Exponential NDF

This is code to accompany the book:

A Hitchhiker's Guide to Multiple Scattering

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notation

$$u = \mathbf{m} \cdot \mathbf{n} = Cos[\theta_m]$$

 $\alpha = roughness$

Definitions and derivations

In[2787]:= Exponential`D[u_,
$$\alpha_$$
] :=
$$\frac{2 e^{-\frac{2\sqrt{1-u^2}}{u\alpha}}}{\pi u^4 \alpha^2}$$
 HeavisideTheta[u]

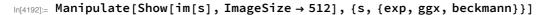
comparison to Beckmann and GGX

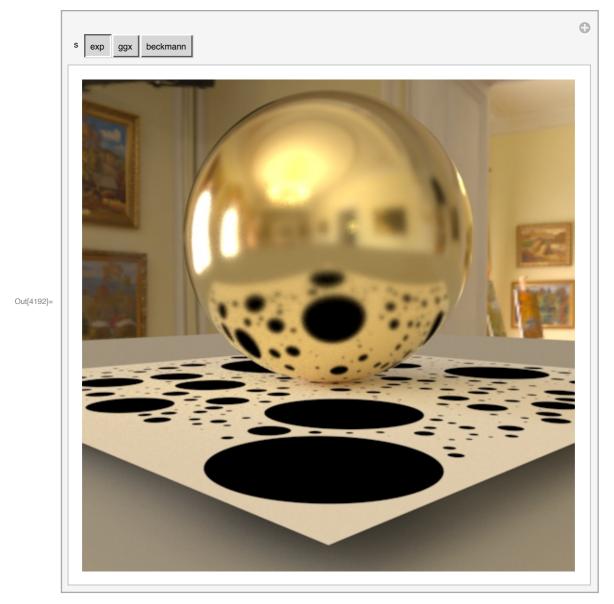
In[4189]:= Clear[im];











relationship to Beckmann

The Exponential NDF can be written as a Gamma superposition of Beckmanns:

In[1900]:= Integrate [Beckmann'D[u, $\alpha \sqrt{m}] \times PDF[GammaDistribution[\frac{3}{2}, 1]][m],$ {m, 0, Infinity}, Assumptions \rightarrow 0 < u < 1 && α > 0 && a > 0

Out[1900]=
$$\frac{2 e^{-\frac{2\sqrt{1-u^2}}{u\alpha}}}{\pi u^4 \alpha^2}$$

or as a Chi-3 superposition:

Integrate [Beckmann`D[u,
$$\frac{\alpha}{\sqrt{2}}$$
 m] × PDF [ChiDistribution[3]] [m],
$$\{\text{m, 0, Infinity}\}\text{, Assumptions} \rightarrow 0 < \text{u} < 1 \&\& \alpha > 0 \&\& \text{a} > 0 \}$$

$$2 e^{-\frac{2\sqrt{1-u^2}}{u\alpha}}$$

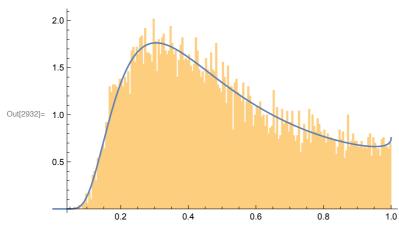
$$\pi u^4 \alpha^2$$

importance sampling (for Walter's method)

Use the Chi-3 expansion of the NDF we can sample the full NDF using 3 Gaussian random variates and 1 uniform random variate:

$$1 / \left(\sqrt{\left(1 - \left(\frac{\alpha}{\sqrt{2}} \; \mathsf{Norm}[\{g_1, \, g_2, \, g_3\}]\right)^2 \left(\mathsf{Log}[u_1]\right)} \right)} \right) \\ = \frac{1}{\sqrt{1 - \frac{1}{2} \; \alpha^2 \; \left(\mathsf{Abs}[g_1]^2 + \mathsf{Abs}[g_2]^2 + \mathsf{Abs}[g_3]^2\right) \; \mathsf{Log}[u_1]}} \\ \\ \mathsf{In}[2932] = \mathsf{With} \left[\{\alpha = 2.3\}, \\ \mathsf{Show} \right[\\ \mathsf{Histogram}[\mathsf{Table}[\\ 1 / \left(\sqrt{\left(1 - \left(\frac{\alpha}{\sqrt{2}} \; \mathsf{Norm}[\{\mathsf{RandomVariate}[\mathsf{NormalDistribution}[]], \; \mathsf{RandomVariate}[\mathsf{NormalDistribution}[]], \right)} \right) \\ \mathsf{RandomVariate}[\mathsf{NormalDistribution}[]], \\ \mathsf{RandomVariate}[\mathsf{NormalDistribution}[]] \right) \right) \\ \left(\mathsf{Log}[\mathsf{RandomReal}[]]) \right) \right), \; \{i, \; \mathsf{Range}[10 \; 000]\} \right], \; 200, \; "\mathsf{PDF"}], \\ \mathsf{Plot}[\mathsf{Exponential} \; \mathsf{D}[\mathsf{u}, \, \alpha] \; 2 \; \mathsf{Pi} \; \mathsf{u}, \; \{\mathsf{u}, \, 0, \, 1\}, \; \mathsf{PlotRange} \; \rightarrow \; \mathsf{All}] \\ \right] \\ \right]$$

