

P値と信頼区間

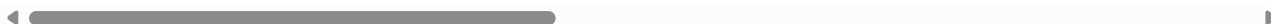
- 黒木玄
- 2025-05-27
- このノートブックの内容の解説は[手書きのノート](#)
<https://github.com/genkuroki/Statistics/blob/master/2022/handwritten/%E6%95%B0%E7%90%86%E7%B5%B1%E8%A8%8F>
にある。
- このノートブックは[Google Colabで実行できる](#)
https://colab.research.google.com/github/genkuroki/Statistics/blob/master/2022/handwritten/08_P-value_and_confidence_interval.ipynb).

P値と信頼区間に関する現代的な理解については

- 音声概要集 (<https://genkuroki.github.io/audio/>)

の01番から06番の6つの音声概要を聴いて欲しい。全部1倍速で聴いても1時間弱。

言葉ではなく、数学的なP値と信頼区間の理解については、このノートブックのコードが役に立つだろう。



```

In [1]: 1 # Google Colabと自分のパソコンの両方で使えるようにするための工夫
2
3 import Pkg
4
5 """すでにPkg.add済みのパッケージのリスト (高速化のために用意)"""
6 _packages_added = [info.name for (uuid, info) in Pkg.dependencies() if info.is_direct_dep]
7
8 """_packages_added内にはパッケージをPkg.addする"""
9 add_pkg_if_not_added_yet(pkg) = if !(pkg in _packages_added)
10     println(stderr, "# $(pkg).jl is not added yet, so let's add it.")
11     Pkg.add(pkg)
12 end
13
14 """expr::Exprからusing内の`.`を含まないモジュール名を抽出"""
15 function find_using_pkgs(expr::Expr)
16     pkgs = String[]
17     function traverse(expr::Expr)
18         if expr.head == :using
19             for arg in expr.args
20                 if arg.head == :. && length(arg.args) == 1
21                     push!(pkgs, string(arg.args[1]))
22                 elseif arg.head == :(:) && length(arg.args[1].args) == 1
23                     push!(pkgs, string(arg.args[1].args[1]))
24                 end
25             end
26         else
27             for arg in expr.args arg isa Expr && traverse(arg) end
28         end
29     end
30     traverse(expr)
31     pkgs
32 end
33
34 """必要そうなPkg.addを追加するマクロ"""
35 macro autoadd(expr)
36     pkgs = find_using_pkgs(expr)
37     :(add_pkg_if_not_added_yet.($(pkgs))); $expr
38 end
39
40 # 以下は黒木玄がよく使っているパッケージ達
41 # 例えばQuadGKパッケージ(数値積分のパッケージ)の使い方は
42 # QuadGK.jl をインターネットで検索すれば得られる.
43
44 ENV["LINES"], ENV["COLUMNS"] = 100, 100
45 using LinearAlgebra
46 using Printf
47 using Random
48 Random.seed!(4649373)
49
50 @autoadd begin
51     using Distributions
52     using StatsPlots
53     default(fmt=:png, legendfontsize=12)
54     #using BenchmarkTools
55     #using Optim
56     #using QuadGK
57     #using RDatasets
58     #using Roots
59     #using StatsBase
60     #using StatsFuns
61     #using SpecialFunctions
62     #using SymPy
63 end

```

```

In [2]: 1 safediv(x, y) = x == 0 ? zero(x/y) : x/y
2       r(x) = round(x; sigdigits=3)

```

Out[2]: r (generic function with 1 method)

```

In [3]: 1 surprisal_value(pval) = -log2(pval)

```

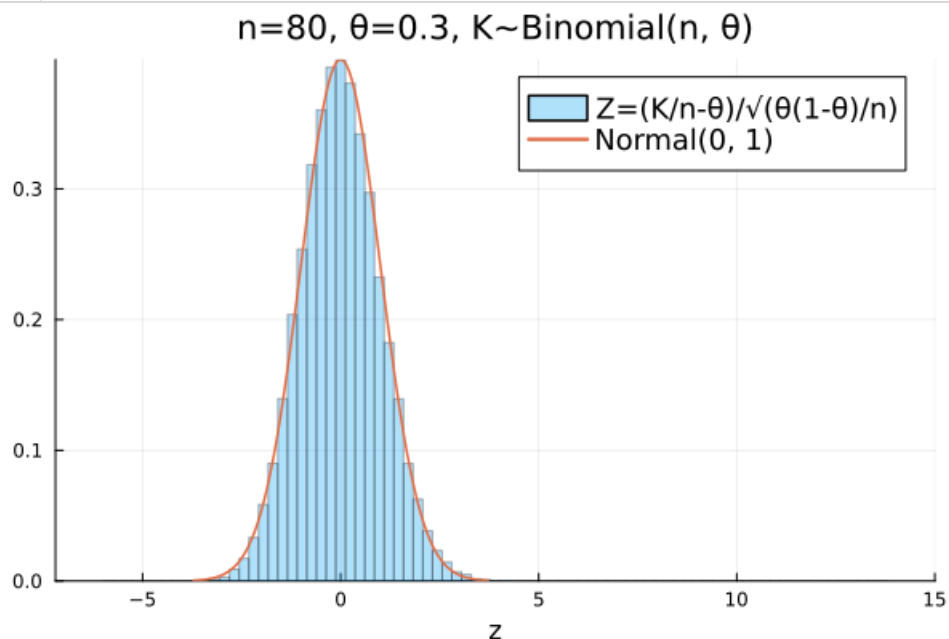
Out[3]: surprisal_value (generic function with 1 method)

```
In [4]: 1 @show surprisal_value(1.0);
2 @show surprisal_value(0.5);
3 @show surprisal_value(0.25);
4 @show surprisal_value(0.125);
5 @show surprisal_value(0.0625);
6 @show surprisal_value(0.03125);
7 @show surprisal_value(0.05) ▷ r;
8 @show surprisal_value(0.01) ▷ r;
9 @show surprisal_value(0.001) ▷ r;
```

```
surprisal_value(1.0) = -0.0
surprisal_value(0.5) = 1.0
surprisal_value(0.25) = 2.0
surprisal_value(0.125) = 3.0
surprisal_value(0.0625) = 4.0
surprisal_value(0.03125) = 5.0
surprisal_value(0.05) ▷ r = 4.32
surprisal_value(0.01) ▷ r = 6.64
surprisal_value(0.001) ▷ r = 9.97
```

```
In [5]: 1 n, theta = 80, 0.3
2 L = 10^5
3 Z = zeros(L)
4 for i in 1:L
5     K = rand(Binomial(n, theta))
6     Z[i] = (K/n - theta) / sqrt(theta*(1-theta)/n)
7 end
8
9 Kbin = -0.5:n+0.5
10 Zbin = @. (Kbin/n - theta) / sqrt(theta*(1-theta)/n)
11 histogram(Z; norm=true, alpha=0.3, bin=Zbin, label="Z=(K/n-θ)/√(θ(1-θ)/n)")
12 plot!(Normal(0, 1); label="Normal(0, 1)", lw=1.5)
13 plot!(xguide="z")
14 title!("n=$n, θ=$theta, K~Binomial(n, θ)")
```

Out[5]:



```

In [6]: 1 theta = 0.5
2 n = 100
3 k = 59
4 @show a = (k/n - theta) / sqrt(theta*(1-theta)/n)
5 @show pval = 2ccdf(Normal(0, 1), abs(a))
6
7 plot(Normal(0, 1), -5, 5; label="Normal(0, 1)", c=:blue)
8 plot!(Normal(0, 1), -5, -abs(a); label="", c=:blue, fillrange=0, fc=:red, fa=0.3)
9 plot!(Normal(0, 1), abs(a), 5; label="", c=:blue, fillrange=0, fc=:red, fa=0.3)
10 vline!([abs(a), -abs(a)]; label="|z|=|a|=$(r(abs(a)))", c=:red, ls=:dot)
11 title!("k=$k, n=$n, θ=$theta, P-value=$(r(pval))")

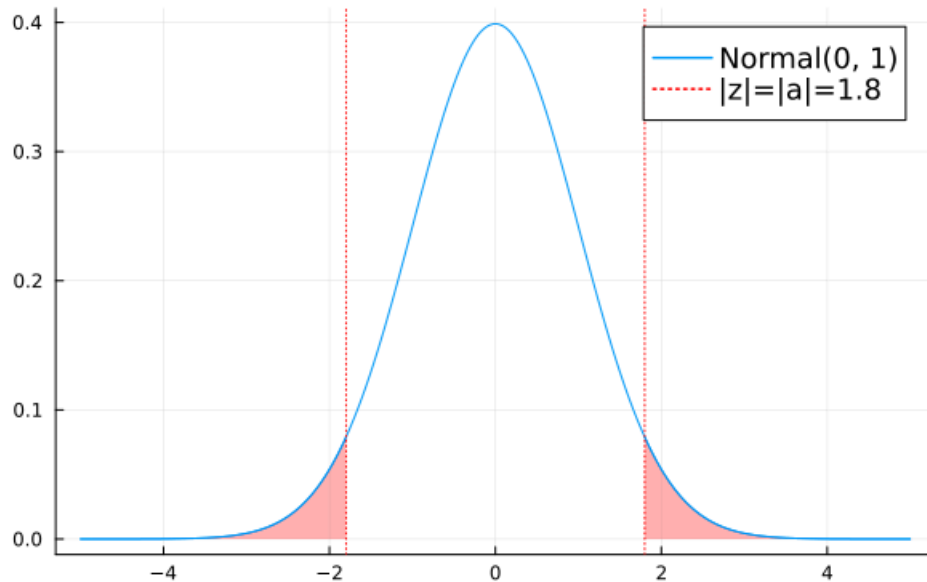
```

```

a = (k / n - theta) / sqrt((theta * (1 - theta)) / n) = 1.7999999999999994
pval = 2 * ccdf(Normal(0, 1), abs(a)) = 0.07186063822585168

