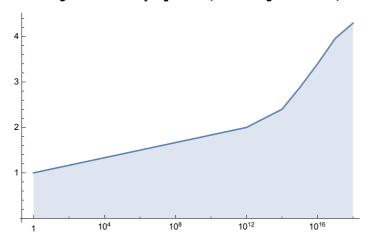
```
Rivat alpha factors determined by running pi(x) benchmarks *)
                  alphaDelegliseRivat = \{ (* \{x, alpha\} *) \{1, 1\}, \{10^1, 1\}, \{10^2, 1\}, 
                         \{10^3, 1\}, \{10^4, 1\}, \{10^5, 1\}, \{10^6, 1.172\}, \{10^7, 1.861\},
                         \{10^8, 2.778\}, \{10^9, 3.955\}, \{10^10, 5.426\}, \{10^11, 7.795\},
                         \{10^12, 10.960\}, \{10^13, 15.22\}, \{10^15, 27.16\}, \{10^16, 34.80\},
                         \{10^17, 43.88\}, \{10^18, 54.56\}, \{10^19, 67.01\}, \{10^20, 81.38\},
                         \{10^21, 97.86\}, \{10^22, 116.61\}, \{10^23, 137.83\}, \{10^24, 161.69\}\}
\text{Out[11]= } \{\{1,1\},\{10,1\},\{100,1\},\{1000,1\},\{10000,1\},\{100000,1\},
                      \{1000000, 1.172\}, \{10000000, 1.861\}, \{100000000, 2.778\},
                      \{10000000000, 3.955\}, \{10000000000, 5.426\}, \{10000000000, 7.795\},
                      \{1000000000000, 10.96\}, \{100000000000, 15.22\},
                      \{100000000000000000, 27.16\}, \{1000000000000000000, 34.8\},
                      \{100\,000\,000\,000\,000\,000,\,43.88\}, \{1\,000\,000\,000\,000\,000\,000,\,54.56\},
                      \{10\,000\,000\,000\,000\,000\,000, 67.01\}, \{100\,000\,000\,000\,000\,000\,000, 81.38\},
                      {1000000000000000000000000, 97.86}, {10000000000000000000000, 116.61},
                      \{100\,000\,000\,000\,000\,000\,000\,000,\,137.83\},\,\{1\,000\,000\,000\,000\,000\,000\,000,\,161.69\}\}
 ln[12]= ListLogLinearPlot[alphaDelegliseRivat, Filling → Bottom, Joined → True]
                  150
                  100
Out[12]=
                    50
                                                                                                                                                                         10^{23}
                                           1000.0
 In[13]:= NonlinearModelFit[alphaDelegliseRivat,
                     a (Log[x])^3 + b (Log[x])^2 + c Log[x] + d, \{a, b, c, d\}, x
\text{Out} [13] = \text{ FittedModel } \Big[ \Big[ 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x]^3 \Big] \Big] \Big] = 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x]^3 \Big] \Big] \Big] \Big[ 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x]^3 \Big] \Big] \Big] \Big[ 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x]^3 \Big] \Big] \Big[ 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x]^3 \Big] \Big] \Big[ 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x]^3 \Big] \Big] \Big[ 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x]^3 \Big] \Big] \Big[ 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x]^3 \Big] \Big] \Big[ 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x]^3 \Big] \Big] \Big[ 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x]^3 \Big] \Big] \Big[ 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x]^3 \Big] \Big] \Big[ 0.0672202 + 0.483613 \, \text{Log}[x] - 0.0508992 \, \text{Log}[x]^2 + 0.0017154 \, \text{Log}[x] - 0.0017154 \, \text{Lo
                  (* Below is another formula which is quite accurate for calculating
                         the Deleglise-Rivat alpha factor in primecount. The constant
                         1200 has been obtained by running many pi(10^20) benchmarks. *)
                  alpha[x_] := (Log[x])^3 / (1200 (Log[Log[10^20]] / Log[Log[x]])^3)
                  (* List of fast Lagarias-Miller-
                     Odlyzko alpha factors determined by running pi(x) benchmarks *)
                  alphaLMO = \{(* \{x, alpha\} *) \{1, 1\}, \{10^12, 2\}, \{10^13, 2.2\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14, 2.4\}, \{10^14,
                         \{10^15, 2.877\}, \{10^16, 3.398\}, \{10^17, 3.960\}, \{10^18, 4.295\}\}
                  \{\{1,1\},\{1000000000000,2\},\{100000000000,2.2\},\{1000000000000,2.4\},
                     \{10000000000000000, 2.877\}, \{10000000000000000, 3.398\},
                      \{10000000000000000000, 3.96\}, \{1000000000000000000, 4.295\}\}
```

In[11]:= (* List of fast Deleglise-

 ${\tt ListLogLinearPlot[alphaLMO, Filling \rightarrow Bottom, Joined \rightarrow True]}$



 $Nonlinear Model Fit[alpha LMO, a (Log[x])^2 + b Log[x] + c, \{a, b, c\}, x]$

 $\label{eq:fittedModel} \textbf{FittedModel} \big[\big| \ 1.00454 - 0.0656652 \, \text{Log[x]} + 0.00352628 \, \text{Log[x]}^2 \\$