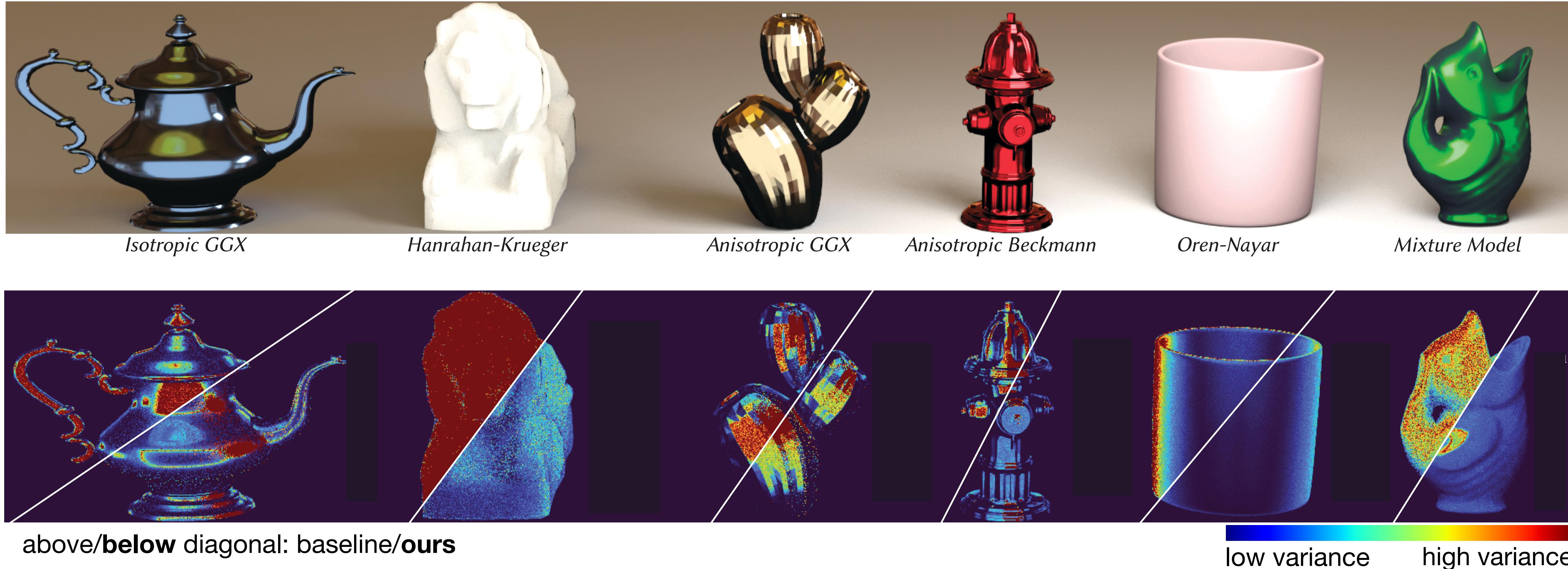


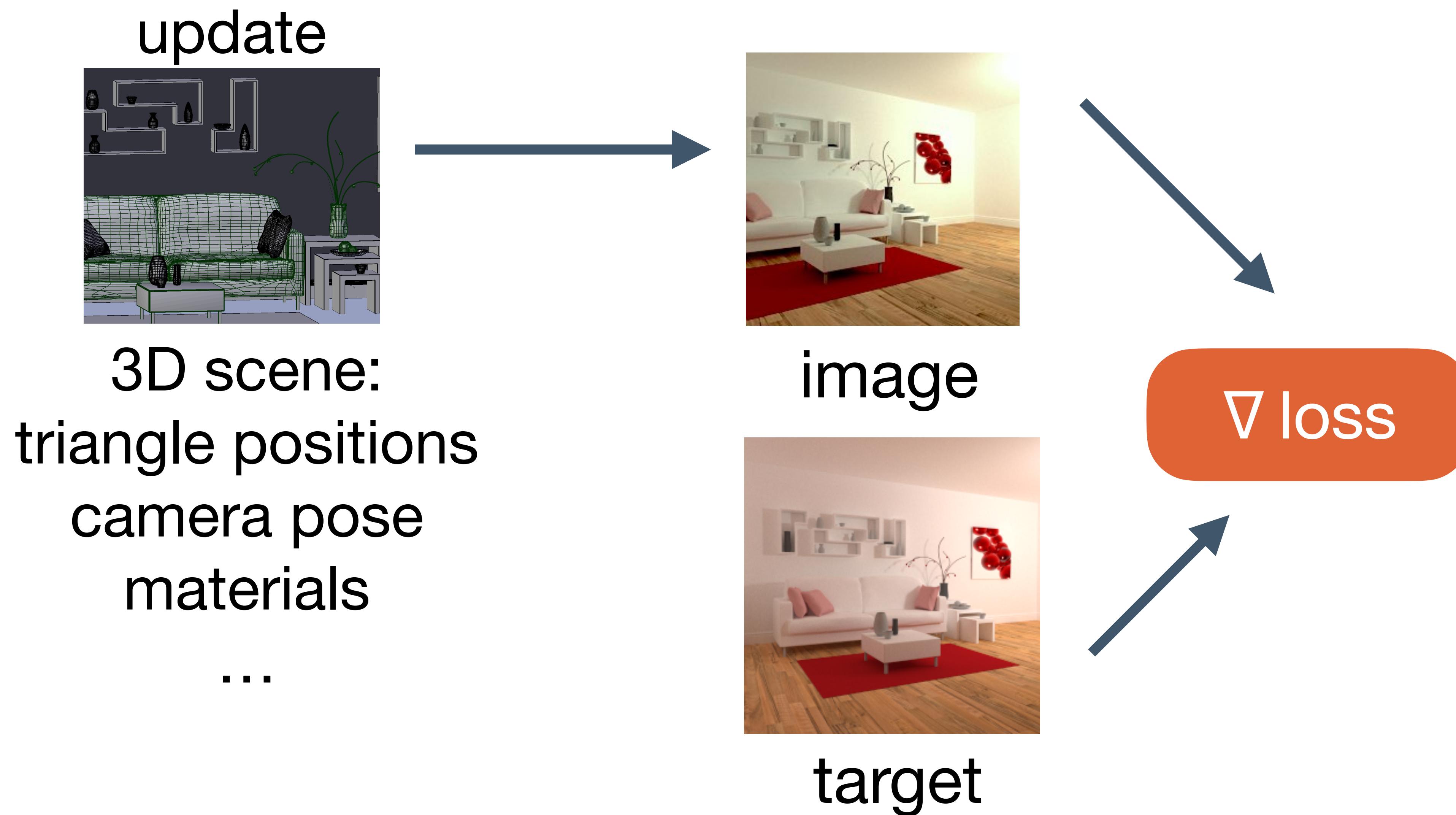
Importance Sampling BRDF Derivatives



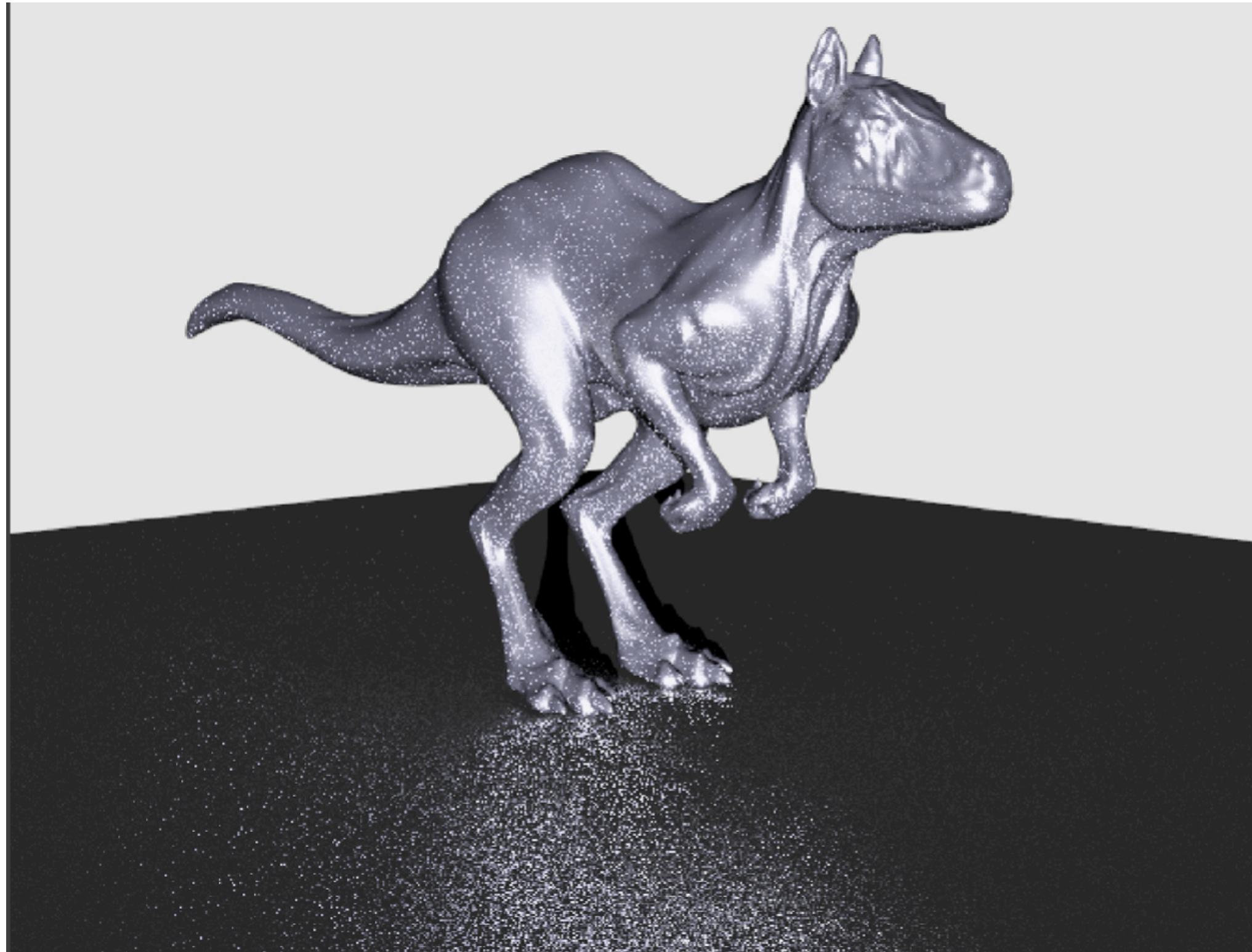
Variance reduction in differentiable rendering by correctly handling the derivative's sign