```

Out[6]: **k=59, n=100, θ=0.5, P-value=0.0719**

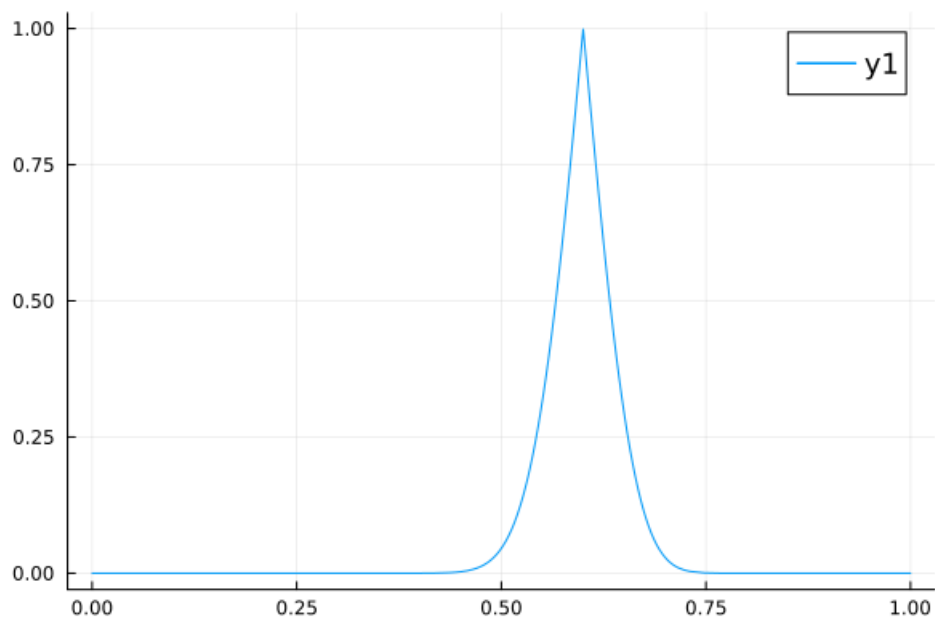


```

In [7]: 1 function pvalue_wilson(k, n, theta)
2         z = safediv(k/n - theta, sqrt(theta*(1-theta)/n))
3         2ccdf(Normal(0, 1), abs(z))
4     end
5
6     # 学習のためにはこういう素朴なプロットでも十分な場合が多い。
7     k, n = 60, 100
8     plot(theta → pvalue_wilson(k, n, theta), 0, 1)

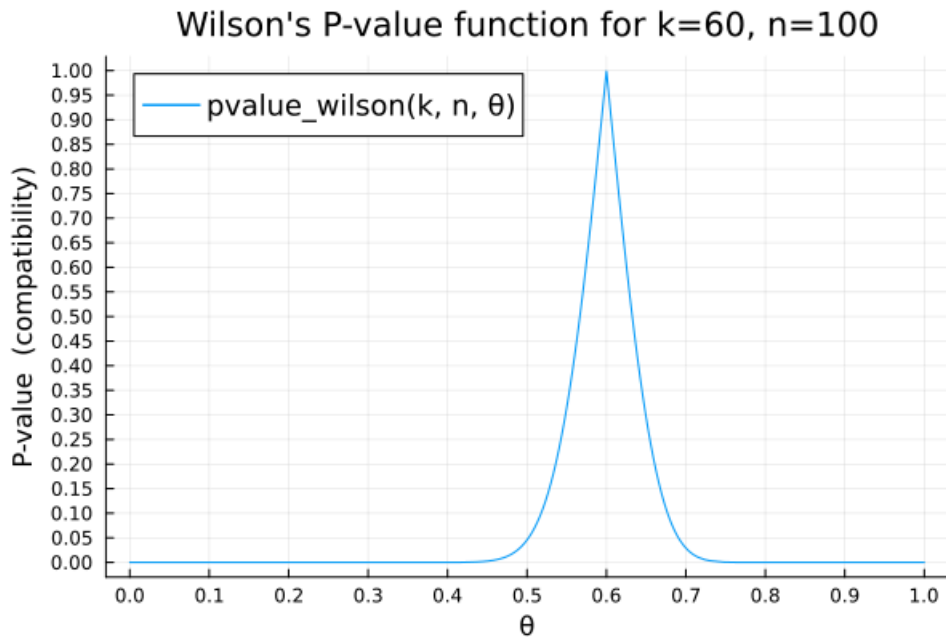
```

Out[7]:



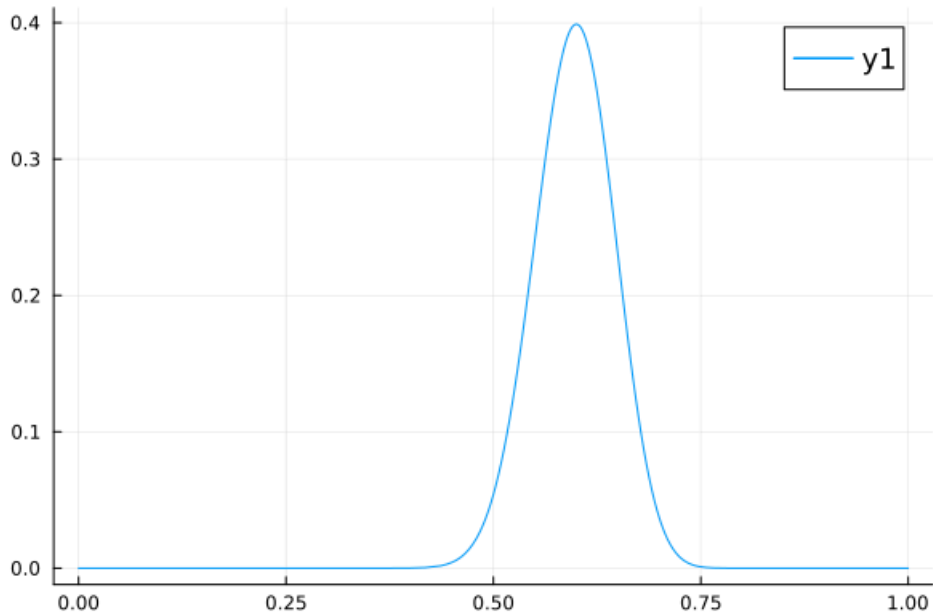
```
In [8]: 1 function plot_pvalue_wilson(k, n; c=1, kwargs...)
2         plot(theta → pvalue_wilson(k, n, theta), 0, 1; label="pvalue_wilson(k, n, θ)", c)
3         plot!(xtick=0:0.1:1, ytick=0:0.05:1)
4         plot!(xguide="θ", yguide="P-value (compatibility)")
5         title!("Wilson's P-value function for k=$k, n=$n")
6         plot!(; kwargs...)
7     end
8
9     plot_pvalue_wilson(60, 100; legend=:topleft)
```

Out[8]:



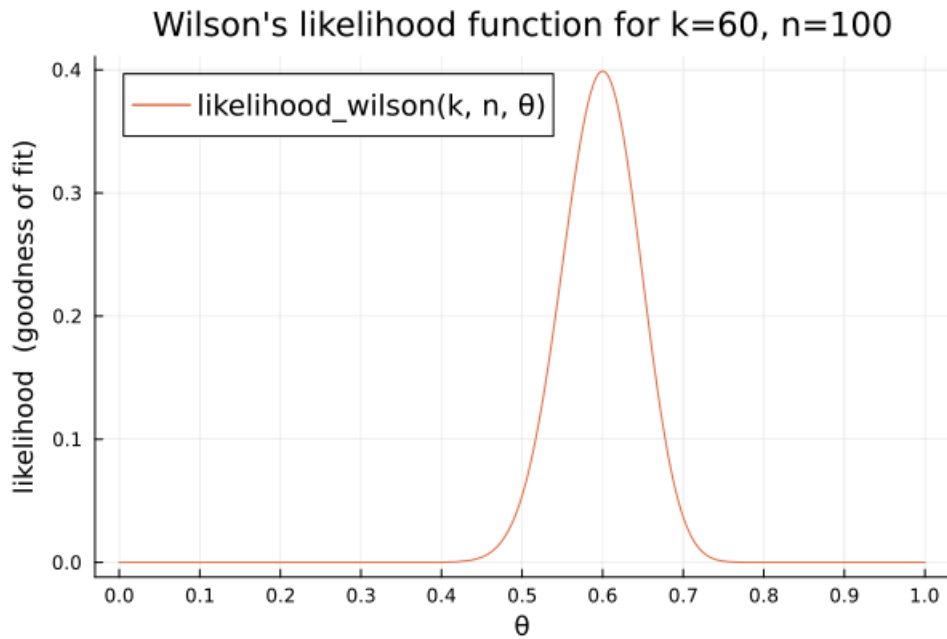
```
In [9]: 1 function likelihood_wilson(k, n, theta)
2         z = safediv(k/n - theta, sqrt(theta*(1-theta)/n))
3         pdf(Normal(0, 1), abs(z))
4     end
5
6     # 学習のためにはこういう素朴なプロットでも十分な場合が多い。
7     k, n = 60, 100
8     plot(theta → likelihood_wilson(k, n, theta), 0, 1)
```

Out[9]:



```
In [10]: 1 function plot_likelihood_wilson(k, n; c=2, kwargs...)
2         plot(theta → likelihood_wilson(k, n, theta), 0, 1; label="likelihood_wilson(k, n, θ)")
3         plot!(xtick=0:0.1:1)
4         plot!(xguide="θ", yguide="likelihood (goodness of fit)")
5         title!("Wilson's likelihood function for k=$k, n=$n")
6         plot!(; kwargs...)
7     end
8
9     plot_likelihood_wilson(60, 100; legend=:topleft)
```

Out[10]:

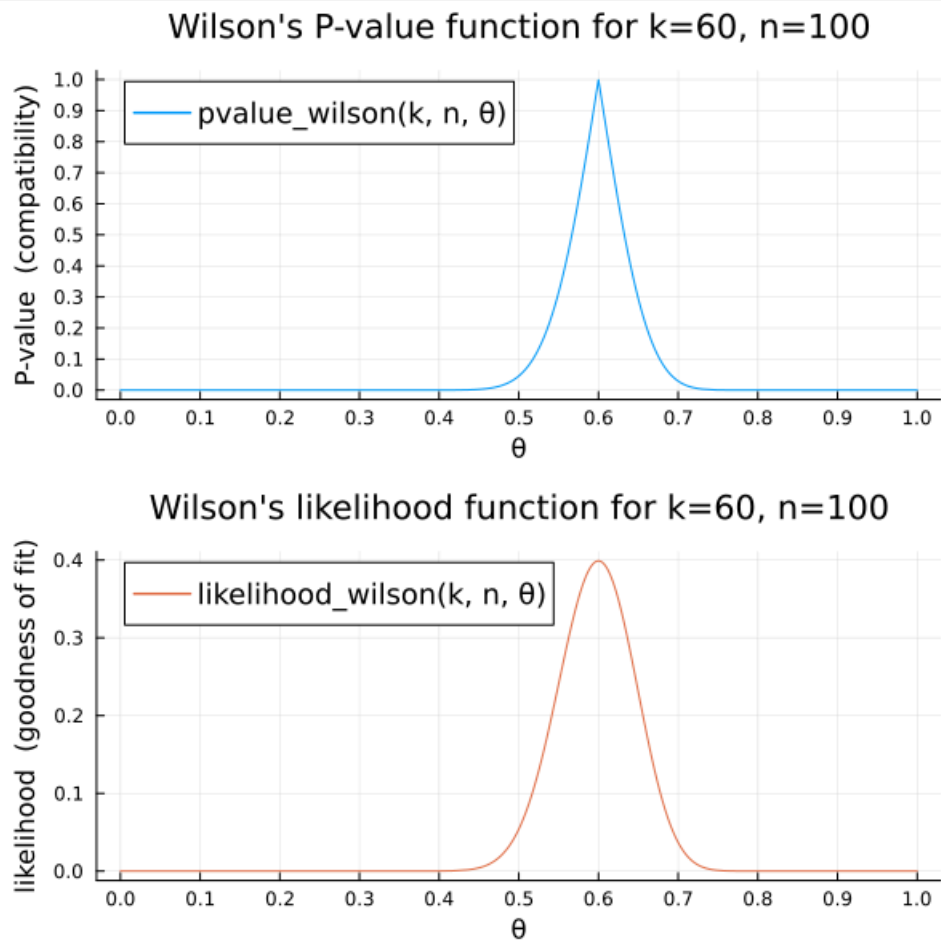


```
In [11]: 1 function plot_pvalue_likelihood_wilson(k, n; kwargs...)
2         P = plot_pvalue_wilson(k, n; kwargs..., ytick=0:0.1:1)
3         Q = plot_likelihood_wilson(k, n; kwargs...)
4         plot(P, Q; size=(600, 600), layout=(2, 1))
5     end
```

Out[11]: plot_pvalue_likelihood_wilson (generic function with 1 method)

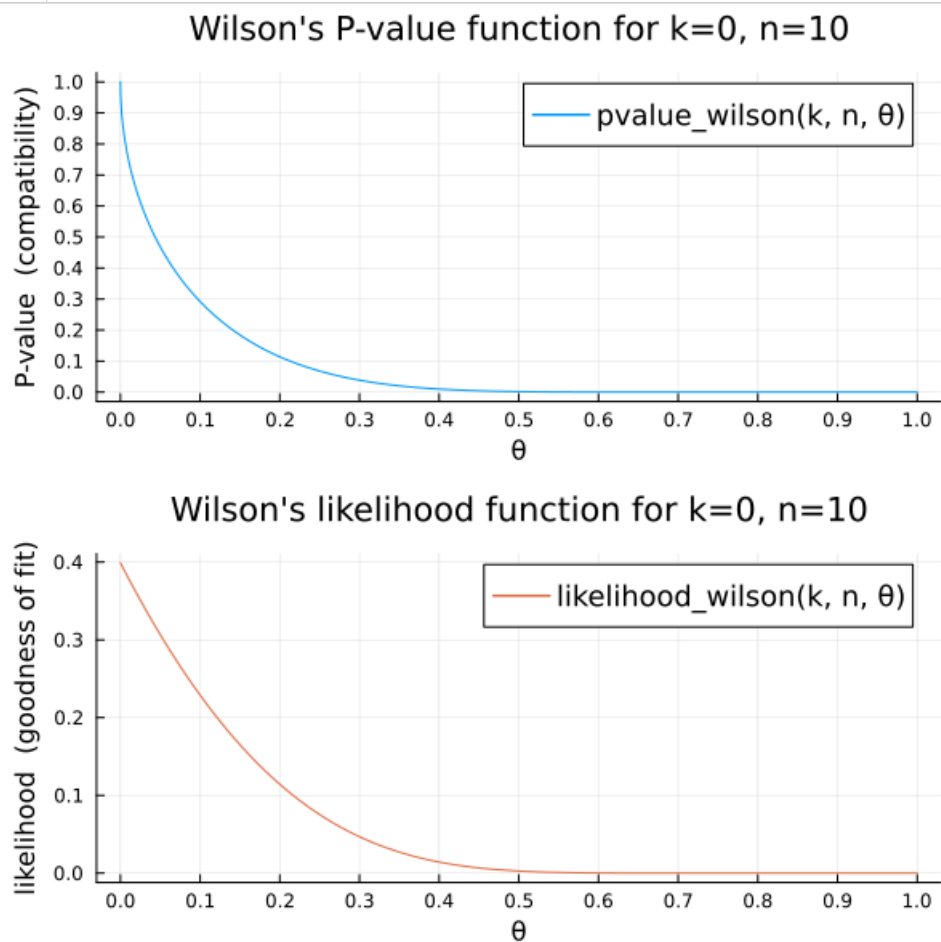
```
In [12]: 1 plot_pvalue_likelihood_wilson(60, 100; legend=:topleft)
```

