 $\Delta\mu = 0.0$

Out[106]:

X:  $\text{Normal}(\mu=0.0, \sigma=1.0)$ , m=10  
 Y:  $\text{Normal}(\mu=0.0, \sigma=2.0)+(a, \Delta\mu)$ , n=10  
 $a=0.0, \Delta\mu=0.0$



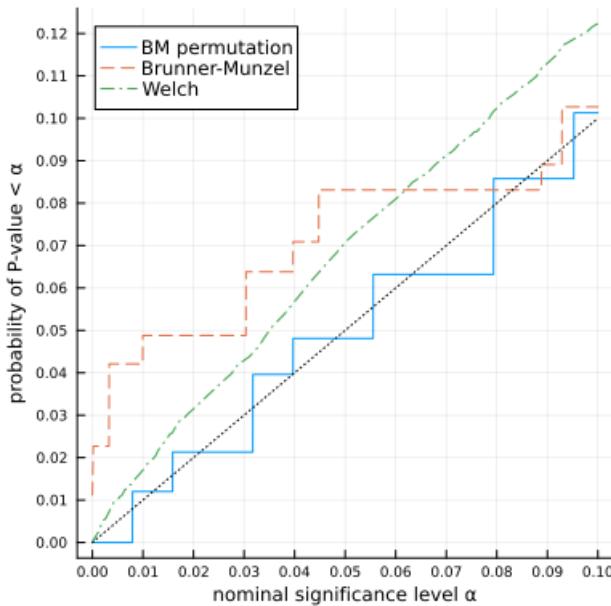
In [107]:

```
1 plot_pvals_with_perm(
2     distx = Exponential(1), disty = Exponential(2),
3     m = 5, n = 5, L = 10^4)
```

0.343144 seconds (1.23 M allocations: 351.582 MiB, 221.71% compilation time)  
 0.035708 seconds (14.38 k allocations: 2.305 MiB, 1145.51% compilation time)  
 0.030994 seconds (14.93 k allocations: 2.644 MiB, 1141.61% compilation time)  
 $a = -0.5753641445892759$   
 $\Delta\mu = -1.0$

Out[107]:

X:  $\text{Exponential}(\theta=1.0)$ , m=5  
 Y:  $\text{Exponential}(\theta=2.0)+(a, \Delta\mu)$ , n=5  
 $a=-0.5754, \Delta\mu=-1.0$



In [108]:

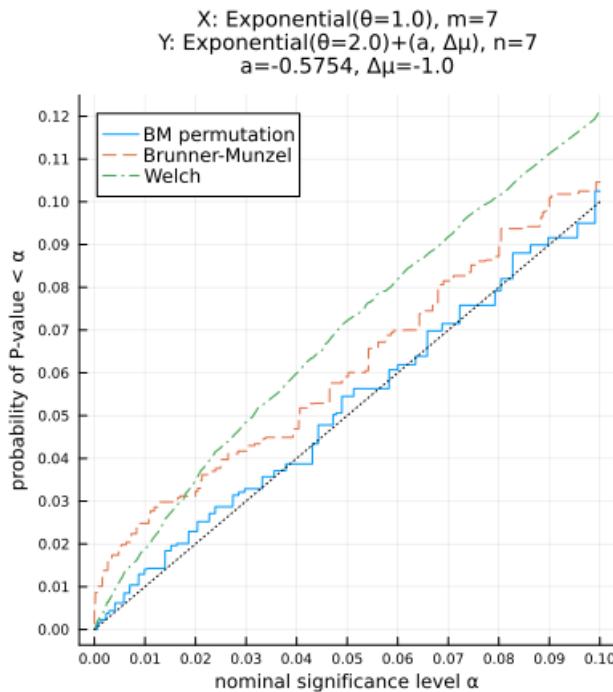
```

1 plot_pvals_with_perm(
2     distx = Exponential(1), disty = Exponential(2),
3     m = 7, n = 7, L = 10^4)

