

# Predicting performance in Psychometric Tests

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## Load packages and dataset

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```

• begin
•   using DataFrames
•   using Empirikos
•   using Plots
•   using PGFPlotsX
•   using LaTeXStrings
•   using MosekTools
•   using JuMP
• end

```

```

• begin
•   pgfplotsx()
•   empty!(PGFPlotsX.CUSTOM_PREAMBLE)
•   push!(PGFPlotsX.CUSTOM_PREAMBLE, raw"\usepackage{amssymb}")
•   push!(PGFPlotsX.CUSTOM_PREAMBLE, raw"\newcommand{\PP}[2][[]
•     {\mathbb{P}_{\#1}\left[\#2\right]}")
•   push!(PGFPlotsX.CUSTOM_PREAMBLE, raw"\newcommand{\EE}[2][[]
•     {\mathbb{E}_{\#1}\left[\#2\right]}")
• end;

```

```
lord_cressie =
```

	x	N1	N2	post_mean	Lower1	Upper1	Lower2	Upper2
<b>1</b>	20	63	2	0.898	0.852	0.952	0.713	0.999
<b>2</b>	19	141	2	0.863	missing	missing	missing	missing
<b>3</b>	18	220	2	0.823	0.78	0.846	0.637	0.936
<b>4</b>	17	319	2	0.773	missing	missing	missing	missing
<b>5</b>	16	424	6	0.716	0.69	0.743	0.588	0.846
<b>6</b>	15	622	4	0.663	missing	missing	missing	missing
<b>7</b>	14	776	8	0.619	0.605	0.646	0.521	0.739
<b>8</b>	13	1001	4	0.583	missing	missing	missing	missing
<b>9</b>	12	1203	9	0.553	0.534	0.564	0.447	0.627
<b>10</b>	11	1443	19	0.526	missing	missing	missing	missing
more								
<b>21</b>	0	2	0	0.28	0.01	0.373	0.0	0.479

```
• lord_cressie = Empirikos.LordCressie.load_table() |> DataFrame
```

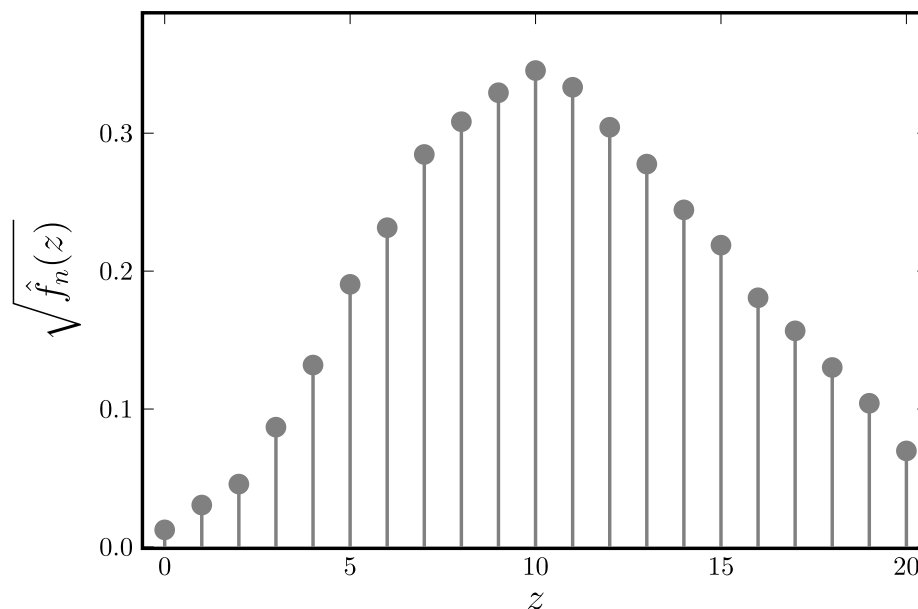
```
• Zs = Empirikos.MultinomialSummary(BinomialSample.(lord_cressie.x, 20),
•   lord_cressie.N1);
•
```

## Plot empirical frequencies (Fig. 1a)

```
empirical_probs =
```

```
[0.000153965, 0.000923788, 0.00207852, 0.00754426, 0.017398, 0.0362587, 0.0535797, 0.08098
```

```
• empirical_probs = weights(Zs)/nobs(Zs)
```



```
lord_cressie_freq =
```

```
• lord_cressie_freq = plot(
•   0:20,
•   sqrt(empirical_probs), seriestype=:sticks, frame=:box,
•   grid=nothing, color=:grey, markershape=:circle,
•   legendfontalign = :left,
•   markerstrokealpha = 0, ylim=(-0.001,sqrt(0.15)),
•   xguide=L"z",yguide=L"\sqrt{\hat{f}_n(z)}",thickness_scaling=1.3,
•   label=nothing, size=(500,350)
• )
```

```
• # savefig(lord_cressie_freq, "lord_cressie_pdf.tikz")
```

## Construct confidence intervals

We first create a discretization object that will combine all counts  $\leq 1$ .