Yash Belhe, Bing Xu, Sai Praveen Bangaru*, Ravi Ramamoorthi, Tzu-Mao Li – UCSD (*MIT)

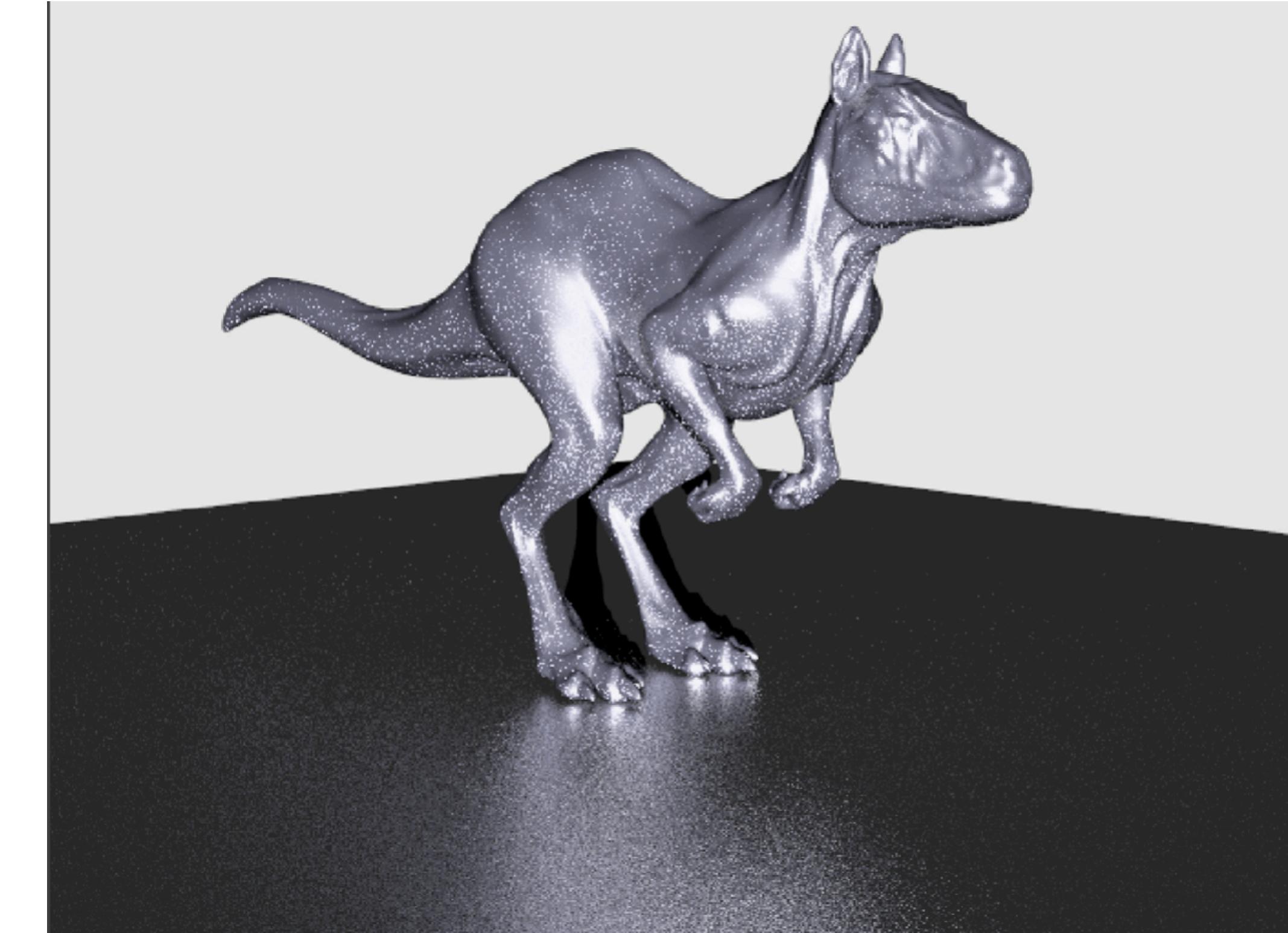
Inverse rendering via derivatives of images



BRDF importance sampling is important in forward rendering



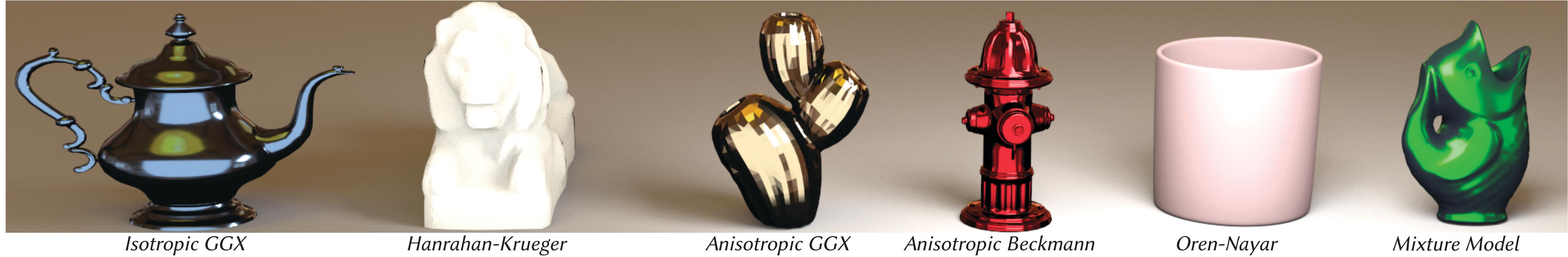
Uniform Sampling



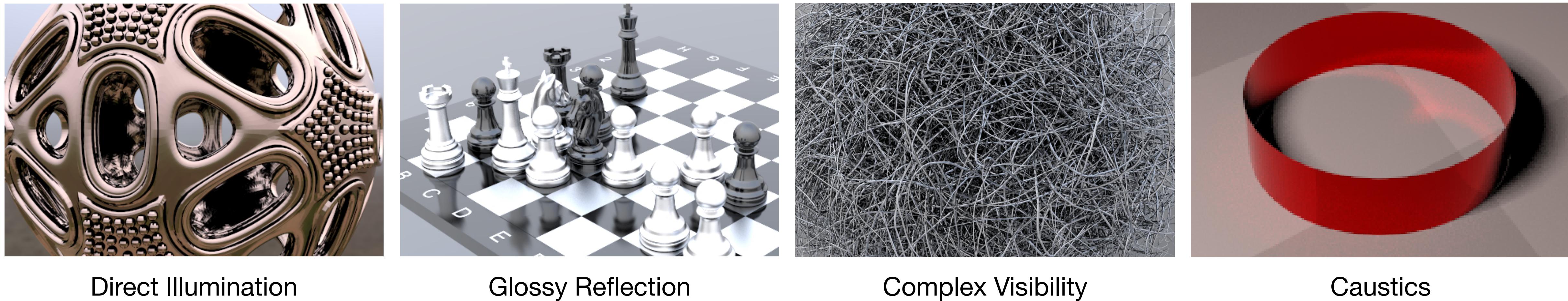
BRDF Importance Sampling

Goal: low-variance importance sampling of BRDF derivatives

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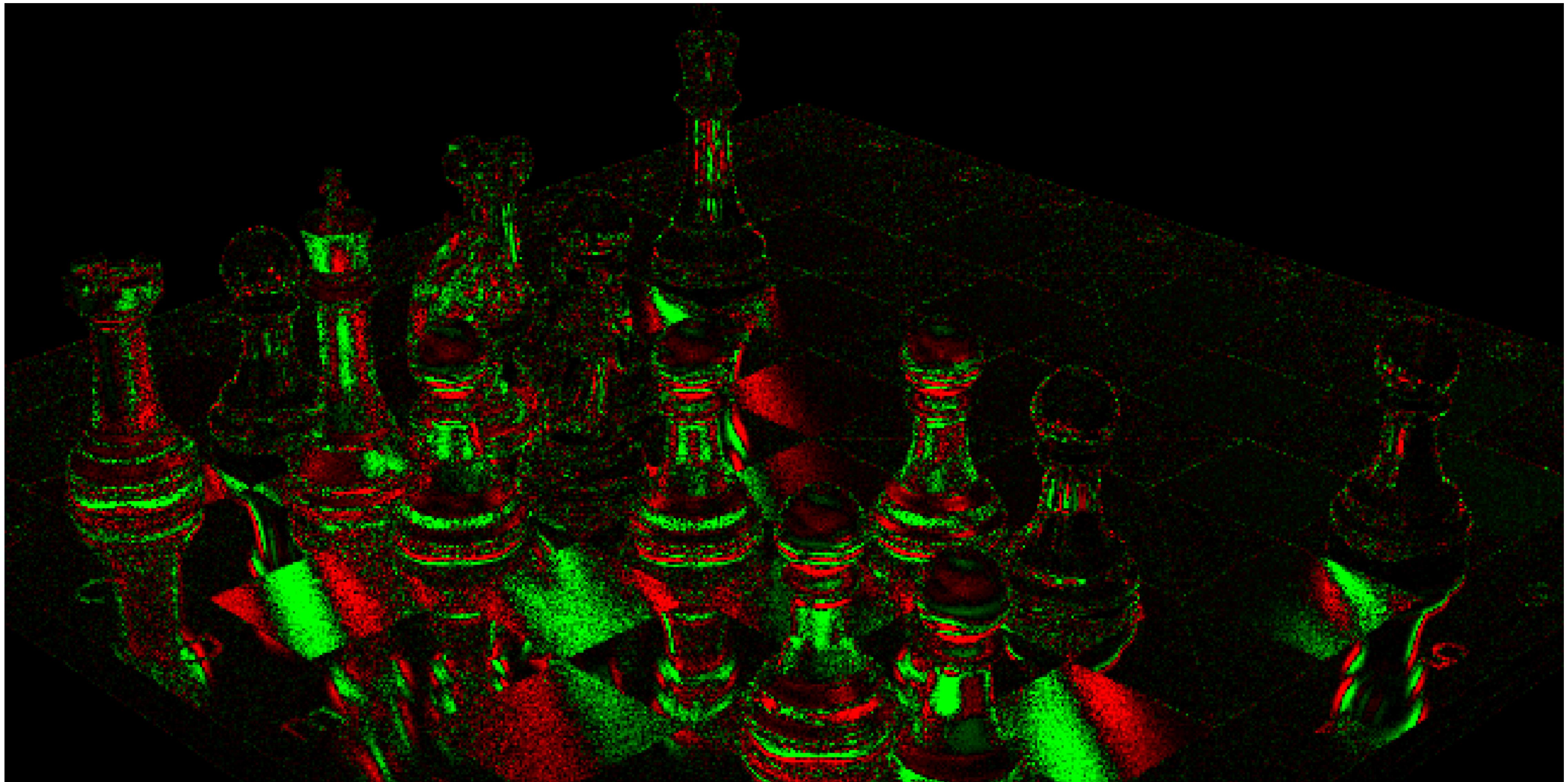


What do BRDF derivatives look like?