```

5.384772 seconds (16.26 M allocations: 4.601 GiB, 16.60% gc time)  
 0.020699 seconds (5.12 k allocations: 1.768 MiB)  
 0.002418 seconds (5.96 k allocations: 2.017 MiB)  
 $a = -0.5753641445892759$   
 $\Delta\mu = -1.0$

Out[108]:



In [109]:

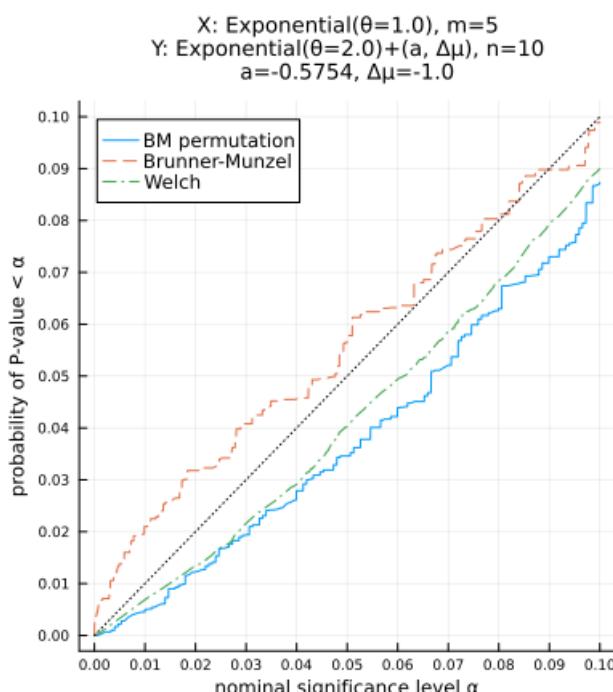
```

1 plot_pvals_with_perm(
2     distx = Exponential(1), disty = Exponential(2),
3     m = 5, n = 10, L = 10^4)

```

4.838760 seconds (15.70 M allocations: 4.442 GiB, 17.10% gc time)  
 0.001775 seconds (5.13 k allocations: 1.769 MiB)  
 0.001546 seconds (5.77 k allocations: 1.960 MiB)  
 $a = -0.5753641445892759$   
 $\Delta\mu = -1.0$

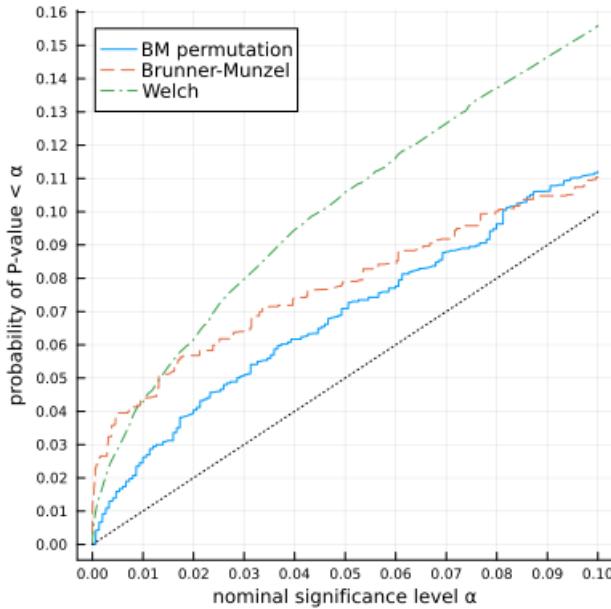
Out[109]:



```
In [110]: 1 plot_pvals_with_perm(
2     distx = Exponential(1), disty = Exponential(2),
3     m = 10, n = 5, L = 10^4)
```

4.919973 seconds (15.70 M allocations: 4.443 GiB, 18.50% gc time)  
 0.015176 seconds (5.34 k allocations: 1.829 MiB)  
 0.002276 seconds (6.43 k allocations: 2.149 MiB)  
 $a = -0.5753641445892759$   
 $\Delta\mu = -1.0$

