

Odhady hustot na základě reálných dat

Pomocné funkce

```
Clear[α, λ, t, f1, f2, EX]
f1[t_] := α λ (1 - e-tλ)α-1 e-tλ
f2[t_] := α λ (1 + λ t)α-1 Exp[1 - (1 + λ t)α]

Logf[f1, t_] := Log[α λ (1 - e-tλ)α-1] - t λ
Logf[f2, t_] := Log[α λ (1 + λ t)α-1] + 1 - (1 + λ t)α

EX[K_, f_] := EX[K, f] =
  Assuming[{α > 0, λ > 0}, Integrate[t^K f[t], {t, 0, Infinity}]] // FullSimplify

(*Metoda maximální věrohodnosti*)
MLE[f_] := Module[{},
  Clear[α, λ];
  Clear[sectiLogy, ucfc];
  sectiLogy = Table[Logf[f, t], {t, data}];
  ucfc = Total[sectiLogy];
  r = NMaximize[{ucfc, α > 0, λ > 0},
    {{α, 0.1, 20}, {λ, 0.9, 1.2}}, MaxIterations → 100000];
  {α, λ} /. r[[2]]
]

(*Momentová metoda*)
MM[f_] :=
Module[
  {},
  Clear[α, λ];
  mean = Mean[data];
  meanTo2 = Mean[data^2];
  Clear[ex1, ex2];
  ex1[α_, λ_] := EX[1, f];
  ex2[α_, λ_] := EX[2, f];

  res = FindRoot[{ex1[α, λ] == mean, ex2[α, λ] == meanTo2},
    {{α, 5}, {λ, 0.3}}, MaxIterations → 10000];
  {α, λ} /. res
]
```

První datová sada

```

data1 =
  WeatherData["KBOS", "TotalPrecipitation", {{2000, 1, 1}, {2005, 12, 31}}, "Month"];
data = data1["Path"][[All, 2]]

r1 = MLE[f1];
r2 = MM[f1];

{α, λ} = r1
f3p = f1[t];
{α, λ} = r2
f2p = f1[t];

{7.1, 4.4, 8.5, 6.85, 6.54, 16.71, 14.21, 5.87, 7.56, 7.77, 11.55, 10.41, 4.21, 3.31, 18.78,
 2.24, 3.25, 7.11, 5.64, 11.89, 5.86, 2.51, 2.06, 8.31, 6.5, 5.31, 9.21, 7.65, 12.37,
 12.81, 3.79, 5.02, 8.72, 8.58, 13.07, 10.81, 4.18, 6.96, 9.75, 9.76, 10.34, 12.43, 5.11,
 6.82, 6.92, 15.7, 5.39, 9.85, 2.51, 3.71, 5.34, 27.75, 7.87, 4.6, 9.46, 11.24, 15.02,
 5.28, 8.25, 8.48, 8.83, 5.17, 9.06, 7.6, 11.14, 7.45, 8.55, 7.39, 4.65, 24.27, 7.62, 6.62}

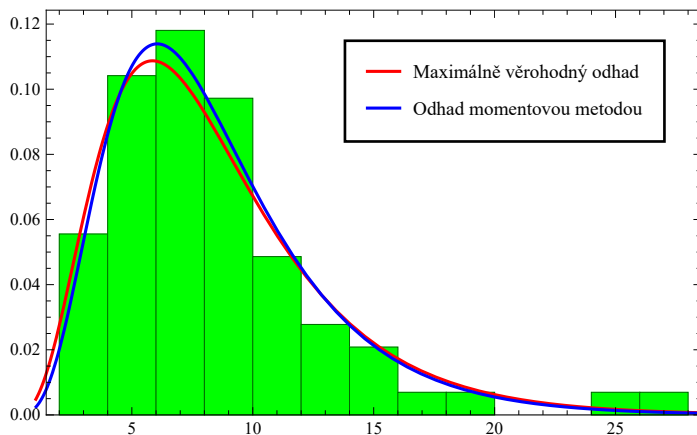
... FindRoot: The line search decreased the step size to within tolerance specified by AccuracyGoal and PrecisionGoal but
was unable to find a sufficient decrease in the merit function. You may need more than MachinePrecision digits of
working precision to meet these tolerances.

{5.47008, 0.280892}

{4.66872, 0.263254}

p1 = Plot[{f2p, f3p}, {t, 1, 30},
  PlotRange → All, PlotStyle → {Red, Blue}, PlotTheme → "Scientific",
  PlotLegends → Placed[LineLegend[{"Maximálně věrohodný odhad", 10},
  Style["Odhad momentovou metodou", 10]}, LegendFunction → Framed], {{.7, .8}}];
h1 = Histogram[data, {2}, "PDF", PlotTheme → "Scientific", ChartStyle → Green];
Show[h1, p1]