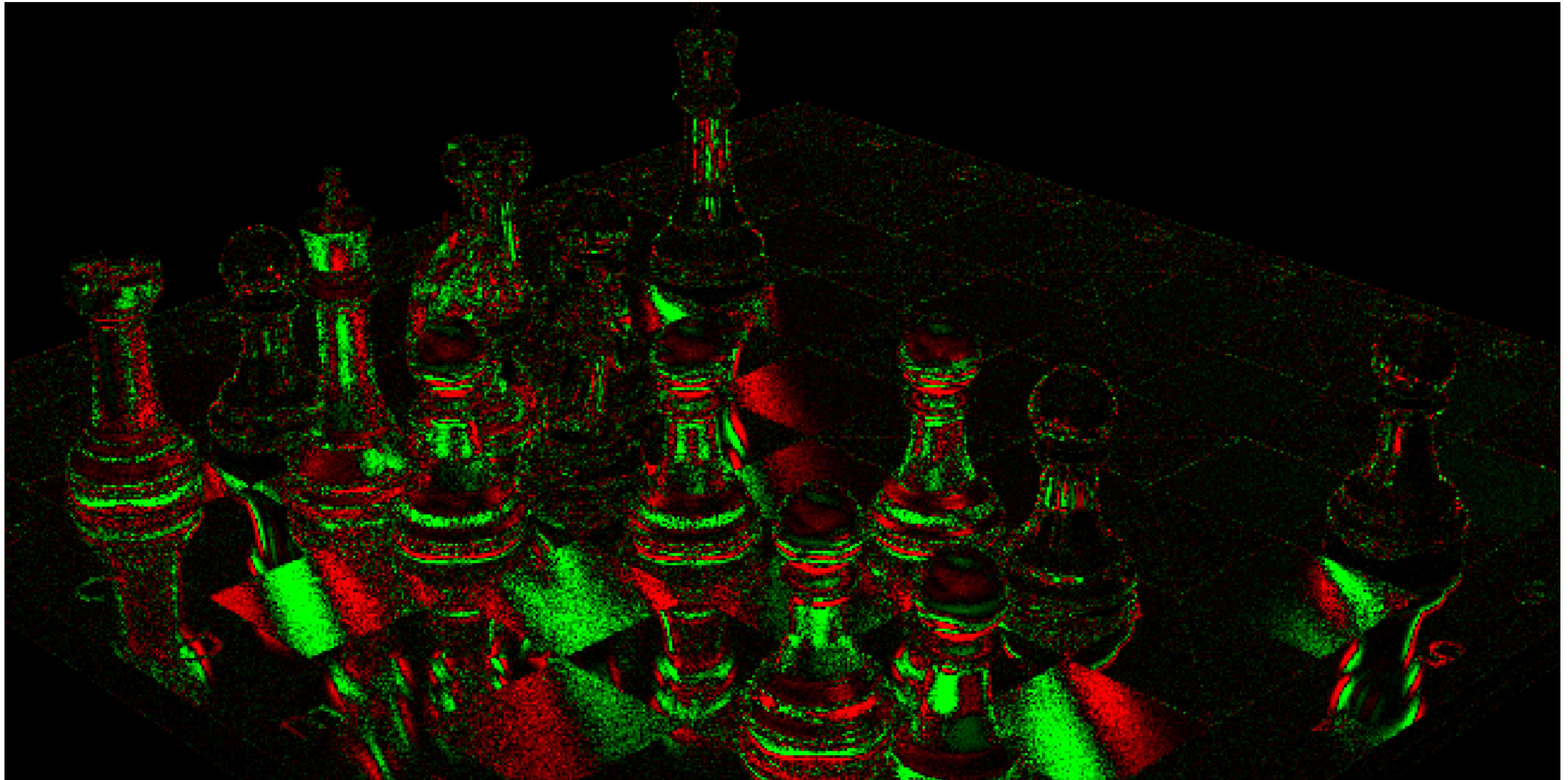


500 samples per pixel

BRDF derivatives can take both **positive and **negative** values**

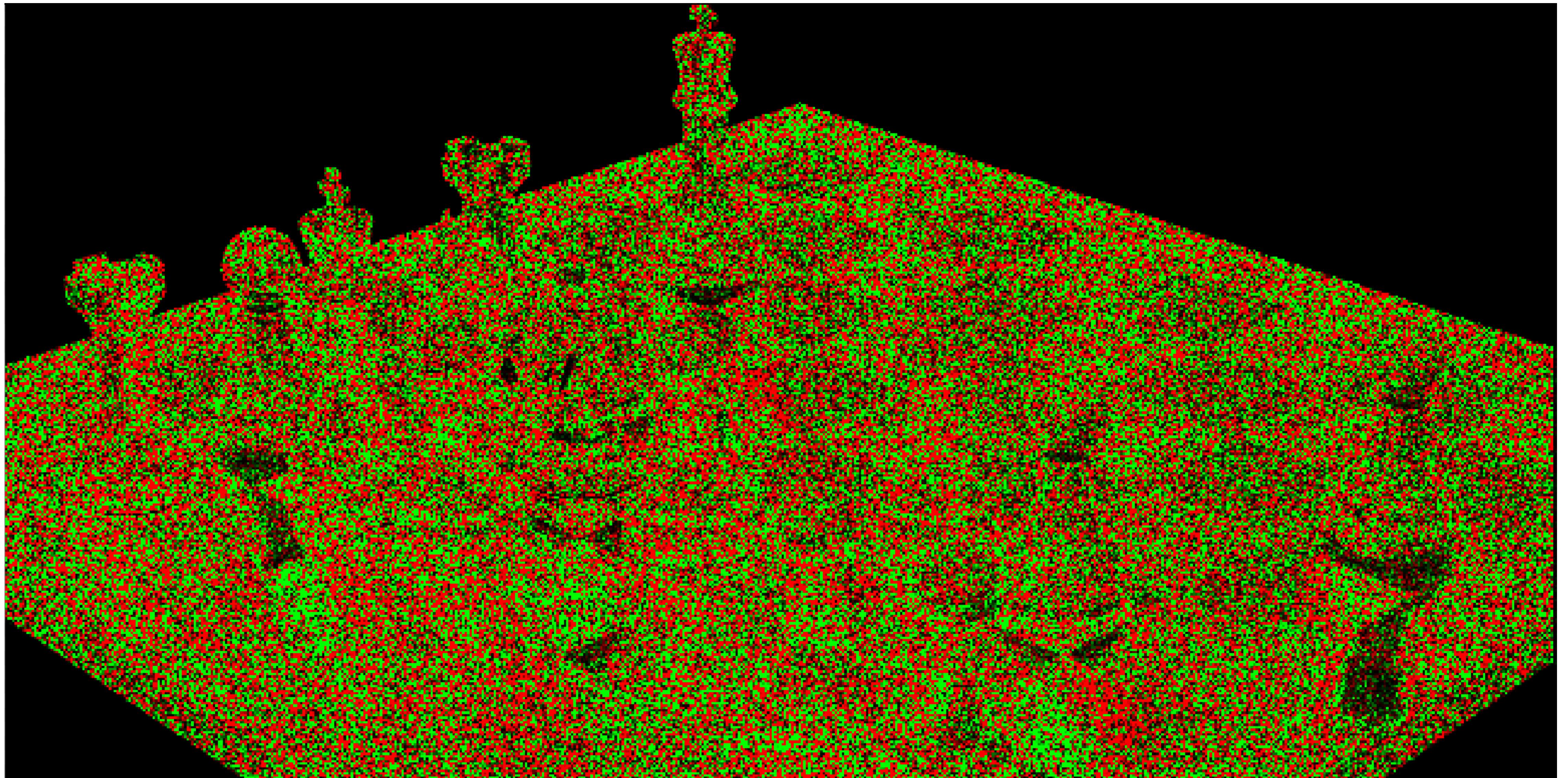


BRDF derivatives can take both **positive and **negative** values**