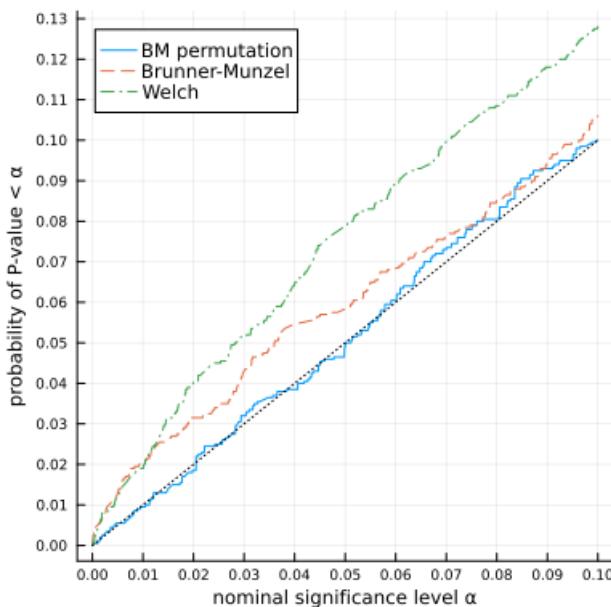
Out[110]: X: Exponential( $\theta=1.0$ ), m=10  
 Y: Exponential( $\theta=2.0$ )+(a,  $\Delta\mu$ ), n=5  
 $a=-0.5754$ ,  $\Delta\mu=-1.0$



```
In [111]: 1 plot_pvals_with_perm(
2     distx = Exponential(1), disty = Exponential(2),
3     m = 10, n = 10, L = 2000)
```

66.374139 seconds (203.34 M allocations: 57.586 GiB, 16.94% gc time)  
 0.000406 seconds (1.14 k allocations: 382.258 KiB)  
 0.000322 seconds (1.27 k allocations: 422.703 KiB)  
 $a = -0.5753641445892759$   
 $\Delta\mu = -1.0$

Out[111]: X: Exponential( $\theta=1.0$ ), m=10  
 Y: Exponential( $\theta=2.0$ )+(a,  $\Delta\mu$ ), n=10  
 $a=-0.5754$ ,  $\Delta\mu=-1.0$

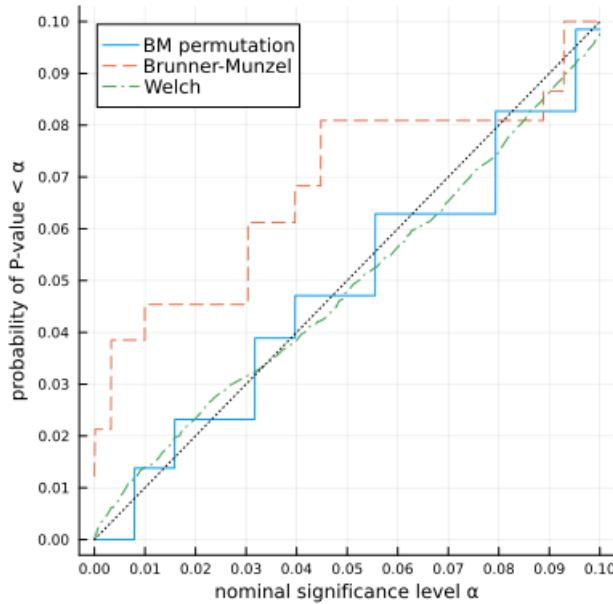


```
In [112]: 1 plot_pvals_with_perm(
2     distx = Uniform(-1, 1), disty = Exponential(0.5773502691896257),
3     m = 5, n = 5, L = 10^4)
```

0.369498 seconds (1.23 M allocations: 351.212 MiB, 17.51% gc time, 105.53% compilation time)  
0.024506 seconds (4.75 k allocations: 1.658 MiB)  
0.056172 seconds (14.33 k allocations: 2.446 MiB, 1154.43% compilation time)  
a = -0.5370568188698567  
 $\Delta\mu$  = -0.5773502691896257

Out[112]:

X: Uniform(a=-1.0, b=1.0), m=5  
Y: Exponential( $\theta=0.57735$ )+(a,  $\Delta\mu$ ), n=5  
a=-0.5371,  $\Delta\mu$ =-0.5774