Out[12]:



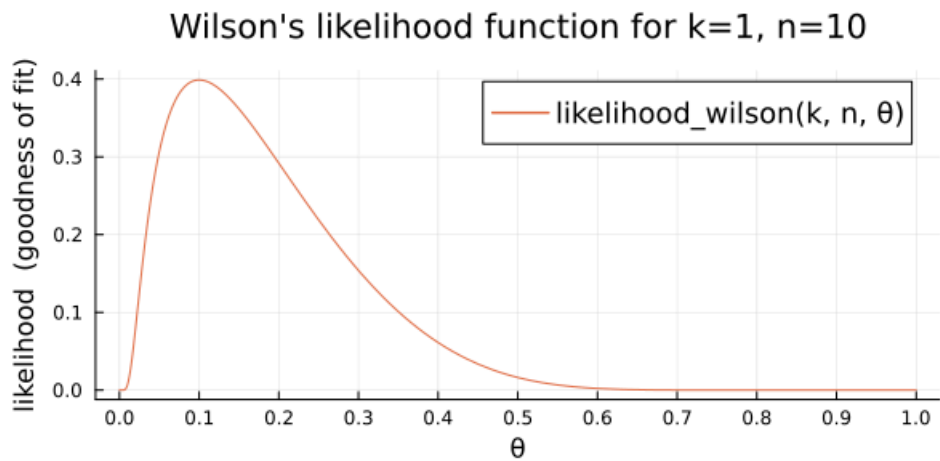
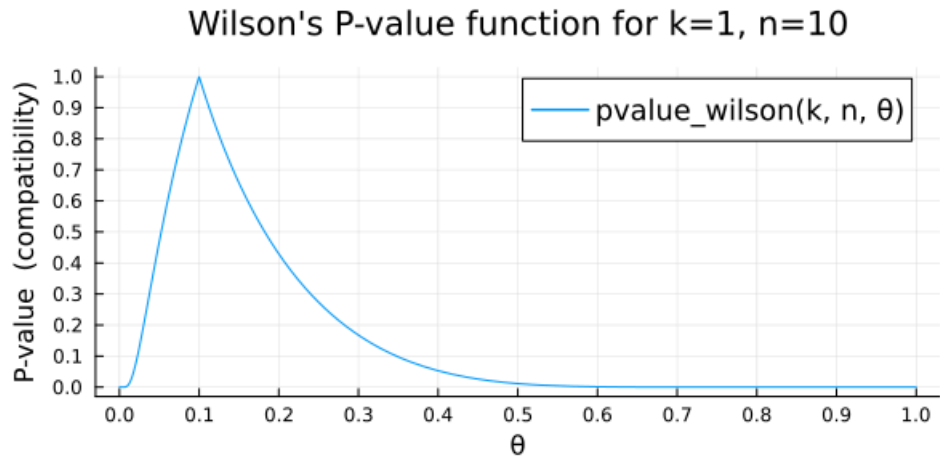
```
In [13]: 1 plot_pvalue_likelihood_wilson(0, 10)
```

Out[13]:



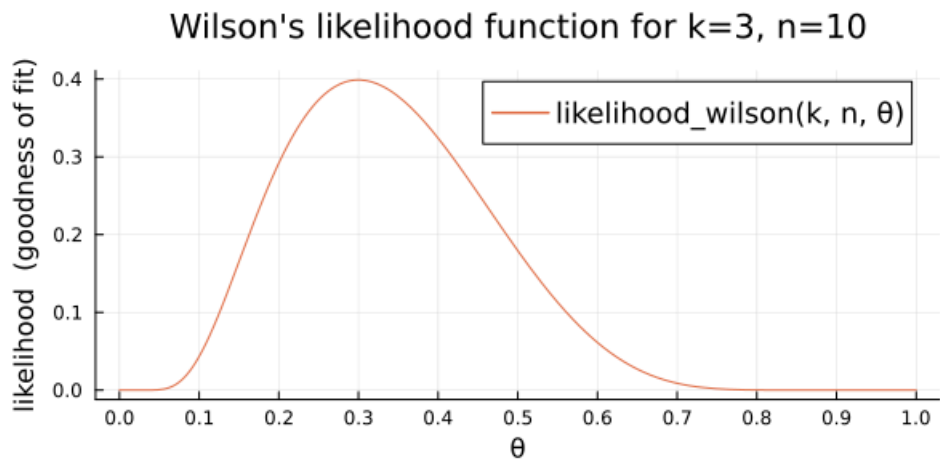
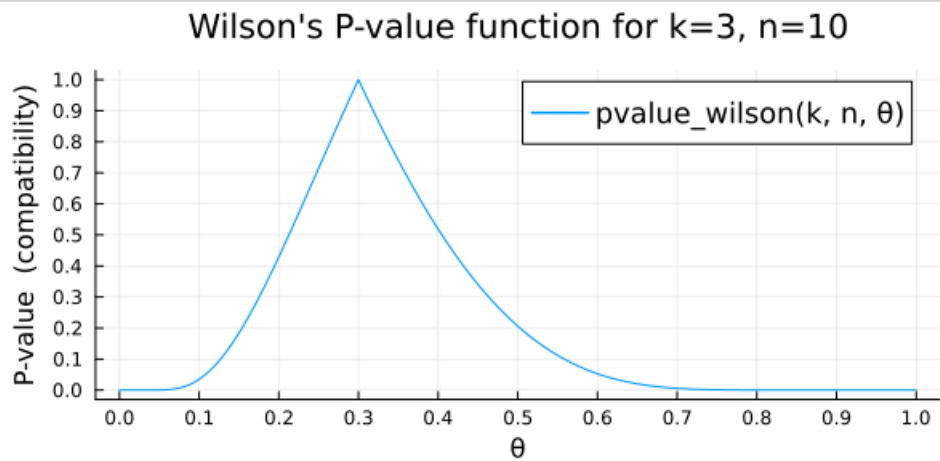
```
In [14]: 1 plot_pvalue_likelihood_wilson(1, 10)
```

Out[14]:



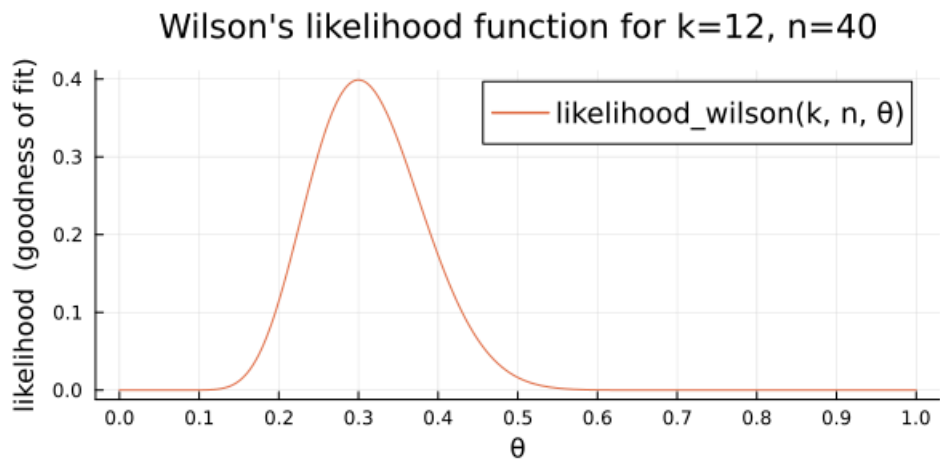
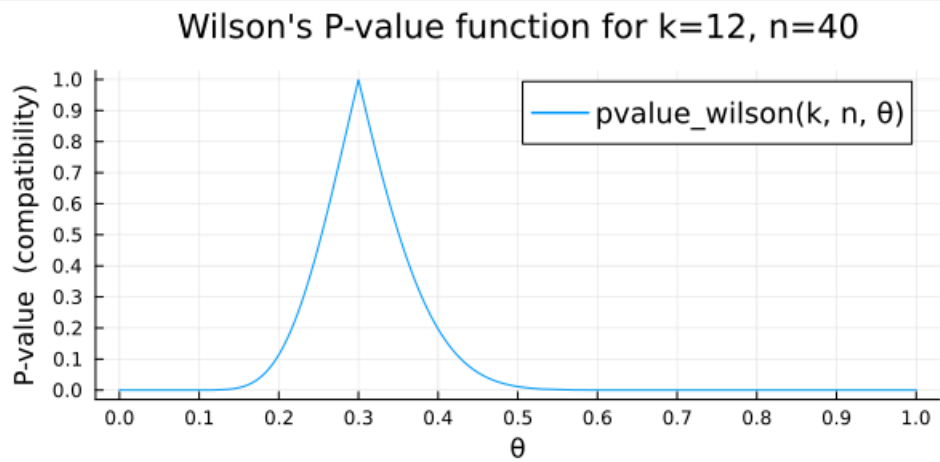
```
In [15]: 1 plot_pvalue_likelihood_wilson(3, 10)
```