```



Druhá datová sada

```

data = {12.20, 23.56, 23.74, 25.87, 31.98,
        37, 41.35, 47.38, 55.46, 58.36, 63.47, 68.46, 78.26,
        74.47, 81.43, 84, 92, 94, 110, 112, 119, 127, 130, 133, 140, 146,
        155, 159, 173, 179, 194, 195, 209, 249, 281, 319, 339, 432, 469,
        519, 633, 725, 817, 1776}

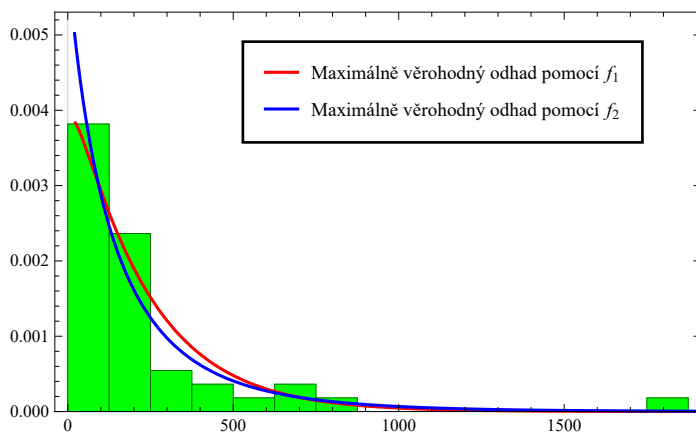
{12.2, 23.56, 23.74, 25.87, 31.98, 37, 41.35, 47.38, 55.46, 58.36, 63.47, 68.46,
 78.26, 74.47, 81.43, 84, 92, 94, 110, 112, 119, 127, 130, 133, 140, 146, 155, 159,
 173, 179, 194, 195, 209, 249, 281, 319, 339, 432, 469, 519, 633, 725, 817, 1776}

r1 = MLE[f1];
r2 = MLE[f2];

{ $\alpha$ ,  $\lambda$ } = r1
f2p = f1 [t] ;
{ $\alpha$ ,  $\lambda$ } = r2
f3p = f2 [t] ;
p1 = Plot[{f2p, f3p}, {t, 20, 2500},
  PlotRange  $\rightarrow$  All, PlotStyle  $\rightarrow$  {Red, Blue}, PlotTheme  $\rightarrow$  "Scientific",
  PlotLegends  $\rightarrow$  Placed[LineLegend[{Style["Maximálně věrohodný odhad pomocí  $f_1$ ", 10],
  Style["Maximálně věrohodný odhad pomocí  $f_2$ ", 10]},
  LegendFunction  $\rightarrow$  Framed], {{.6, .8}}]];
{1.07137, 0.00468621}
{0.694028, 0.00855385}

h1 = Histogram[data, {125}, "PDF", PlotTheme  $\rightarrow$  "Scientific", ChartStyle  $\rightarrow$  Green];
Show[h1, p1]