```
discr = Discretizer([( .. 1], 2, 3, 4, 5, 6, 7, 8, 9, more ,[20 .. ]])
```

```
• discr = integer_discretizer(1:20)
```

```
Zs_collapse =
```

```
MultinomialSummary(SortedDict(Z ∈ ( .. 1] | n=20 ⇒ 14, Z=2 | n=20 ⇒ 27, Z=3 | n=20 ⇒
```

```
• Zs_collapse = discr(Zs)
```

```
quiet_mosek =
```

```
OptimizerWithAttributes(Optimizer (generic function with 2 methods), [RawParameter("QUIET
```

```
• quiet_mosek = optimizer_with_attributes(Mosek.Optimizer, "QUIET" => true)
```

```
•
```

```
• G = DiscretePriorClass(range(0.0,stop=1.0,length=300));
```

```
• postmean_targets = Empirikos.PosteriorMean.(BinomialSample.(0:20,20));
```

# $\chi^2 - F$ -localization intervals

```
chisq_floc = ChiSquaredFLocalization(0.05)
```

```
• chisq_floc = Empirikos.ChiSquaredFLocalization(0.05)
```

```
floc_method_chisq =
```

```
EB intervals with F-Localization: ChiSquaredFLocalization{Float64}(0.05)  
   $\mathcal{G}$ : DiscretePriorClass | support = 0.0:0.0033444816053511705:1.0
```

```
• floc_method_chisq = FLocalizationInterval(flocalization = chisq_floc,  
• convexclass=  $\mathcal{G}$ , solver=quiet_mosek)
```

```
chisq_cis =
```

```
[lower = 0.003731, upper = 0.365,  $\alpha$  = 0.05 (PosteriorMean{BinomialSample{Int64, Int64}}{
```

```
• chisq_cis = confint.(floc_method_chisq, postmean_targets, Zs_collapse)
```

```
lower_chisq_ci =
```

```
[0.00373087, 0.106113, 0.235178, 0.305643, 0.340864, 0.364113, 0.387145, 0.410672, 0.43582
```

```
• lower_chisq_ci = getproperty.(chisq_cis, :lower)
```

```
upper_chisq_ci =
```

```
[0.364997, 0.391926, 0.393431, 0.39507, 0.398937, 0.406593, 0.419615, 0.439773, 0.463009,
```

```
• upper_chisq_ci = getproperty.(chisq_cis, :upper)
```

	x	N1	N2	post_mean	Lower1	Upper1	Lower2	Upper2	lower_chisq	u
1	20	63	2	0.898	0.852	0.952	0.713	0.999	0.852604	0
2	19	141	2	0.863	missing	missing	missing	missing	0.813444	0
3	18	220	2	0.823	0.78	0.846	0.637	0.936	0.780462	0
4	17	319	2	0.773	missing	missing	missing	missing	0.73414	0
5	16	424	6	0.716	0.69	0.743	0.588	0.846	0.689779	0
6	15	622	4	0.663	missing	missing	missing	missing	0.647067	0
7	14	776	8	0.619	0.605	0.646	0.521	0.739	0.60518	0
8	13	1001	4	0.583	missing	missing	missing	missing	0.565566	0
9	12	1203	9	0.553	0.534	0.564	0.447	0.627	0.533905	0
10	11	1443	19	0.526	missing	missing	missing	missing	0.507679	0
more										
21	0	2	0	0.28	0.01	0.373	0.0	0.479	0.00373087	0

```

• begin
•     lord_cressie.lower_chisq = reverse(lower_chisq_ci)
•     lord_cressie.upper_chisq = reverse(upper_chisq_ci)
•     lord_cressie
• end

```

## DKW $F$ -localization intervals

```

floc_method_dkw =
EB intervals with F-Localization: DvoretzkyKieferWolfowitz{Float64, Int64}(0.05, 1000)
G: DiscretePriorClass | support = 0.0:0.0033444816053511705:1.0

```

```

• floc_method_dkw = FLocalizationInterval(
•     flocalization = DvoretzkyKieferWolfowitz(0.05),
•     convexclass = G, solver=quiet_mosek)
•

```

```

• dkw_cis = confint.(floc_method_dkw, postmean_targets, Zs_collapse);
•

```

```

lower_dkw_ci =
[0.000191288, 0.026525, 0.114512, 0.221245, 0.299652, 0.344311, 0.375797, 0.40453, 0.42863

```

```

• lower_dkw_ci = getproperty.(dkw_cis, :lower)

```

```

upper_dkw_ci =
[0.418168, 0.440815, 0.440832, 0.440867, 0.441617, 0.443737, 0.449232, 0.45908, 0.4773, 0.

```