Out[15]:



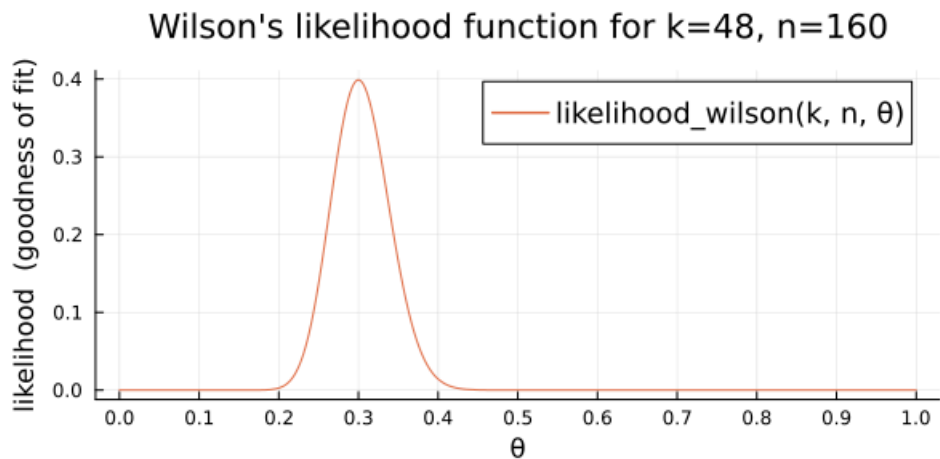
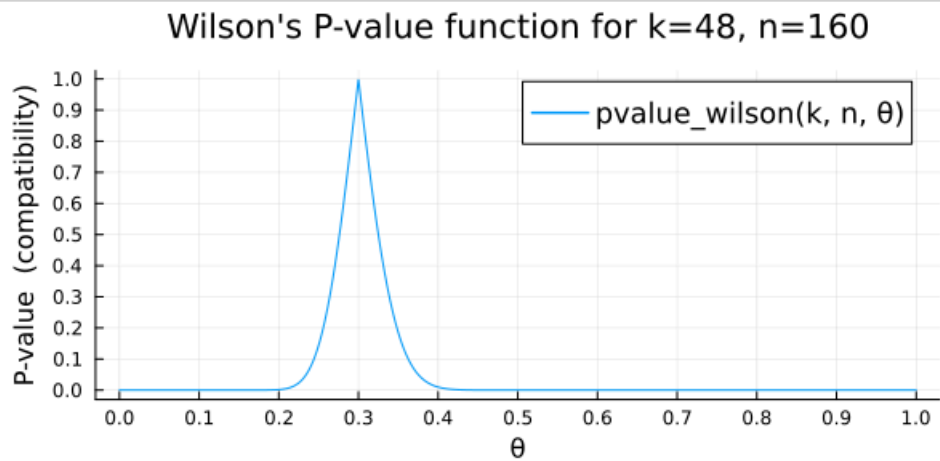

```
In [16]: 1 plot_pvalue_likelihood_wilson(12, 40)
```

Out[16]:



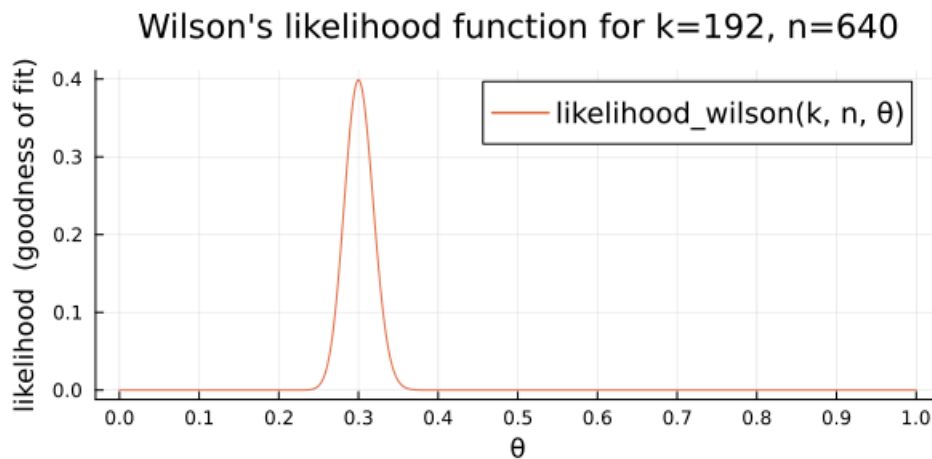
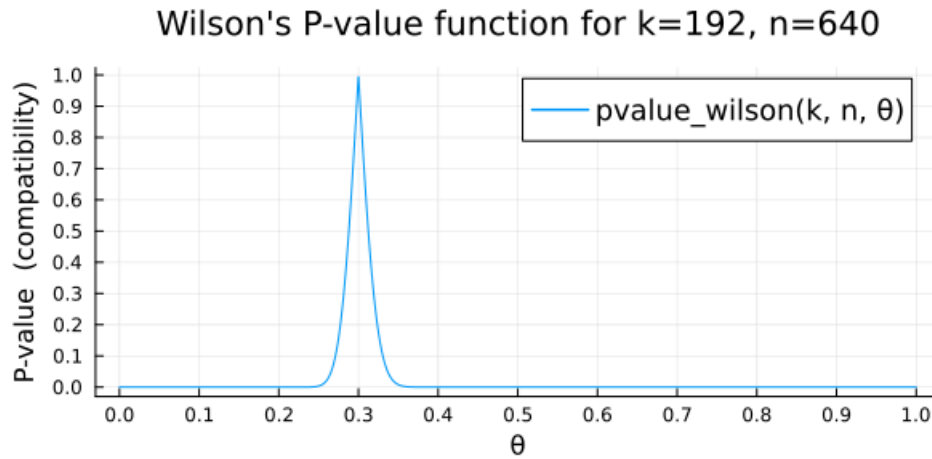
```
In [17]: 1 plot_pvalue_likelihood_wilson(48, 160)
```

Out[17]:



```
In [18]: 1 plot_pvalue_likelihood_wilson(192, 640)
```

Out[18]:



```
In [19]: 1 @show quantile(Normal(0, 1), 0.05/2);  
2 @show quantile(Normal(0, 1), 0.01/2);  
3 @show quantile(Normal(0, 1), 0.001/2);
```

```
quantile(Normal(0, 1), 0.05 / 2) = 1.9599639845400592  
quantile(Normal(0, 1), 0.01 / 2) = 2.5758293035489053  
quantile(Normal(0, 1), 0.001 / 2) = 3.2905267314919
```