```



Simulace

```
(*pomocné funkce*)
Clear[α, λ, t, f1, f2, EX]
f1[t_] := α λ (1 - e-t λ)α-1 e-t λ
f2[t_] := α λ (1 + λ t)α-1 Exp[1 - (1 + λ t)α]

Logf[f1, t_] := Log[α λ (1 - e-t λ)α-1] - t λ
Logf[f2, t_] := Log[α λ (1 + λ t)α-1] + 1 - (1 + λ t)α

EX[K_, f_] := EX[K, f] =
  Assuming[{α > 0, λ > 0}, Integrate[t^K f[t], {t, 0, Infinity}]] // FullSimplify

(*Metoda maxiální věrohodnosti*)
MLE[ $\mathcal{D}$ _, f_, N_] := Module[{},
  data = RandomVariate[ $\mathcal{D}$ , N];
  Clear[α, λ];
  Clear[sectiLogy, ucfc];
  sectiLogy = Table[Logf[f, t], {t, data}];
  ucfc[α_, λ_] := Total[sectiLogy];
  r =
    NMaximize[{ucfc[α, λ], α > 0, λ > 0, α < 0.8, λ < 2}, {α, λ}, MaxIterations → 100000];
  {α, λ} /. r[[2]]
]

(*Momentová metoda*)
MM[ $\mathcal{D}$ _, f_, N_] :=
  Module[
    {},
    data = RandomVariate[ $\mathcal{D}$ , N];
    Clear[α, λ];
    mean = Mean[data];
    meanTo2 = Mean[data^2];
    Clear[ex1, ex2];
    ex1[α_, λ_] := EX[1, f];
    ex2[α_, λ_] := EX[2, f];

    res = FindRoot[{ex1[α, λ] == mean, ex2[α, λ] == meanTo2},
      {{α, 1}, {λ, 1}}, MaxIterations → 10000];
    {α, λ} /. res
  ]
```

```

(*Momentová metoda speciálně pro zobecnění II pro lepší stabilitu řešení*)
MM2[ $\mathcal{D}$ _, f_, N_] :=
Module[
  {},
  data = RandomVariate[ $\mathcal{D}$ , N];
  Clear[ $\alpha$ ,  $\lambda$ ];
  mean = Mean[data];
  meanTo2 = Mean[data^2];
  Clear[ex1, ex2];
   $\lambda = \left(-1 + e \text{Gamma}\left[1 + \frac{1}{\alpha}, 1\right]\right) / \text{mean};$ 
  
$$\text{ex2}[\alpha_] := \frac{1 + e \text{ExpIntegralE}\left[-\frac{2}{\alpha}, 1\right] - 2 e \text{Gamma}\left[1 + \frac{1}{\alpha}, 1\right]}{\left(\left(-1 + e \text{Gamma}\left[1 + \frac{1}{\alpha}, 1\right]\right) / \text{mean}\right)^2};$$

  res = FindRoot[ex2[ $\alpha$ ] == meanTo2, { $\alpha$ , 1}, MaxIterations -> 10000]; {
     $\alpha$ ,  $\lambda$ } /. res
]