... General: Exp[-42566.1] is too small to represent as a normalized machine number; precision may be lost.



shape invariant f(x)

$$\label{eq:linear_line$$

distribution of slopes

In[1887]:= FullSimplify [Exponential`D [
$$\frac{1}{\sqrt{p^2+q^2+1}}$$
, α] $\left(\frac{1}{\sqrt{p^2+q^2+1}}\right)^4$, Assumptions \rightarrow 0 < α < 1 && p > 0 && q > 0]
$$\frac{2 e^{-\frac{2\sqrt{p^2+q^2}}{\alpha}}}{\pi \ \alpha^2}$$

In[1889]:= Exponential P22[p_, q_,
$$\alpha_$$
] := $\frac{2 e^{-\frac{2 \sqrt{p^2 + q^2}}{\alpha}}}{\pi \alpha^2}$

In[1890]:= Integrate[Exponential`P22[p, q, α], {p, -Infinity, Infinity}, {q, -Infinity, Infinity}, Assumptions \rightarrow 0 < α < 1]

Out[1890]= 1

sigma - approximate with Ei sigma:

The cross section and Lambda() function are not known analytically, but we noted the following close approximation:

$$\begin{split} &\inf_{\| u \| = \| 1} \; = \; \frac{1}{6 \, \sqrt{\pi} \, \left(-1 + u^2 \right) \, \alpha^2} \left(\alpha \, \left(3 \, \sqrt{\pi} \, u \, \left(-1 + u^2 \right) \, \alpha + 2 \, e^{\frac{u^2}{\left(-1 + u^2 \right) \, \alpha^2}} \, \sqrt{1 - u^2} \, \left(-\alpha^2 + u^2 \, \left(-1 + \alpha^2 \right) \right) \right) + \\ & 3 \, \sqrt{\pi} \, u \, \left(-\alpha^2 + u^2 \, \left(-2 + \alpha^2 \right) \right) \, \text{Erf} \Big[\frac{u}{\sqrt{1 - u^2} \, \alpha} \Big] \, + \\ & 2 \, \sqrt{\pi} \, u^2 \, \text{Abs} [u] \, \left(1 + 2 \, \text{Erf} \Big[\frac{u^2 \, \sqrt{1 - u^2}}{\alpha \, \text{Abs} [u] - u^2 \, \alpha \, \text{Abs} [u]} \Big] \right) \end{split}$$

```
ln[1948] = With[{\alpha = 2.1},
          Plot[{
             Quiet[NIntegrate[Exponential`D[ui, \alpha] × Delta`\sigma[u, ui], {ui, 0, 1}]],
             Quiet[Ei^{\sigma}[u, 1.7\alpha]]
           }, {u, -1, 1}]
        ]
                                      1.0
                                     0.8
                                      0.6
Out[1948]=
                                     0.4
                                      0.2
        -1.0
```

G1 shadow

```
In[•]:= Ei`Λ[u_, α_] :=
                2 e^{\frac{u^{2}}{(-1+u^{2})\,\alpha^{2}}} \sqrt{1-u^{2}} \alpha \left(-\alpha^{2}+u^{2}\,\left(-1+\alpha^{2}\right)\right) + \sqrt{\pi} u \left(3\,\alpha^{2}+u^{2}\,\left(2-3\,\alpha^{2}\right)\right) \, \text{Erfc}\left[\frac{u}{\sqrt{1-u^{2}}\,\alpha}\right] 
                                                             6\sqrt{\pi} u \left(-1+u^2\right) \alpha^2
 In[2628]:= Ei`G1[u_, \alpha_] := \frac{1}{1 + Ei`\Lambda[u, \alpha]}
 In[2634]:= With [\{\alpha = .8\}],
              Plot[{
                   Quiet\big[u\big/NIntegrate[Exponential`D[ui,\,\alpha]\times Delta`\sigma[u,\,ui],\,\{ui,\,0,\,1\}]\big],
                   Quiet[Ei`G1[u, 1.7α]]
                 , {u, 0, 1}, PlotRange \rightarrow All]
             1
             1.0
            0.8
            0.6
Out[2634]=
            0.4
             0.2
                                 0.2
```

Rational approximation for G1:

Similar to Walter's approximation for Beckmann G1 we find:

```
ln[3048]:= Exponential`Glapprox[u_, \alpha_] := If[#1 < 1.5,
               (4.15428848706139`*^-8 + 5.3171271839868055` #1 + 1.5604787398328208` #1<sup>2</sup>) /
                (1+2.9490717540313374`#1+3.812617848664264`#1²-
                   1.0148658798039354\ \pm13 + 0.18113230976734568\ \pm14\)
              1] & \left[\frac{1}{1.7} \alpha^{-1} \left(\frac{\sqrt{1-u^2}}{u}\right)^{-1}\right]
In[2938]:= With [\{\alpha = 0.8\},
          Plot[{
             Exponential`Glapprox[u, \alpha],
             u/Quiet[NIntegrate[Exponential`D[ui, \alpha] \times Delta`\sigma[u, ui], \{ui, 0, 1\}]]
           \}, {u, 0, 1}, PlotRange \rightarrow All
        1.0
        0.8
        0.6
Out[2938]=
        0.4
        0.2
                      0.2
                                  0.4
                                                                       1.0
                                               0.6
                                                           0.8
```