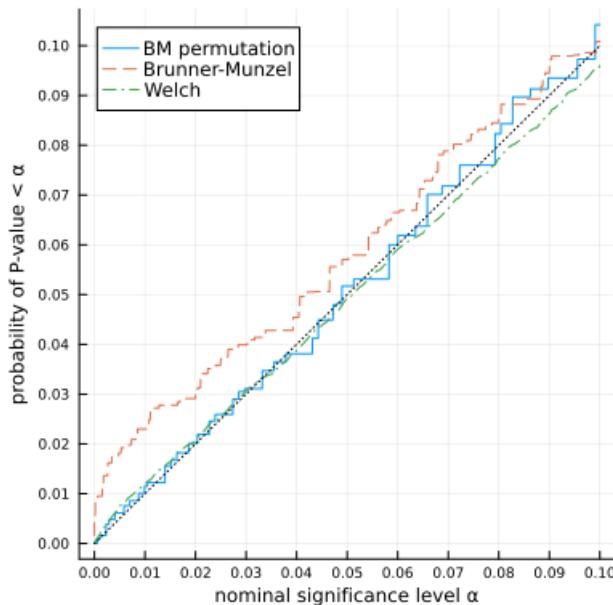


```
In [113]: 1 plot_pvals_with_perm(
2     distx = Uniform(-1, 1), disty = Exponential(0.5773502691896257),
3     m = 7, n = 7, L = 10^4)
```

5.337259 seconds (16.25 M allocations: 4.601 GiB, 17.09% gc time)  
0.001646 seconds (4.90 k allocations: 1.703 MiB)  
0.018274 seconds (5.44 k allocations: 1.864 MiB)  
a = -0.5370568188698567  
 $\Delta\mu$  = -0.5773502691896257

Out[113]:

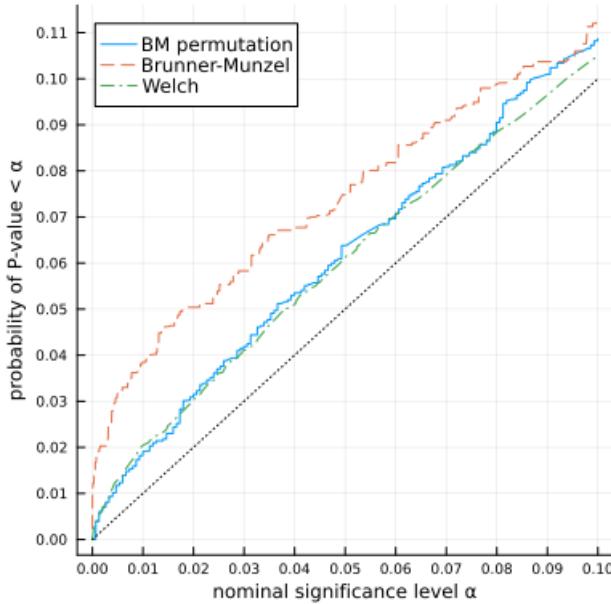
X: Uniform(a=-1.0, b=1.0), m=7  
Y: Exponential( $\theta=0.57735$ )+(a,  $\Delta\mu$ ), n=7  
a=-0.5371,  $\Delta\mu$ =-0.5774



```
In [114]: 1 plot_pvals_with_perm(
2     distx = Uniform(-1, 1), disty = Exponential(0.5773502691896257),
3     m = 5, n = 10, L = 10^4)

4.874172 seconds (15.70 M allocations: 4.442 GiB, 18.91% gc time)
0.002489 seconds (5.45 k allocations: 1.862 MiB)
0.001980 seconds (5.71 k allocations: 1.943 MiB)
a = -0.5370568188698567
Δμ = -0.5773502691896257
```