```

```

runSimulations[ $\alpha_0$ _,  $\lambda_0$ _, f_, estimatorFunction_] := Module[{},
   $\alpha$  =  $\alpha_0$ ;
   $\lambda$  =  $\lambda_0$ ;
  Clear[ $\mathcal{D}$ ];
   $\mathcal{D}$  = ProbabilityDistribution[f[t], {t, 0, Infinity}];
  pocetExperimentu = 10;
  rozsahExperimentu = {100, 1000, 10000};
  SeedRandom[31415926];
  bp = Range[6] 0;
  Table[
    simulationres =
      Table[estimatorFunction[ $\mathcal{D}$ , f, rozsahExperimentu[[j]]], {i, 1, pocetExperimentu}];
    bp[[j]] = simulationres[[All, 1]];
    bp[[j + 3]] = simulationres[[All, 2]]; {j, 3}];
  bp;

  tbl = Table[{ $10^{i+1}$ , Mean[bp[[i]] -  $\alpha_0$ ],
    Variance[bp[[i]]], Mean[(bp[[i]] -  $\alpha_0$ )^2]}, {i, 3}];
  tbl2 = Table[{Mean[bp[[i]] -  $\lambda_0$ ], Variance[bp[[i]]],
    Mean[(bp[[i]] -  $\lambda_0$ )^2]}, {i, 4, 6}];
  restb = Transpose[Join[Transpose[tbl], Transpose[tbl2]]];
  restb = Round[restb, .00001];
  r1 = TableForm[restb, TableHeadings →
    {None, {"N", "E( $\alpha$ - $\alpha$ )", "Var  $\alpha$ ", "MSE  $\alpha$ ", "E( $\lambda$ - $\lambda$ )", "Var  $\lambda$ ", "MSE  $\lambda$ "}}];

  abp = bp[[1 ;; 3]];
  lbp = bp[[4 ;; 6]];
  {r1, GraphicsRow[
    BoxWhiskerChart[abp, ChartLabels → {100, 1000, 10000}, PlotTheme → "Scientific",
      AxesLabel → {"Rozsah experimentu", "Odhad parametru"}],
    BoxWhiskerChart[lbp, ChartLabels → {100, 1000, 10000},
      PlotTheme → "Scientific", ChartStyle → Blue]
  ]}]
]

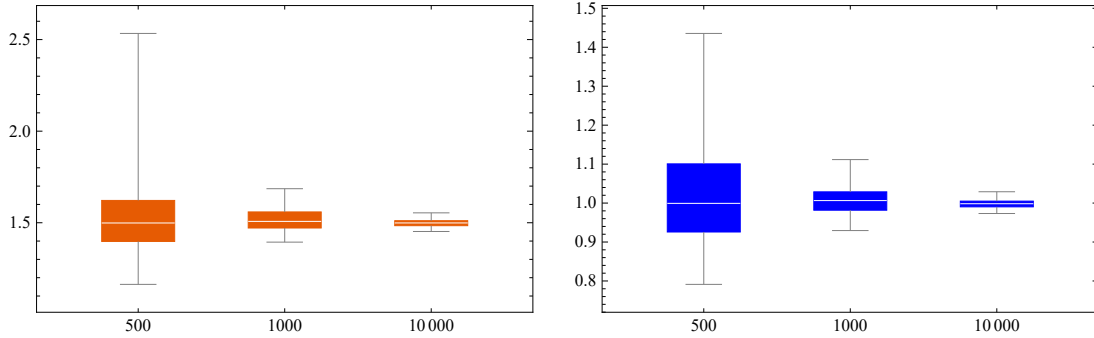
```

```

simulRes2 = runSimulations[1.5, 1, f1, MLE];
simulRes2[[1]]
simulRes2[[2]]

```

| N | $E(\alpha - \hat{\alpha})$ | $\text{Var } \alpha$ | $\text{MSE } \alpha$ | $E(\lambda - \hat{\lambda})$ | $\text{Var } \lambda$ | $\text{MSE } \lambda$ |
|--------|----------------------------|----------------------|----------------------|------------------------------|-----------------------|-----------------------|
| 100. | 0.03627 | 0.05706 | 0.05781 | 0.00837 | 0.01422 | 0.01415 |
| 1000. | 0.01424 | 0.00379 | 0.00395 | 0.00523 | 0.00125 | 0.00127 |
| 10000. | -0.00133 | 0.00039 | 0.00039 | -0.0015 | 0.00012 | 0.00012 |

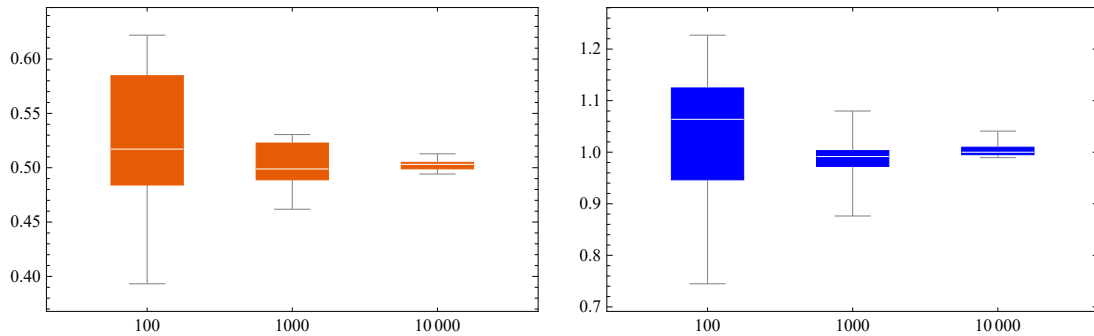


```

simulRes3 = runSimulations[0.5, 1, f1, MLE];
simulRes3[[1]]
simulRes3[[2]]