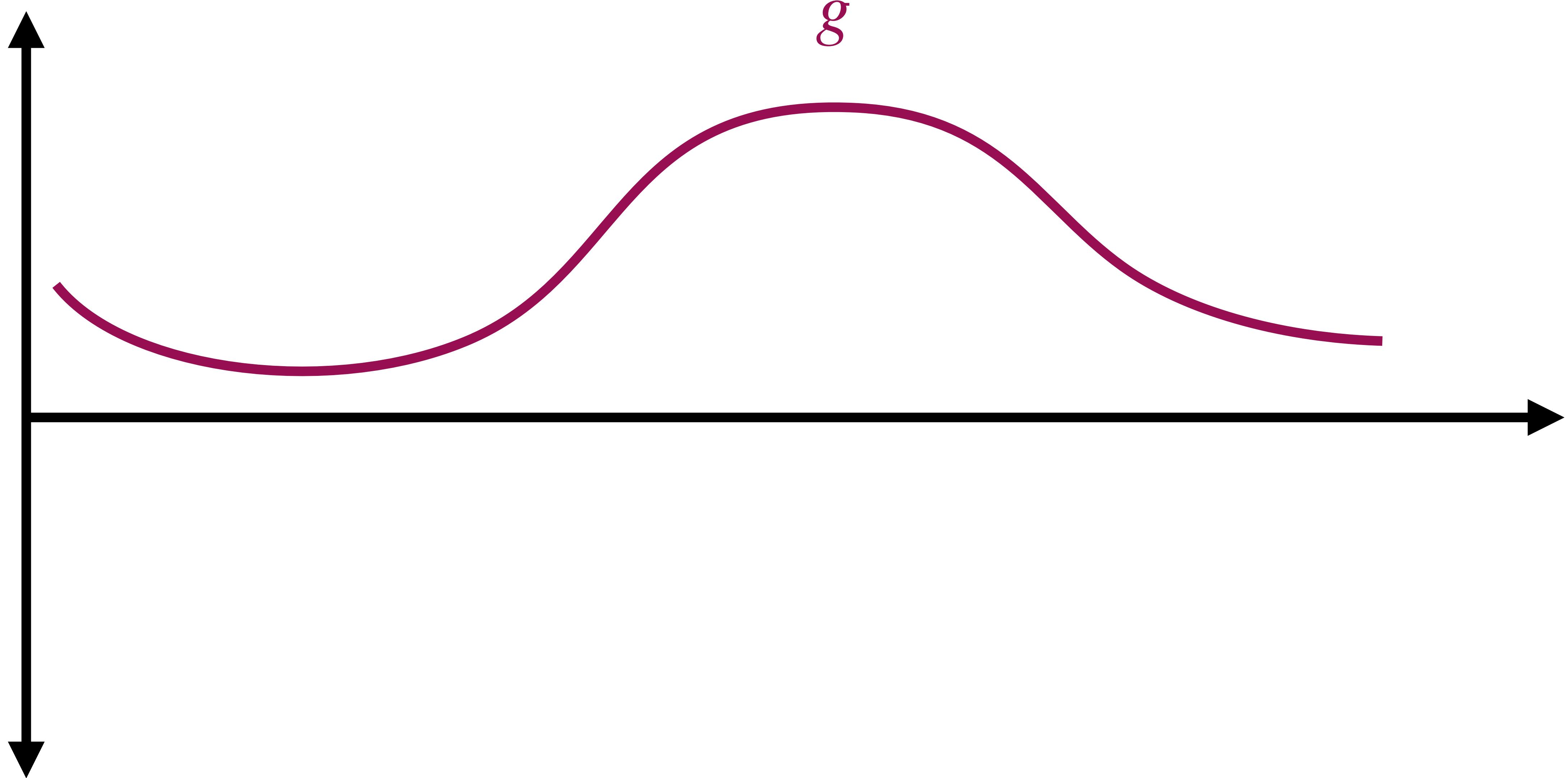
10,000 samples per pixel

BRDF derivatives can be very noisy

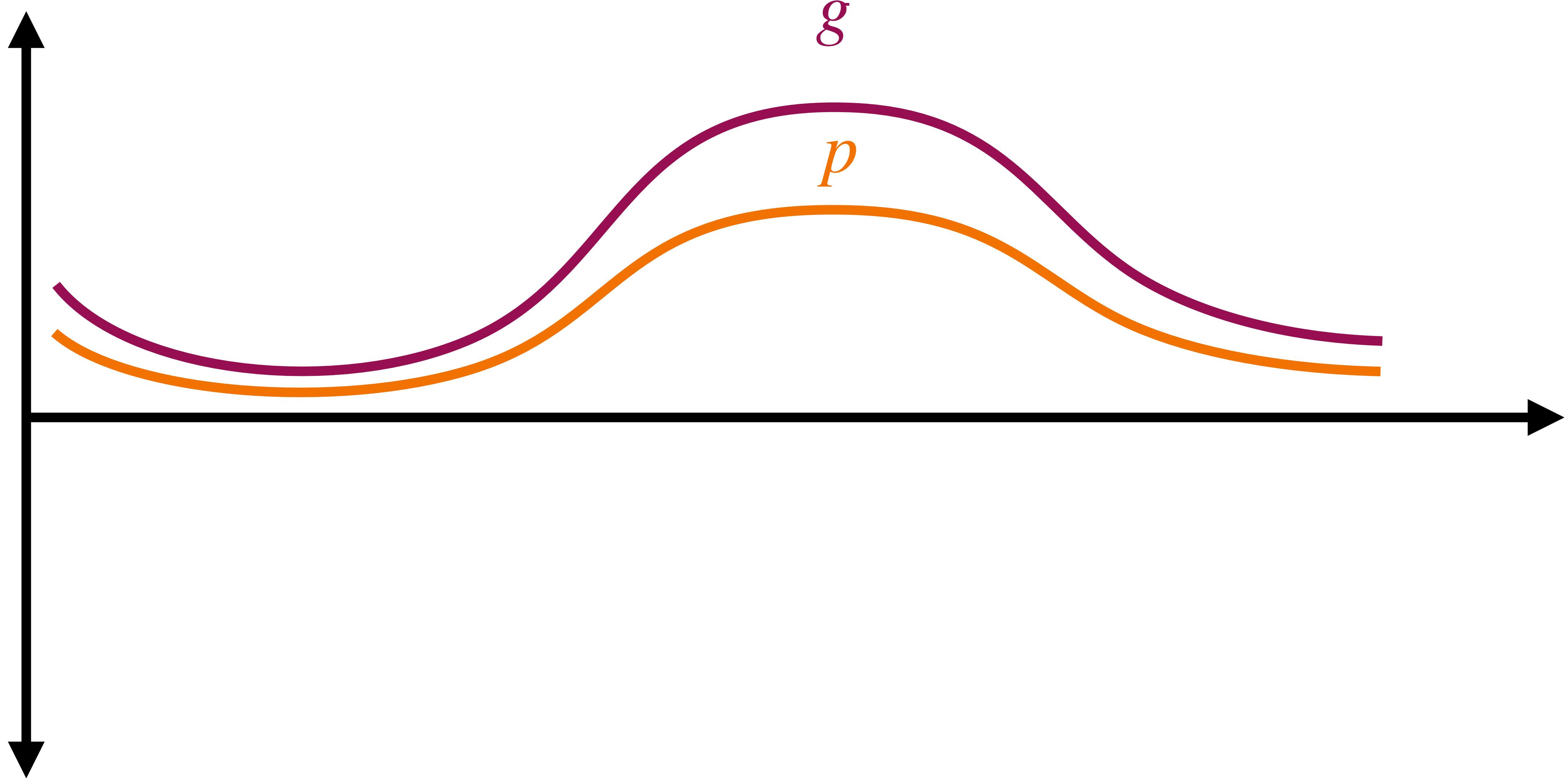


100 samples per pixel

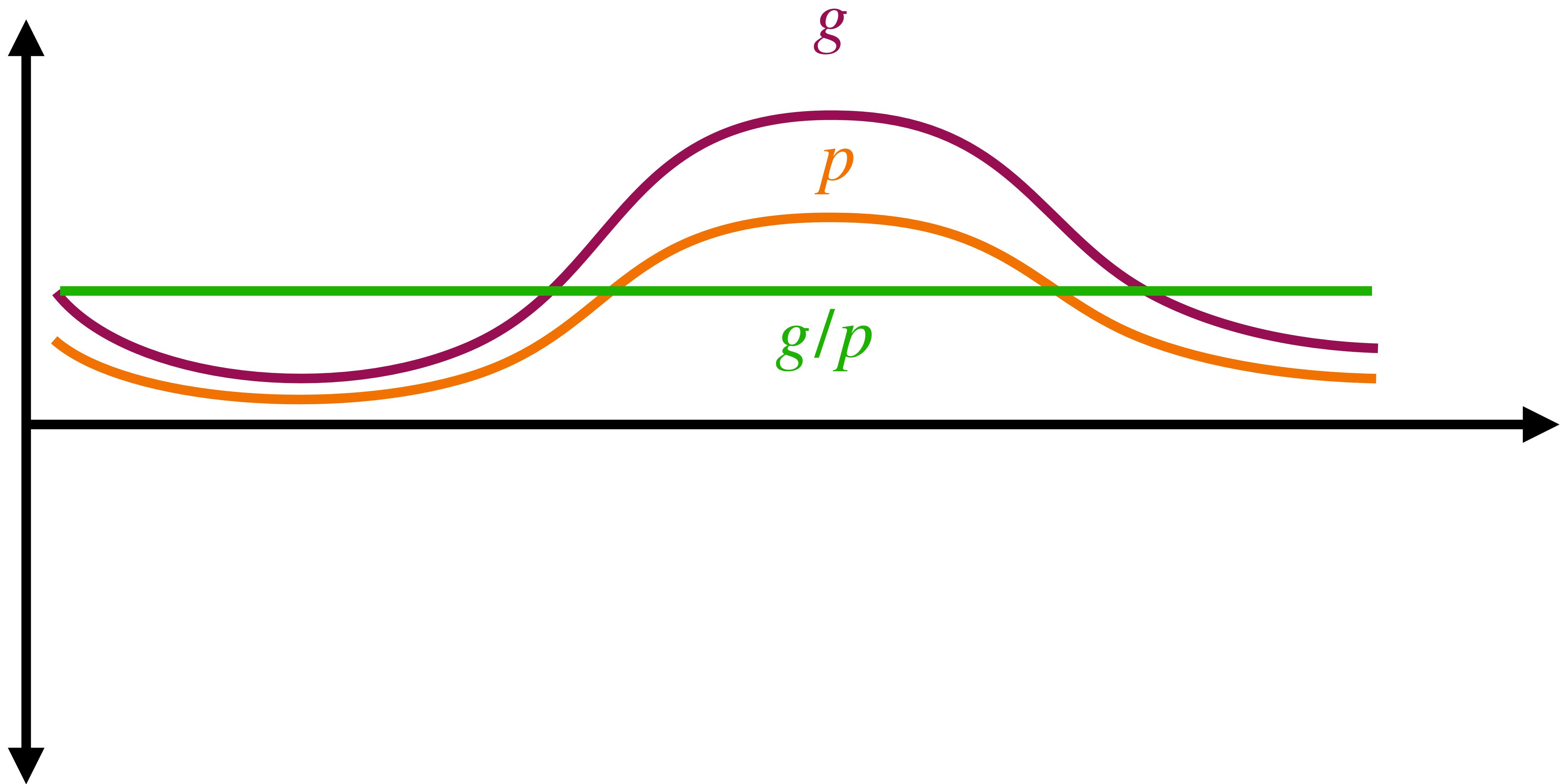
Importance sampling PDF (p) \propto BRDF (g)