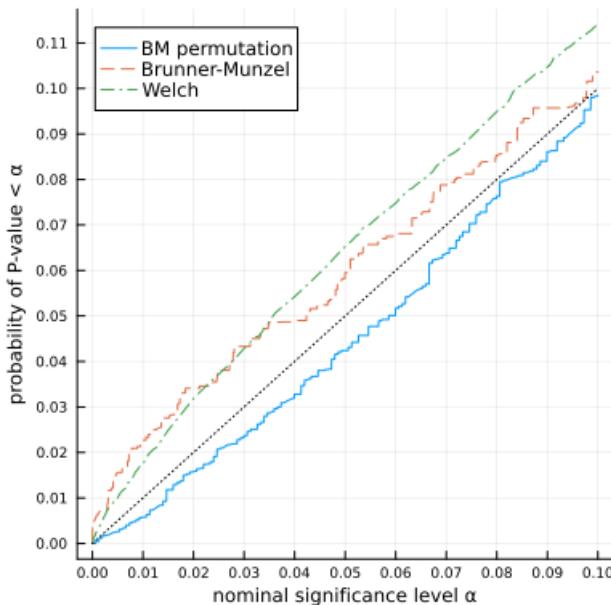
Out[114]:  
X: Uniform(a=-1.0, b=1.0), m=5  
Y: Exponential(θ=0.57735)+(a, Δμ), n=10  
a=-0.5371, Δμ=-0.5774



```
In [115]: 1 plot_pvals_with_perm(
2     distx = Uniform(-1, 1), disty = Exponential(0.5773502691896257),
3     m = 10, n = 5, L = 10^4)

4.875233 seconds (15.70 M allocations: 4.443 GiB, 15.06% gc time)
0.001710 seconds (5.34 k allocations: 1.832 MiB)
0.001474 seconds (6.12 k allocations: 2.059 MiB)
a = -0.5370568188698567
Δμ = -0.5773502691896257
```

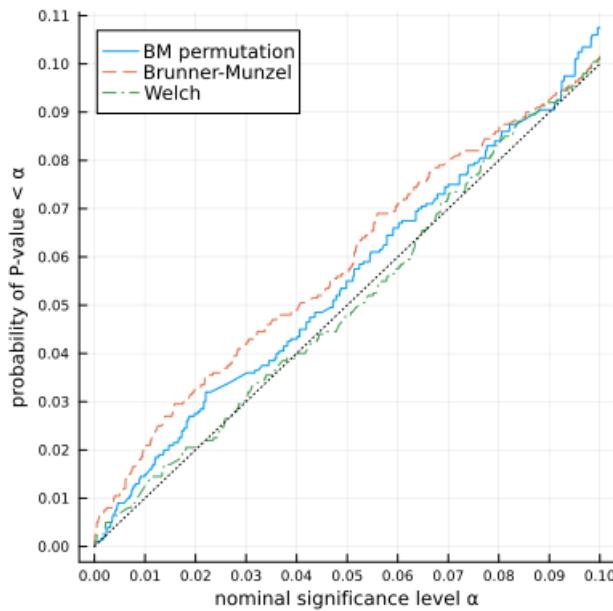
Out[115]:  
X: Uniform(a=-1.0, b=1.0), m=10  
Y: Exponential(θ=0.57735)+(a, Δμ), n=5  
a=-0.5371, Δμ=-0.5774



```
In [116]: 1 plot_pvals_with_perm(
2     distx = Uniform(-1, 1), disty = Exponential(0.5773502691896257),
3     m = 10, n = 10, L = 2000)
```

66.621568 seconds (203.34 M allocations: 57.586 GiB, 17.00% gc time)  
 0.000423 seconds (1.14 k allocations: 380.953 KiB)  
 0.000337 seconds (1.30 k allocations: 432.703 KiB)  
 $a = -0.5370568188698567$   
 $\Delta\mu = -0.5773502691896257$