```

| N | $E(\alpha - \hat{\alpha})$ | $\text{Var } \alpha$ | $\text{MSE } \alpha$ | $E(\lambda - \hat{\lambda})$ | $\text{Var } \lambda$ | $\text{MSE } \lambda$ |
|--------|----------------------------|----------------------|----------------------|------------------------------|-----------------------|-----------------------|
| 100. | 0.02607 | 0.00511 | 0.00528 | 0.03415 | 0.02011 | 0.01927 |
| 1000. | 0.00036 | 0.00049 | 0.00044 | -0.00855 | 0.00299 | 0.00276 |
| 10000. | 0.00318 | 0.00004 | 0.00004 | 0.00595 | 0.00029 | 0.00029 |

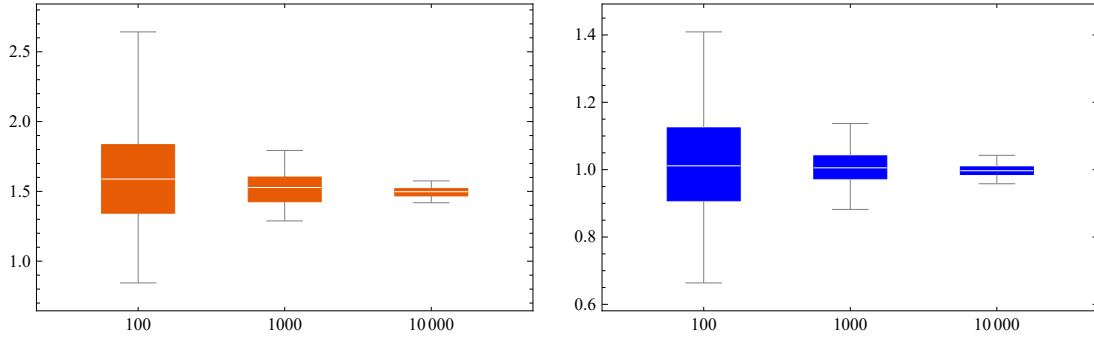


```

simulRes5 = runSimulations[1.5, 1, f1, MM];
simulRes5[[1]]
simulRes5[[2]]

```

| N | $E(\alpha-\alpha)$ | Var α | MSE α | $E(\lambda-\lambda)$ | Var λ | MSE λ |
|--------|--------------------|--------------|--------------|----------------------|---------------|---------------|
| 100. | 0.10622 | 0.14127 | 0.15114 | 0.02865 | 0.02756 | 0.0281 |
| 1000. | 0.02206 | 0.01243 | 0.0128 | 0.00755 | 0.00274 | 0.00277 |
| 10000. | -0.00416 | 0.00128 | 0.00129 | -0.00269 | 0.00024 | 0.00024 |



```

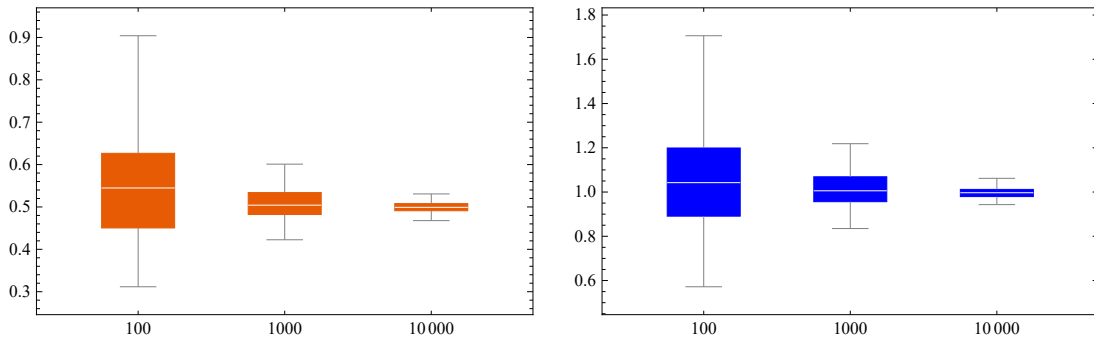
simulRes6 = runSimulations2[0.5, 1, f1, MM];
simulRes6[[1]]
simulRes6[[2]]