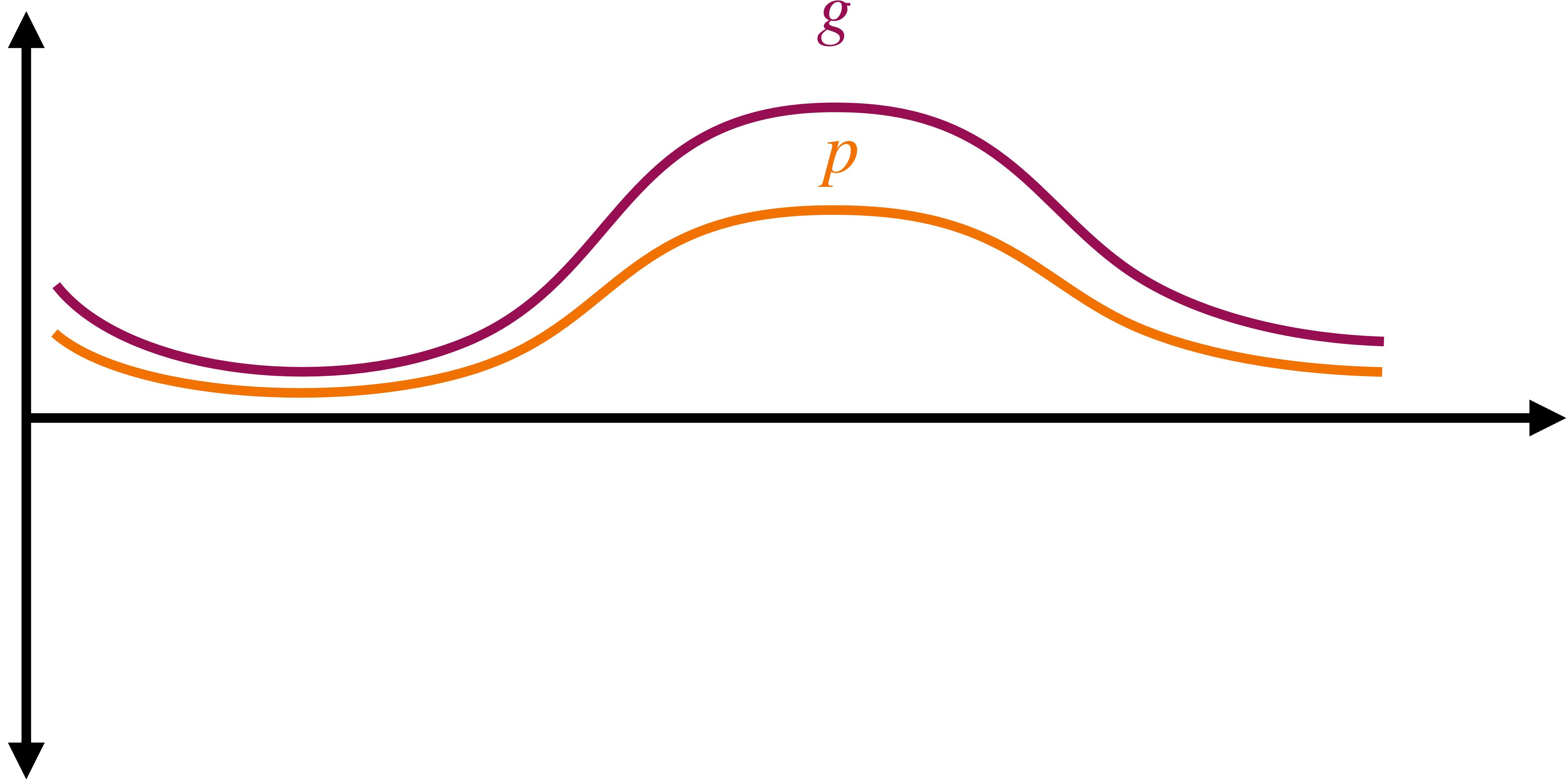
Importance sampling PDF (p) \propto BRDF (g)



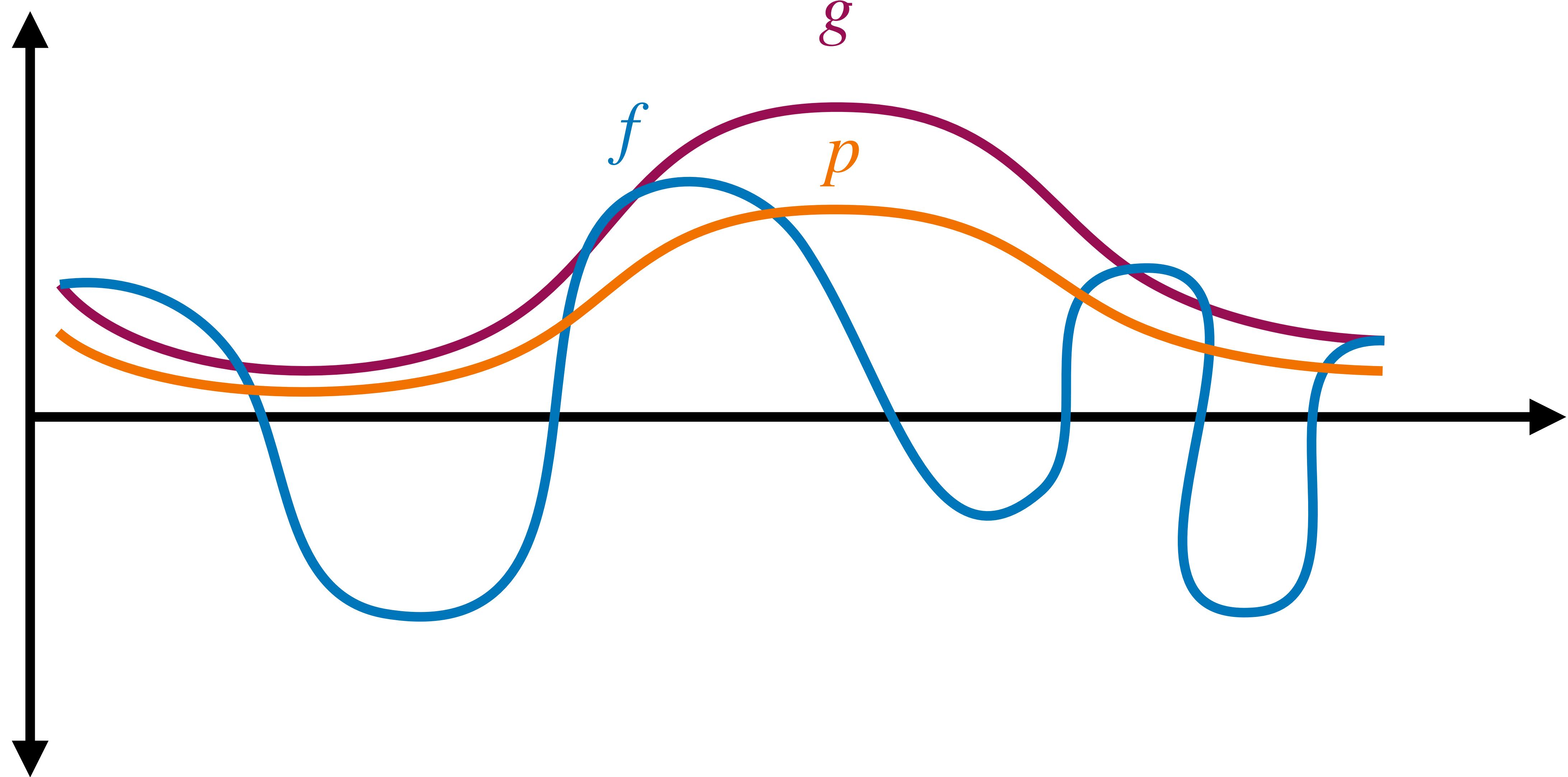
Importance sampling PDF (p) \propto BRDF (g)



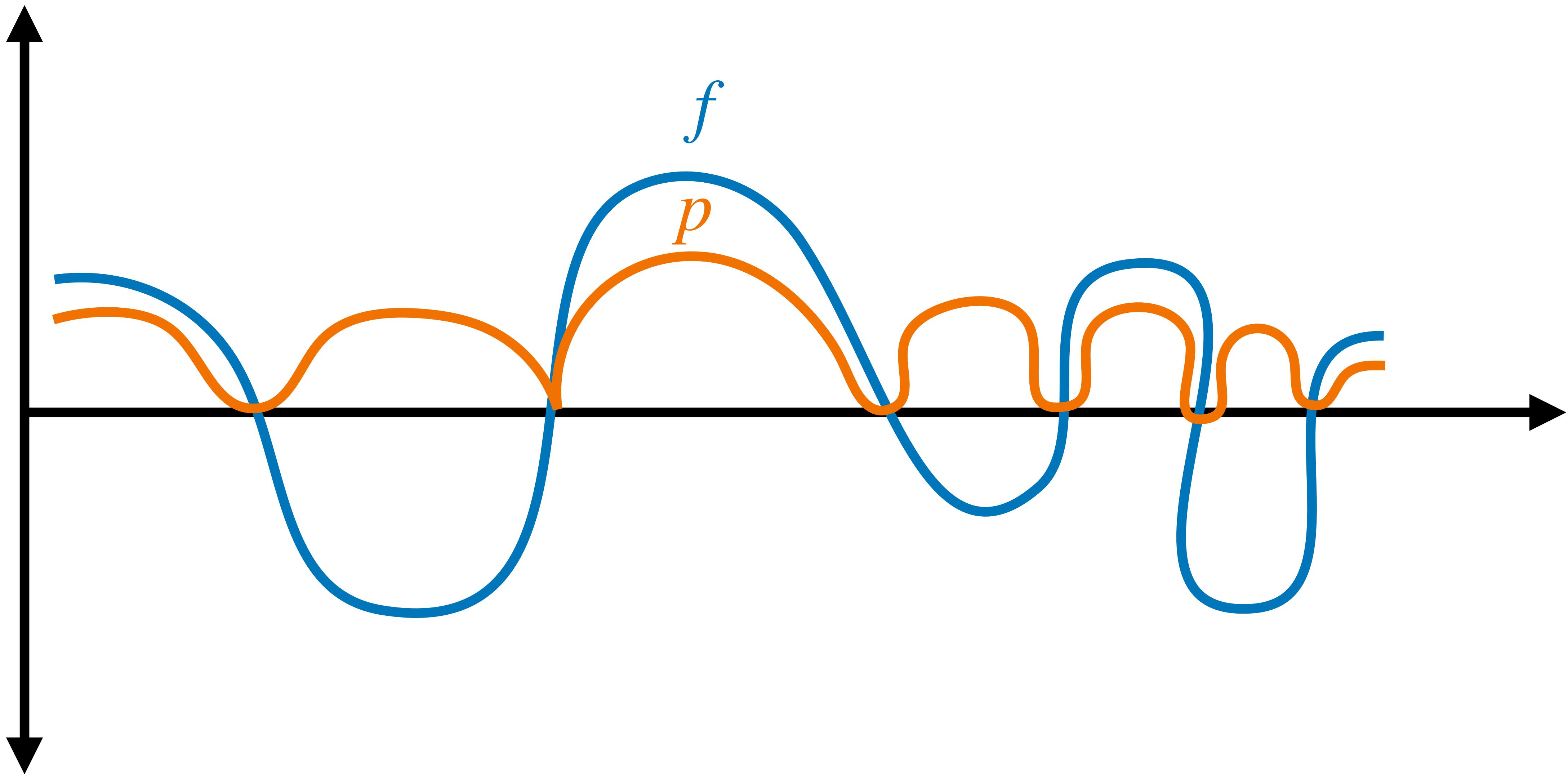
Same PDF to importance sample BRDF derivative?