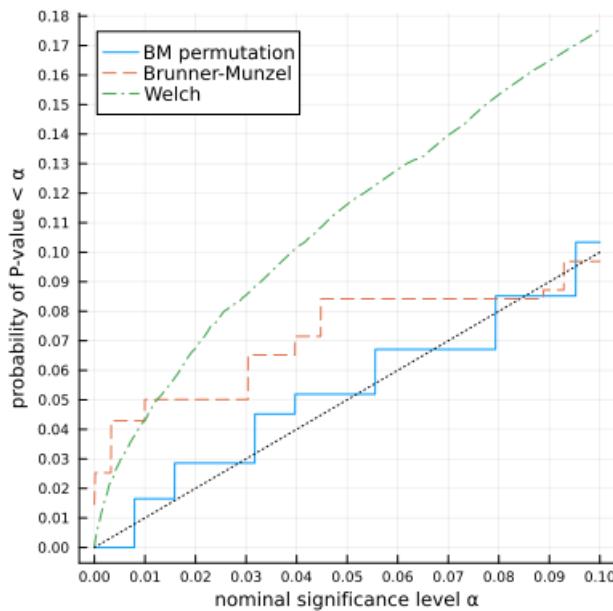
Out[116]: X: Uniform( $a=-1.0, b=1.0$ ),  $m=10$   
 Y: Exponential( $\theta=0.57735$ ) $+(a, \Delta\mu)$ ,  $n=10$   
 $a=-0.5371, \Delta\mu=-0.5774$



```
In [117]: 1 plot_pvals_with_perm(
2     distx = LogNormal(), disty = LogNormal(1), m = 5, n = 5, L = 10^4)
```

0.356817 seconds (1.23 M allocations: 351.341 MiB, 154.41% compilation time)  
 0.002541 seconds (4.76 k allocations: 1.663 MiB)  
 0.002543 seconds (6.17 k allocations: 2.077 MiB)  
 $a = -1.4744426128871542$   
 $\Delta\mu = -2.8329677996379363$

Out[117]: X: LogNormal( $\mu=0.0, \sigma=1.0$ ),  $m=5$   
 Y: LogNormal( $\mu=1.0, \sigma=1.0$ ) $+(a, \Delta\mu)$ ,  $n=5$   
 $a=-1.4744, \Delta\mu=-2.833$

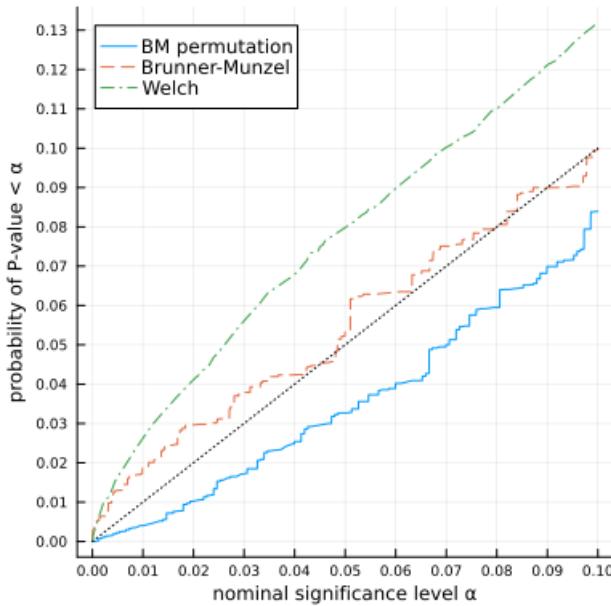


```
In [118]: 1 plot_pvals_with_perm(
2     distx = LogNormal(), disty = LogNormal(1), m = 5, n = 10, L = 10^4)
```

4.881429 seconds (15.70 M allocations: 4.442 GiB, 17.95% gc time)  
 0.026256 seconds (5.26 k allocations: 1.808 MiB)  
 0.015550 seconds (5.63 k allocations: 1.921 MiB)  
 $a = -1.4744426128871542$   
 $\Delta\mu = -2.8329677996379363$

Out[118]:

X:  $\text{LogNormal}(\mu=0.0, \sigma=1.0)$ , m=5  
 Y:  $\text{LogNormal}(\mu=1.0, \sigma=1.0) + (a, \Delta\mu)$ , n=10  
 $a=-1.4744, \Delta\mu=-2.833$