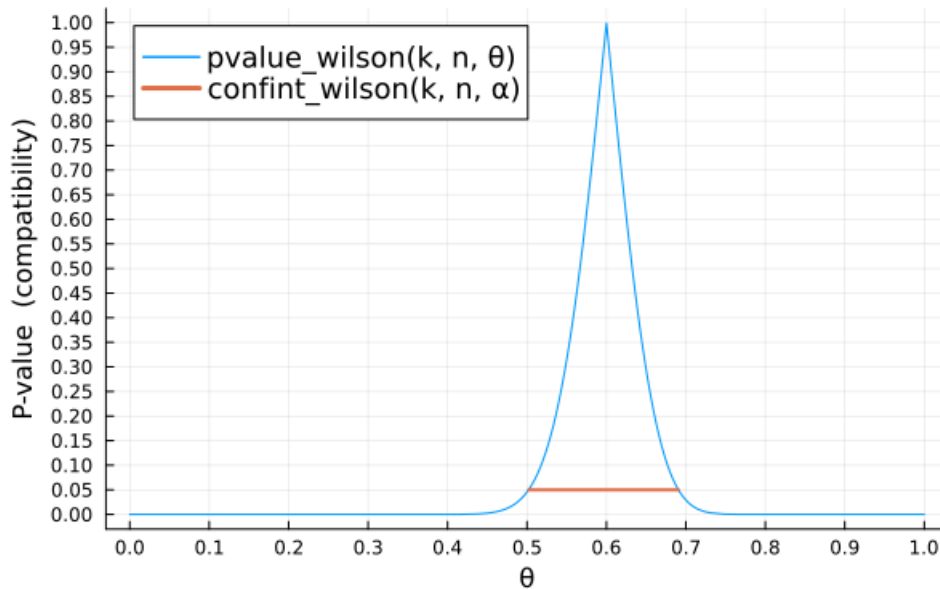
```
In [20]: 1 function confint_wilson(k, n, alpha=0.05)  
2     c = quantile(Normal(0, 1), alpha/2)  
3     thetahat = k/n  
4     A, B, C = 1+c^2/n, thetahat+c^2/(2n), thetahat^2  
5     sqrtD = sqrt(B^2 - A*C)  
6     [(B - sqrtD)/A, (B + sqrtD)/A]  
7 end  
8  
9 function plot_pvalue_confint_wilson(k, n, alpha=0.05; kwargs...)  
10    ci = confint_wilson(k, n, alpha)  
11    println("confint_wilson($k, $n, $alpha) ≈ ", r.(ci))  
12    plot_pvalue_wilson(k, n; c=1)  
13    plot!(ci, fill(alpha, 2); label="confint_wilson(k, n,  $\alpha$ )", lw=2, c=2)  
14    title!("k=$k, n=$n,  $\alpha$ =$alpha")  
15    plot!(; kwargs...)  
16 end
```

Out[20]: plot_pvalue_confint_wilson (generic function with 2 methods)

```
In [21]: 1 # 100(1-α)% 信頼区間はP値関数の高さ α での切断に過ぎない。
2 # 以下は α = 5% の場合である。
3
4 plot_pvalue_confint_wilson(60, 100; legend=:topleft, ytick=0:0.05:1)
```

confint_wilson(60, 100, 0.05) ≈ [0.502, 0.691]

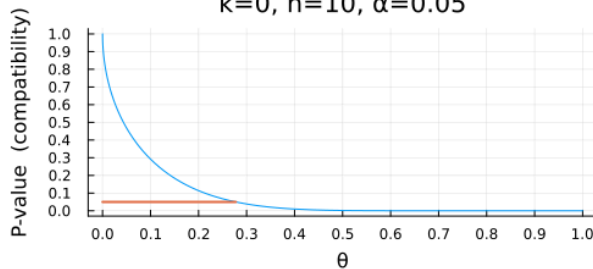
Out[21]: k=60, n=100, α=0.05



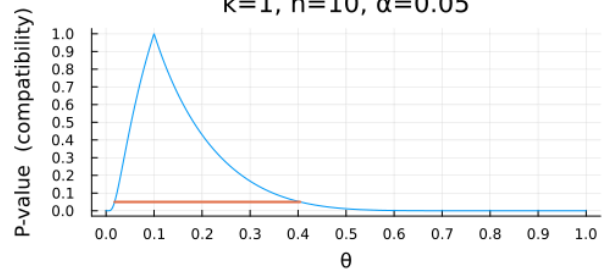
```
In [22]: 1 PP = []
2 for k in 0:5
3     P = plot_pvalue_confint_wilson(k, 10; ytick=0:0.1:1, legend=false)
4     push!(PP, P)
5 end
6 plot(PP...; size=(1000, 700), layout=(3, 2), leftmargin=4Plots.mm)
```

confint_wilson(0, 10, 0.05) ≈ [0.0, 0.278]
 confint_wilson(1, 10, 0.05) ≈ [0.0179, 0.404]
 confint_wilson(2, 10, 0.05) ≈ [0.0567, 0.51]
 confint_wilson(3, 10, 0.05) ≈ [0.108, 0.603]
 confint_wilson(4, 10, 0.05) ≈ [0.168, 0.687]
 confint_wilson(5, 10, 0.05) ≈ [0.237, 0.763]

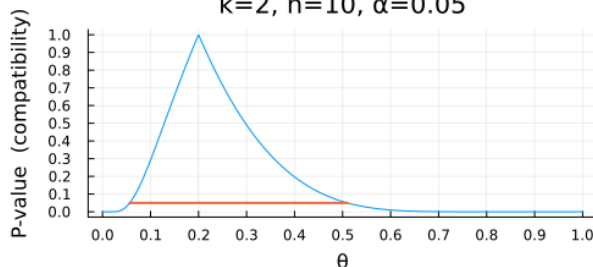
Out[22]: k=0, n=10, α=0.05



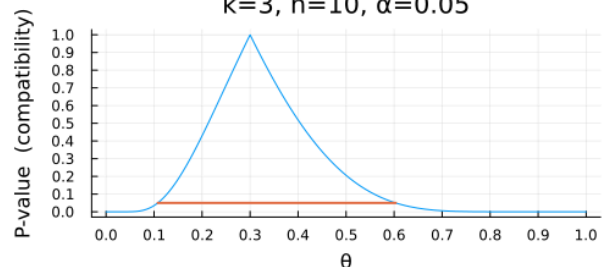
k=1, n=10, α=0.05



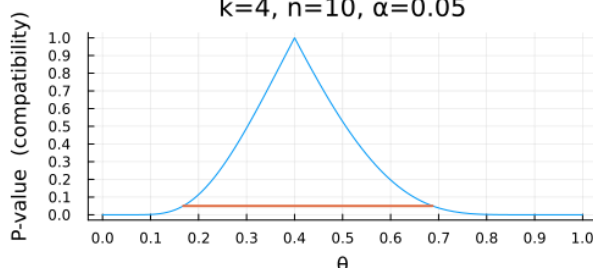
k=2, n=10, α=0.05



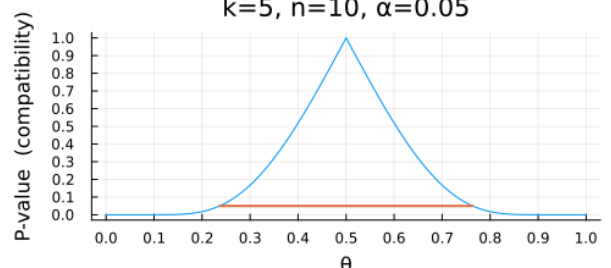
k=3, n=10, α=0.05



k=4, n=10, α=0.05



k=5, n=10, α=0.05

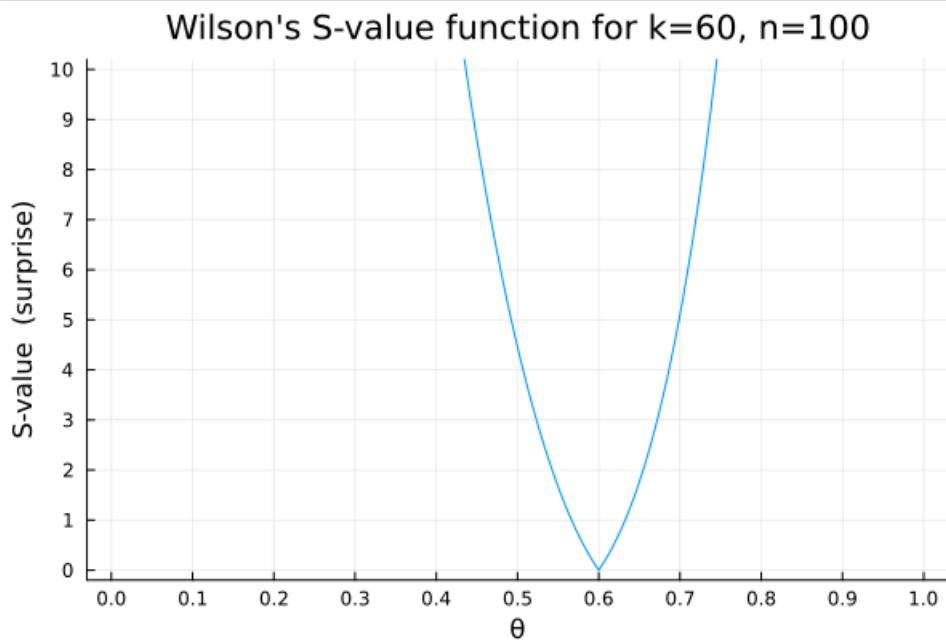


P値はその小ささが重要なのだが、P値関数のグラフではP値の小ささを判別し難い。

P値の代わりにS値(surprisal value)をプロットするとその問題が解消される。
(S値を使うことの欠点はS値はP値と違ってよく知られていないことである。)