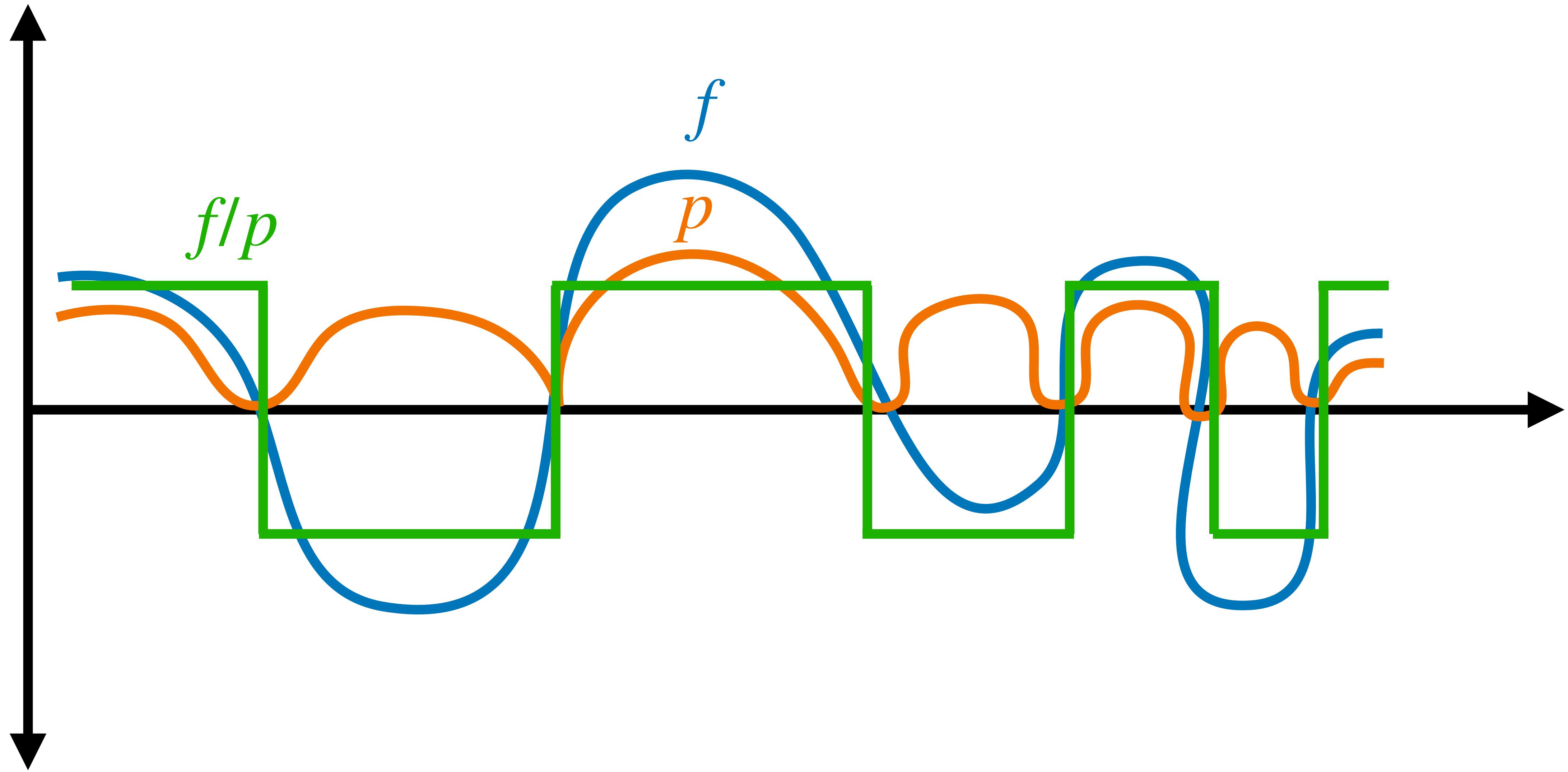
BRDF g is very different from its derivative f



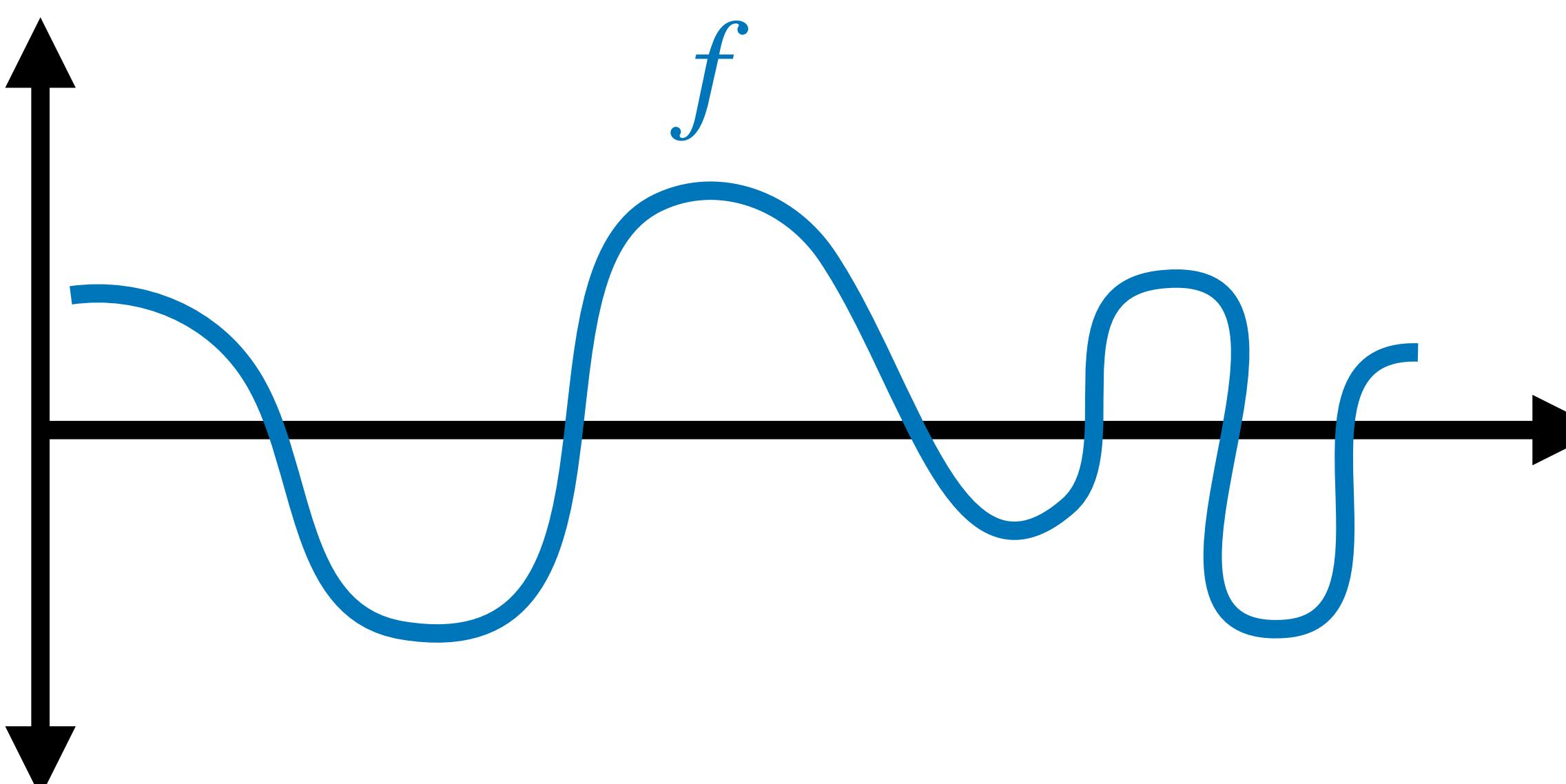
We can construct $p \propto |f|$?