```

... Interpolation: The point 0.99999997697141` in dimension 1 is duplicated.

... Interpolation: The point 0.9999999728094472` in dimension 1 is duplicated.

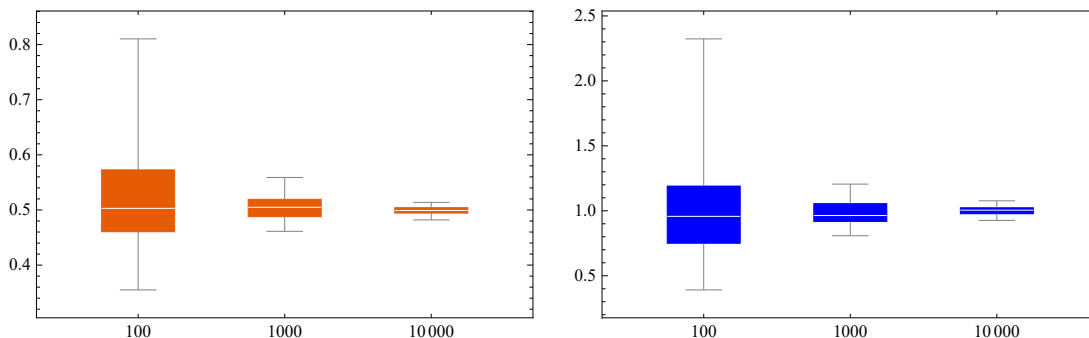
| N | $E(\alpha-\alpha)$ | Var α | MSE α | $E(\lambda-\lambda)$ | Var λ | MSE λ |
|--------|--------------------|--------------|--------------|----------------------|---------------|---------------|
| 100. | 0.04488 | 0.01622 | 0.01807 | 0.06472 | 0.05971 | 0.0633 |
| 1000. | 0.0073 | 0.00159 | 0.00163 | 0.01249 | 0.0059 | 0.006 |
| 10000. | -0.00089 | 0.00018 | 0.00018 | -0.00312 | 0.00051 | 0.00052 |




```

simulRes23 = runSimulations[0.5, 1, f2, MLE];
simulRes23[[1]]
simulRes23[[2]]
    
```

| N | $E(\alpha-\hat{\alpha})$ | Var α | MSE α | $E(\lambda-\hat{\lambda})$ | Var λ | MSE λ |
|--------|--------------------------|--------------|--------------|----------------------------|---------------|---------------|
| 100. | 0.02335 | 0.00814 | 0.0086 | 0.00696 | 0.13383 | 0.13254 |
| 1000. | 0.00521 | 0.00052 | 0.00054 | -0.01442 | 0.00939 | 0.00951 |
| 10000. | -0.00071 | 0.00005 | 0.00005 | 0.00262 | 0.00122 | 0.00122 |



```

simulRes23 = runSimulations[0.5, 1, f2, MM2];
simulRes23[[1]]
simulRes23[[2]]
    
```

| N | $E(\alpha-\hat{\alpha})$ | Var α | MSE α | $E(\lambda-\hat{\lambda})$ | Var λ | MSE λ |
|--------|--------------------------|--------------|--------------|----------------------------|---------------|---------------|
| 100. | 0.05557 | 0.01092 | 0.01291 | -0.08375 | 0.21198 | 0.19779 |
| 1000. | 0.00259 | 0.0007 | 0.00064 | -0.01259 | 0.01395 | 0.01271 |
| 10000. | 0.00344 | 0.00013 | 0.00013 | -0.01315 | 0.00296 | 0.00283 |