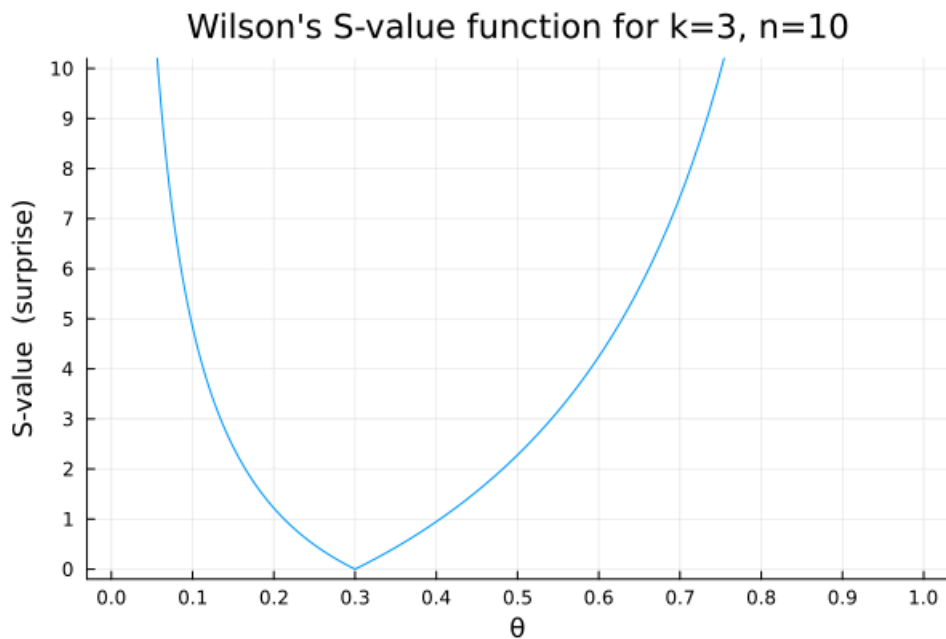
```
In [23]: 1 svalue_wilson(k, n, theta) = surprisal_value(pvalue_wilson(k, n, theta))
2
3 function plot_svalue_wilson(k, n; thetas=range(0, 1, 1001), c=1,
4     ylim=(-0.2, 10.2), ytick=0:10, kwargs...)
5     plot(thetas, theta -> svalue_wilson(k, n, theta); label="", c)
6     plot!(; xtick=0:0.1:1, ytick)
7     plot!(xguide="θ", yguide="S-value (surprise)")
8     plot!(); ylim
9     title!("Wilson's S-value function for k=$k, n=$n")
10    plot!(); kwargs...)
11 end
12
13 plot_svalue_wilson(60, 100)
```

Out[23]:



```
In [24]: 1 plot_svalue_wilson(3, 10)
```

Out[24]:

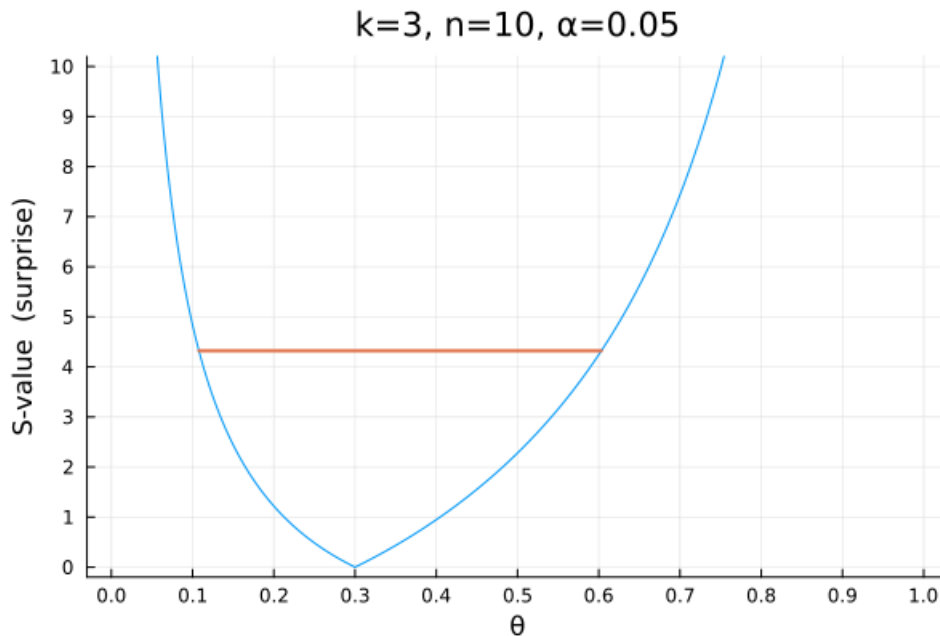


```
In [25]: ► 1 function plot_svalue_confint_wilson(k, n, alpha=0.05; kwargs...)
2           ci = confint_wilson(k, n, alpha)
3           #println("confint_wilson($k, $n, $alpha) ≈ ", r.(ci))
4           plot_svalue_wilson(k, n; c=1)
5           plot!(ci, fill(surprisal_value(alpha), 2); label="", lw=2, c=2)
6           title!("k=$k, n=$n, α=$alpha")
7           plot!(; kwargs...)
8       end
```

Out[25]: plot_svalue_confint_wilson (generic function with 2 methods)

```
In [26]: ► 1 plot_svalue_confint_wilson(3, 10)
```

Out[26]:



```
In [27]: ► 1 function plot_pvalue_svalue_confint_wilson(k, n, alpha=0.05; kwargs...)
2           P = plot_pvalue_confint_wilson(k, n, alpha; kwargs..., ytick=0:0.1:1, legend=false)
3           Q = plot_svalue_confint_wilson(k, n, alpha; kwargs...)
4           plot(P, Q; size=(600, 600), layout=(2, 1))
5       end
```

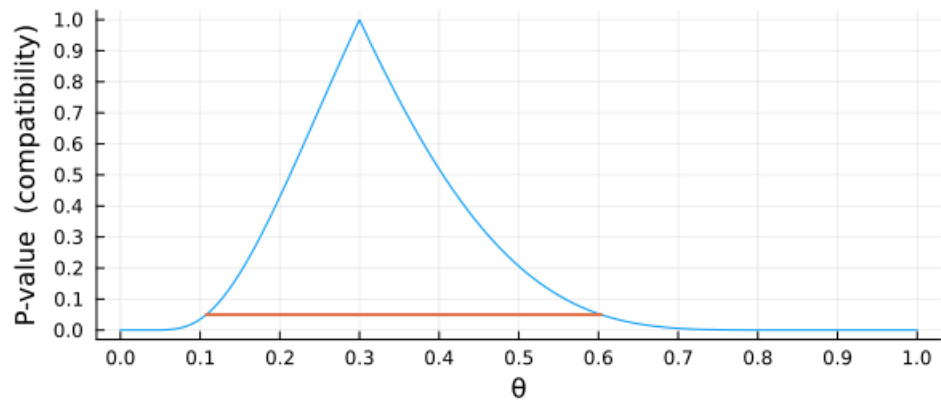
Out[27]: plot_pvalue_svalue_confint_wilson (generic function with 2 methods)

```
In [28]: 1 plot_pvalue_svalue_confint_wilson(3, 10)
```

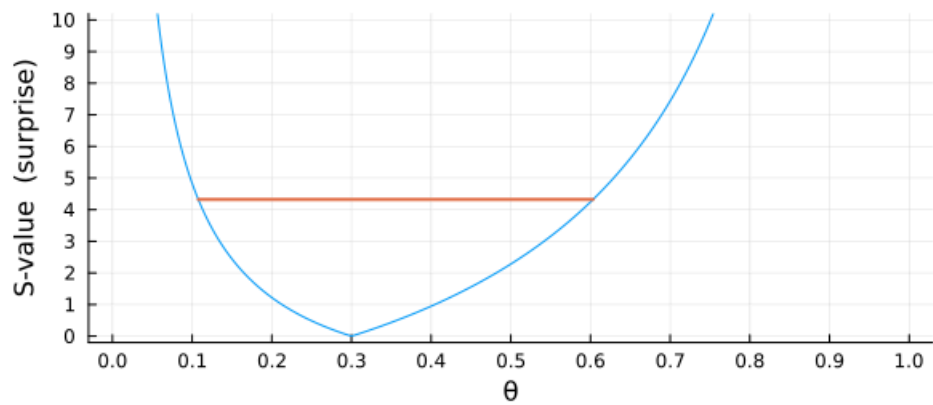
```
confint_wilson(3, 10, 0.05) ≈ [0.108, 0.603]
```