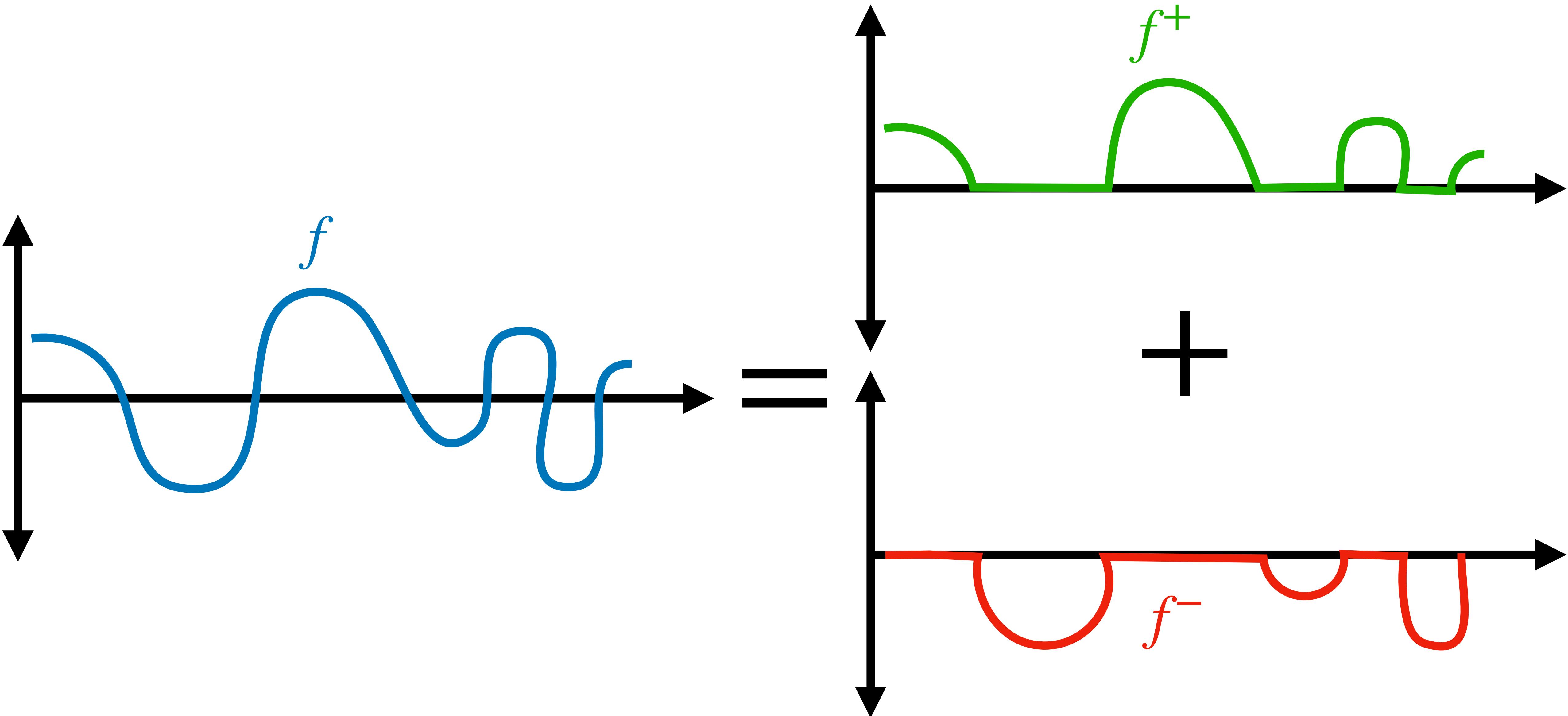
$p \propto |f|$ has sign variance



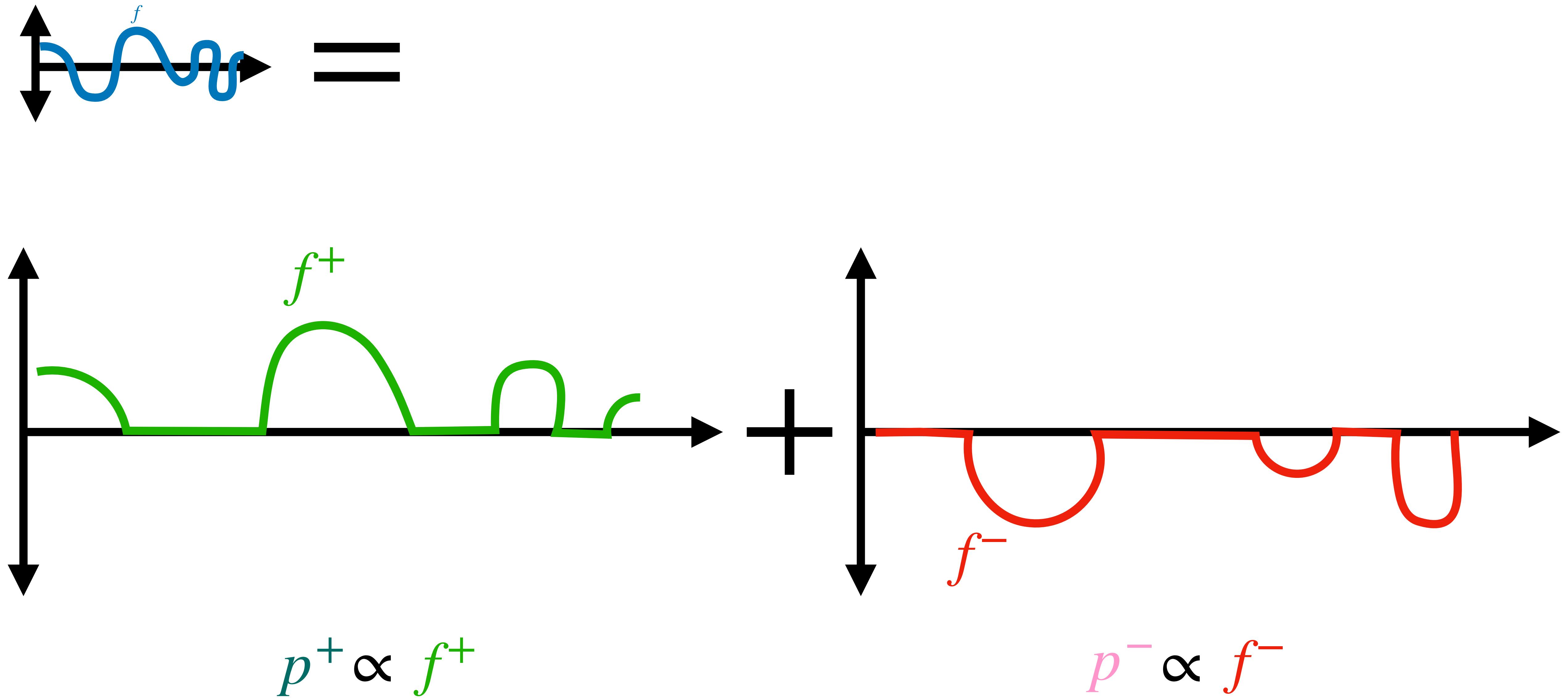
Key Idea: decompose **real-valued into sum of **positive** and **negative****



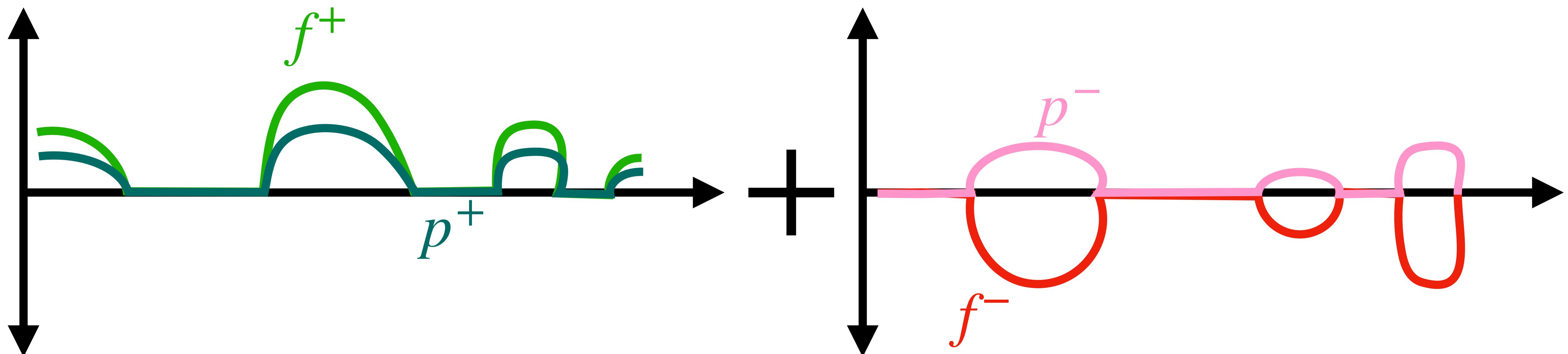
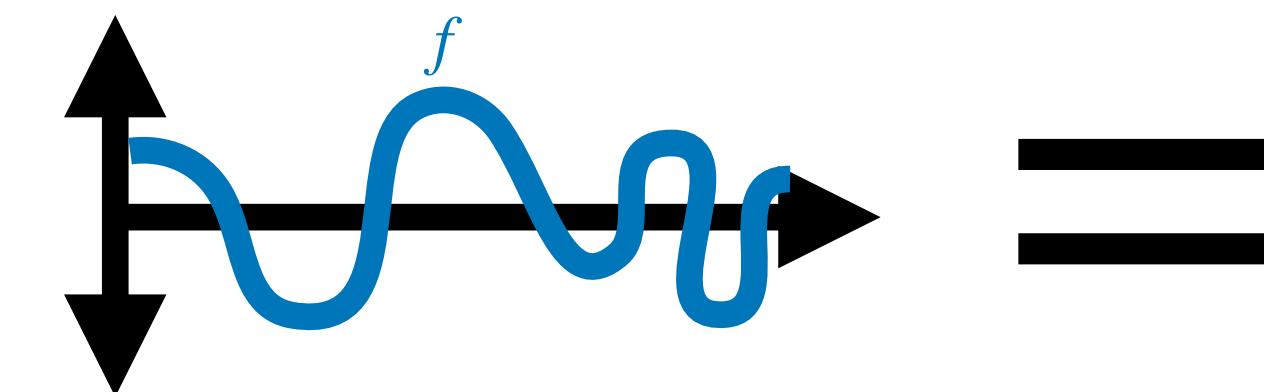
Key Idea: decompose **real-valued into sum of **positive** and **negative****



Positivization – Owen and Zhou 2000.