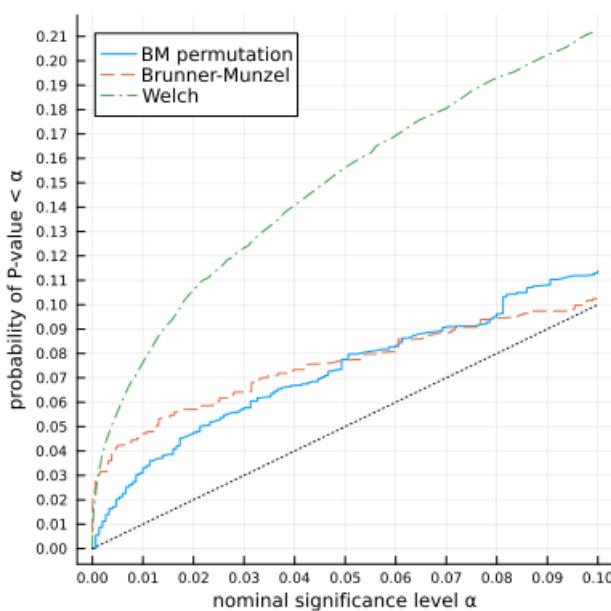


```
In [119]: 1 plot_pvals_with_perm(
2     distx = LogNormal(), disty = LogNormal(1), m = 10, n = 5, L = 10^4)
```

4.735019 seconds (15.70 M allocations: 4.443 GiB, 15.23% gc time)  
 0.013115 seconds (5.25 k allocations: 1.805 MiB)  
 0.015421 seconds (6.43 k allocations: 2.153 MiB)  
 $a = -1.4744426128871542$   
 $\Delta\mu = -2.8329677996379363$

Out[119]:

X:  $\text{LogNormal}(\mu=0.0, \sigma=1.0)$ , m=10  
 Y:  $\text{LogNormal}(\mu=1.0, \sigma=1.0) + (a, \Delta\mu)$ , n=5  
 $a=-1.4744, \Delta\mu=-2.833$

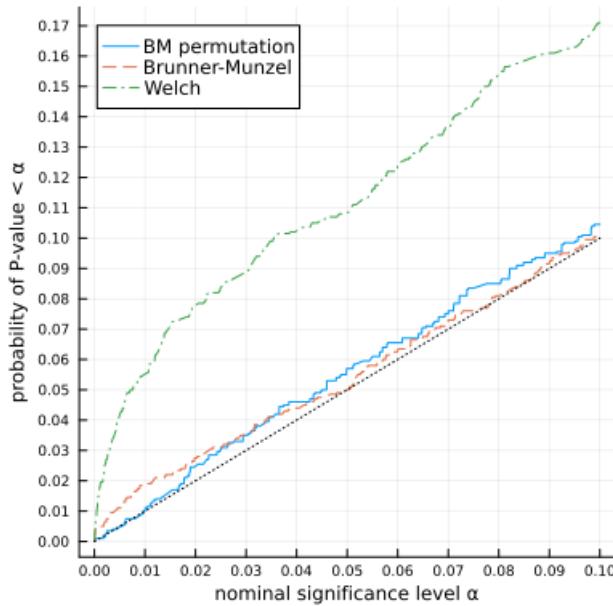


```
In [120]: 1 plot_pvals_with_perm(  
2     distx = LogNormal(0), disty = LogNormal(1), m = 10, n = 10, L = 2000)
```

```
68.285465 seconds (203.34 M allocations: 57.586 GiB, 16.88% gc time)  
0.000460 seconds (1.13 k allocations: 377.789 KiB)  
0.000395 seconds (1.28 k allocations: 426.453 KiB)  
a = -1.4744426128871542  
 $\Delta\mu$  = -2.8329677996379363
```

Out[120]:

X:  $\text{LogNormal}(\mu=0.0, \sigma=1.0)$ , m=10  
Y:  $\text{LogNormal}(\mu=1.0, \sigma=1.0) + (a, \Delta\mu)$ , n=10  
 $a=-1.4744, \Delta\mu=-2.833$



```
In [ ]: 1
```