```

• upper_dkw_ci = getproperty.(dkw_cis, :upper)

```

# Amari intervals

```

amari_chisq =
AMARI with F-Localization: ChiSquaredFLocalization{Float64}(0.01)
G: DiscretePriorClass | support = 0.0:0.00334444816053511705:1.0

• amari_chisq = Empirikos.AMARI(
•                               convexclass = G,
•                               flocalization = Empirikos.ChiSquaredFLocalization(0.01),
•                               solver=quiet_mosek,
•                               discretizer=discr
•                               )
•

postmean_ci_amari =
[lower = 0.003125, upper = 0.3717,  $\alpha$  = 0.05 (PosteriorMean{BinomialSample{Int64, Int64}}]

• postmean_ci_amari = confint.(amari_chisq, postmean_targets, Zs_collapse)

lower_amari_ci =
[0.00312456, 0.117026, 0.259604, 0.335781, 0.355055, 0.37159, 0.392273, 0.416569, 0.443707]

• lower_amari_ci = getproperty.(postmean_ci_amari, :lower)

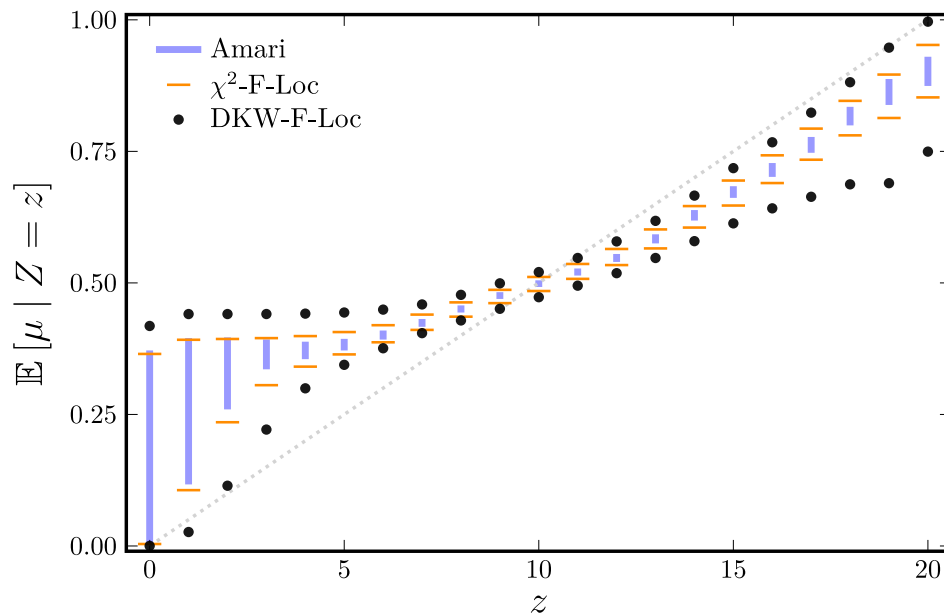
upper_amari_ci =
[0.371684, 0.395092, 0.396436, 0.392493, 0.388285, 0.393683, 0.409529, 0.431459, 0.457264,

• upper_amari_ci = getproperty.(postmean_ci_amari, :upper)

```

## Plot confidence intervals (Fig.2b)

.....



```
postmean_plot =
```

```
• postmean_plot = begin
• plot(0:20, upper_amari_ci, fillrange=lower_amari_ci ,seriestype=:sticks,
•     frame=:box,
•     grid=nothing,
•     xguide = L"z",
•     yguide = L"\mathbb{E}\{\mu \mid Z=z\}",
•     legend = :topleft,
•     linewidth=2,
•     linecolor=:blue,
•     alpha = 0.4,
•     background_color_legend = :transparent,
•     legendfontalign = :left,
•     foreground_color_legend = :transparent, ylim=(-0.01,1.01),
•     thickness_scaling=1.3,
•     label="Amari",
•     size=(500,350))
•
• plot!(0:20, [lower_chisq_ci upper_chisq_ci], seriestype=:scatter, markershape=:hline,
•     label=[L"\chi^2\text{rm}\{-F-Loc\}" nothing], markerstrokecolor=:darkorange,
•     markersize=4.5)
•
• plot!(0:20, [lower_dkw_ci upper_dkw_ci], seriestype=:scatter, markershape=:circle,
•     label=["DKW-F-Loc" nothing], color=:black, alpha=0.9, markersize=2.0,
•     markerstrokealpha=0)
•
• plot!([0;20], [0.0; 1.0], seriestype=:line, linestyle=:dot, label=nothing,
•     color=:lightgrey)
• end

• #savefig(postmean_plot, "lord_cressie_posterior_mean.tikz")
```