Out[28]:

k=3, n=10, $\alpha=0.05$



k=3, n=10, $\alpha=0.05$

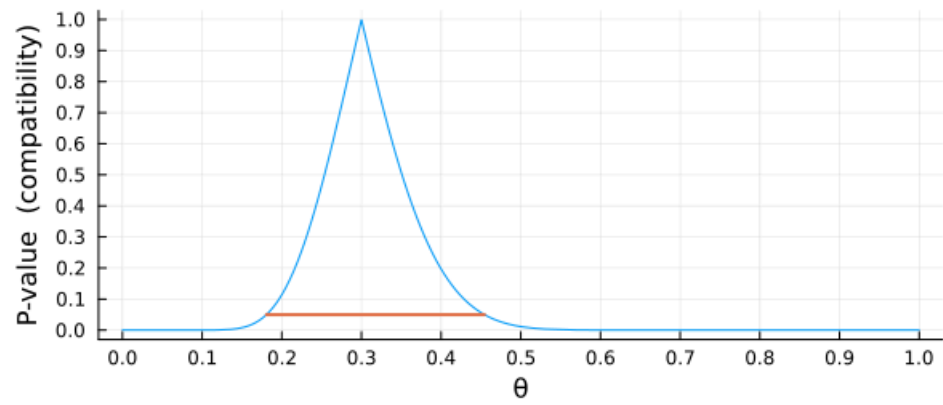


```
In [29]: 1 plot_pvalue_svalue_confint_wilson(12, 40)
```

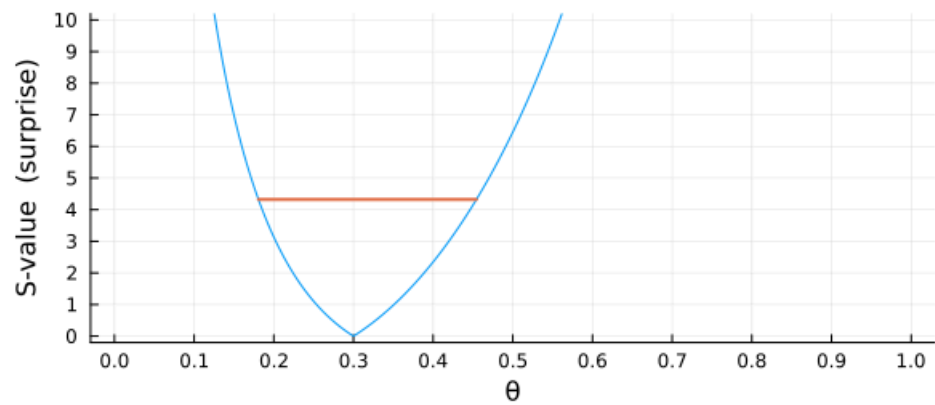
```
confint_wilson(12, 40, 0.05) ≈ [0.181, 0.454]
```

Out[29]:

k=12, n=40, $\alpha=0.05$



k=12, n=40, $\alpha=0.05$

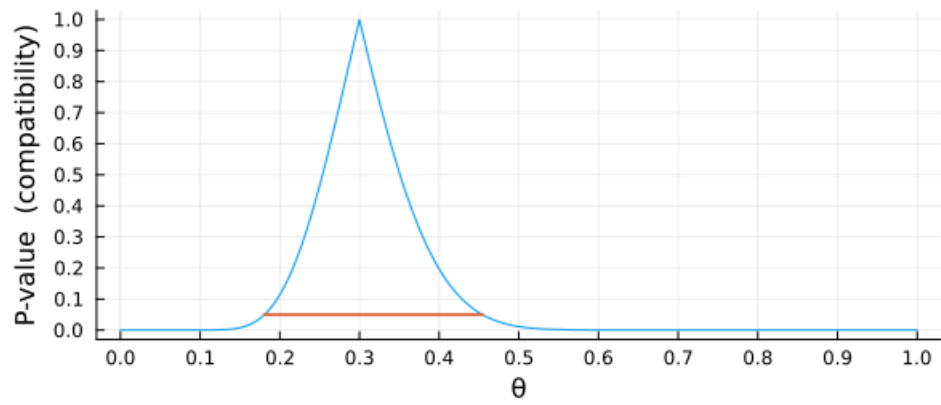


```
In [30]: 1 plot_pvalue_svalue_confint_wilson(12, 40)
```

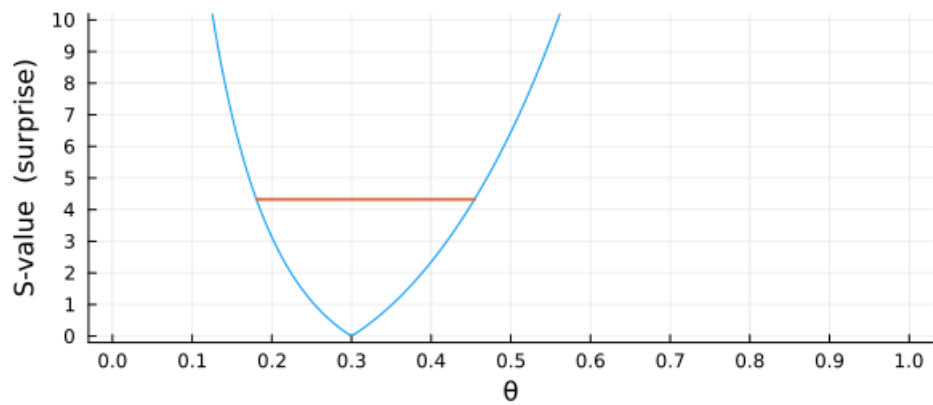
```
confint_wilson(12, 40, 0.05) ≈ [0.181, 0.454]
```

Out[30]:

k=12, n=40, $\alpha=0.05$



k=12, n=40, $\alpha=0.05$

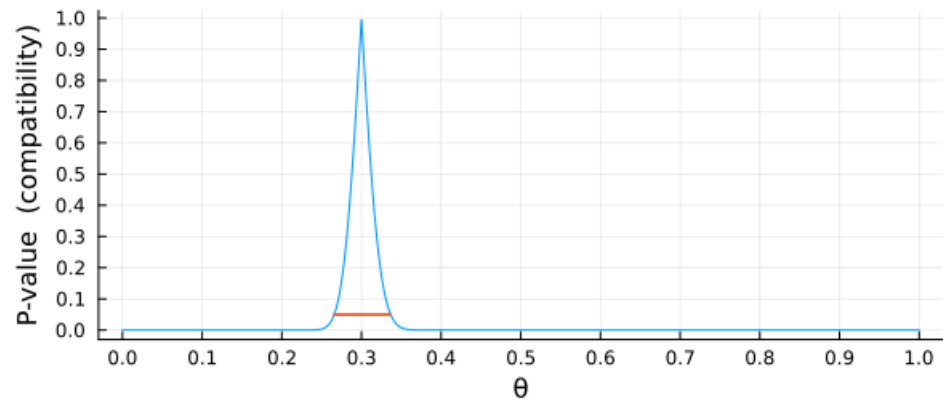


In [31]: 1 plot_pvalue_svalue_confint_wilson(192, 640)

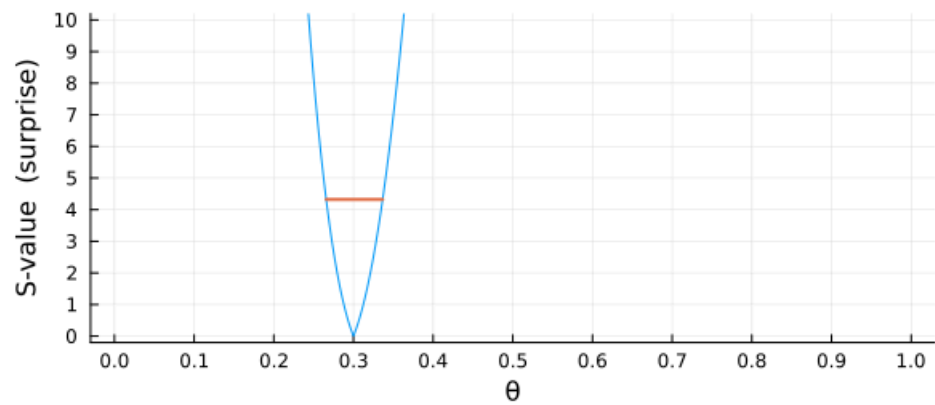
confint_wilson(192, 640, 0.05) \approx [0.266, 0.337]

Out[31]:

k=192, n=640, $\alpha=0.05$



k=192, n=640, $\alpha=0.05$



In []:

1