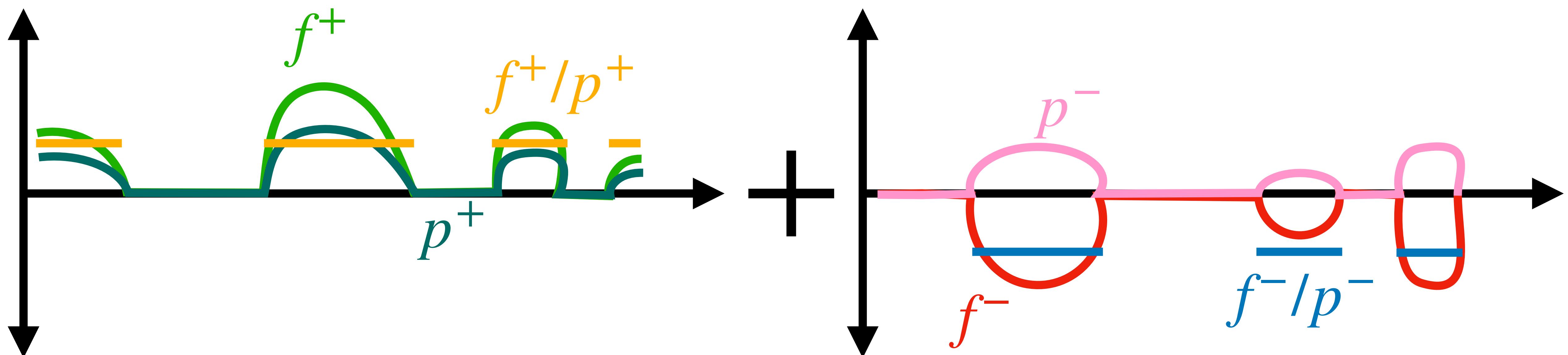
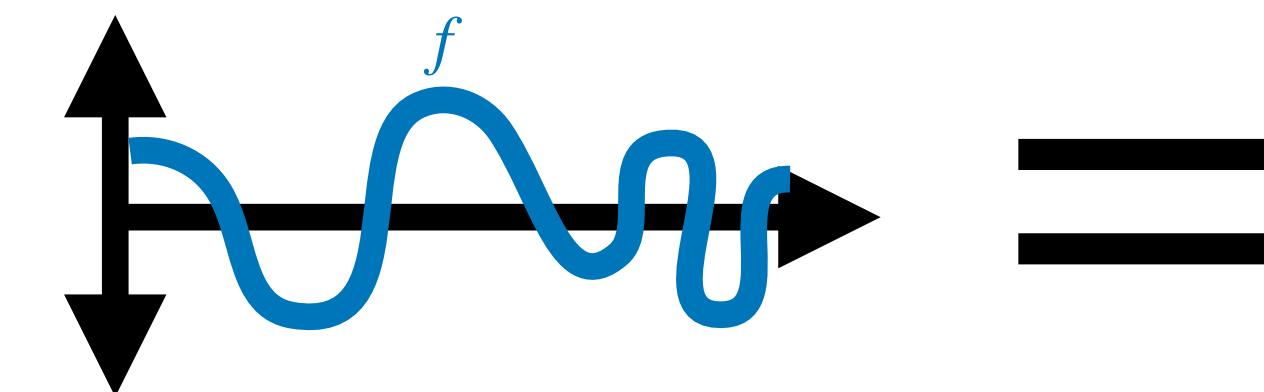
Positivization – Owen and Zhou 2000.



$$p^+ \propto f^+$$

$$p^- \propto f^-$$

Positivization – Owen and Zhou 2000.



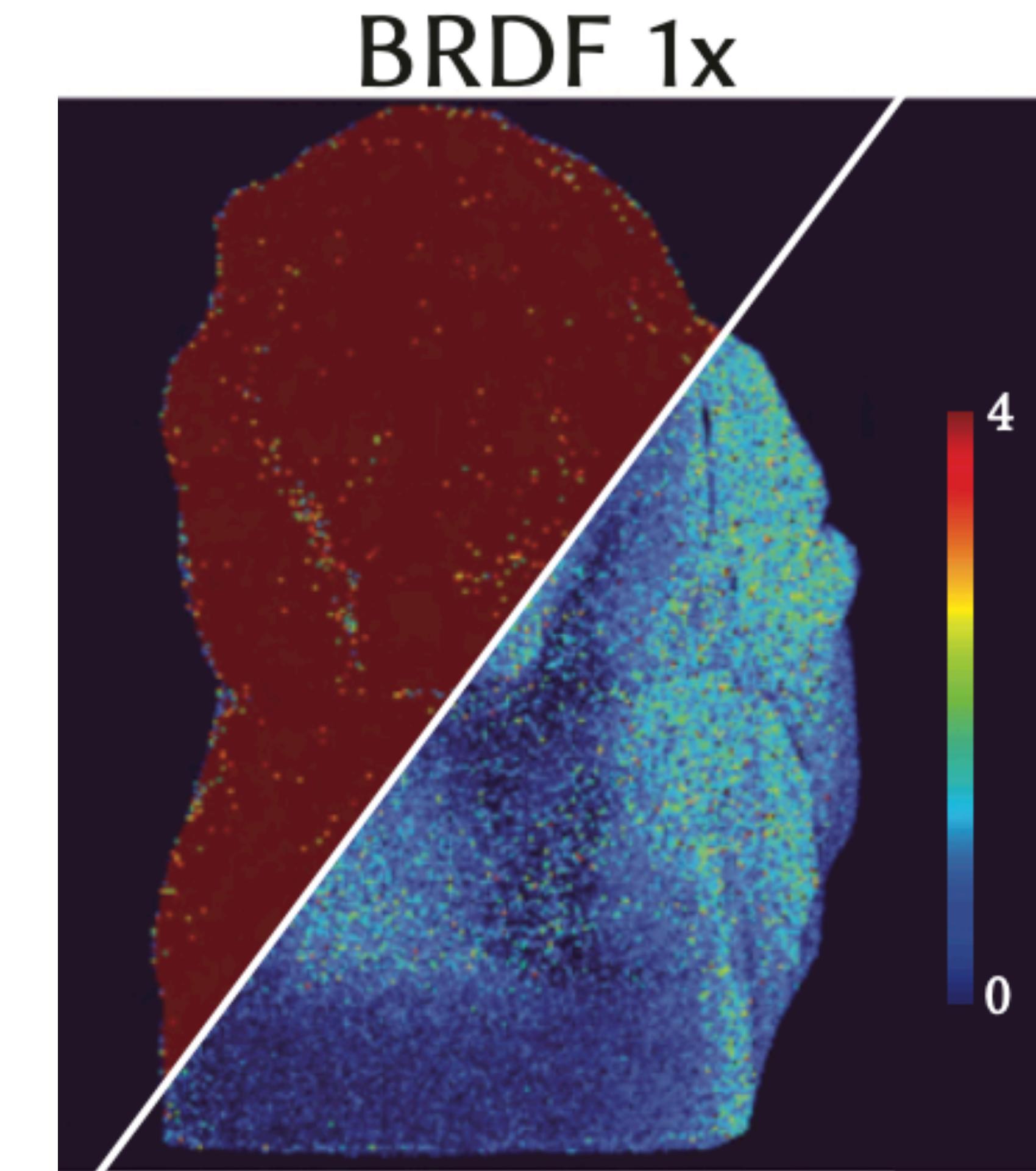
$$p^+ \propto f^+$$

$$p^- \propto f^-$$

Positivization reduces variance by 58x!



Forward rendering

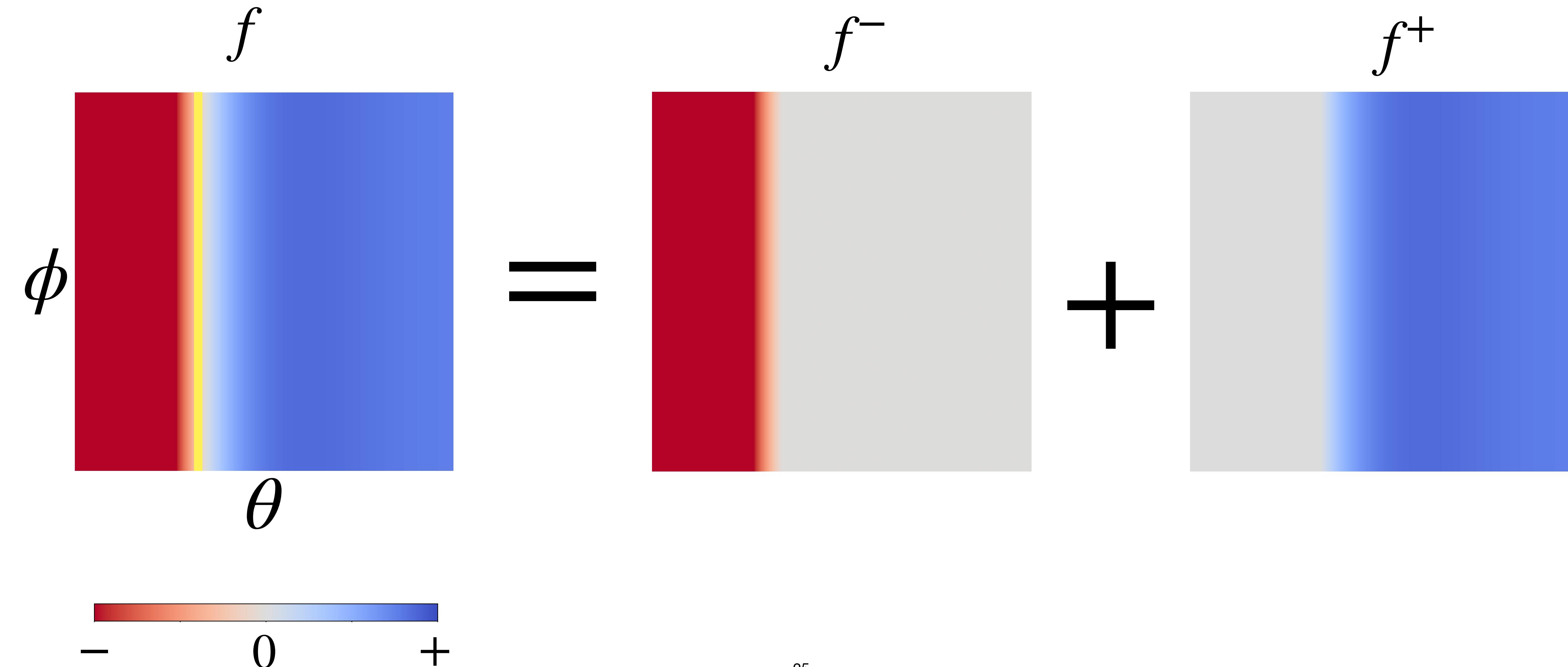


Variance

Positivization is applicable to several BRDF derivatives

- Roughness of isotropic **GGX, Beckmann**
- Exponent of **Blinn-Phong**
- Scattering parameter of **Hanrahan-Krueger BRDF**

Positivization requires analytic root locations

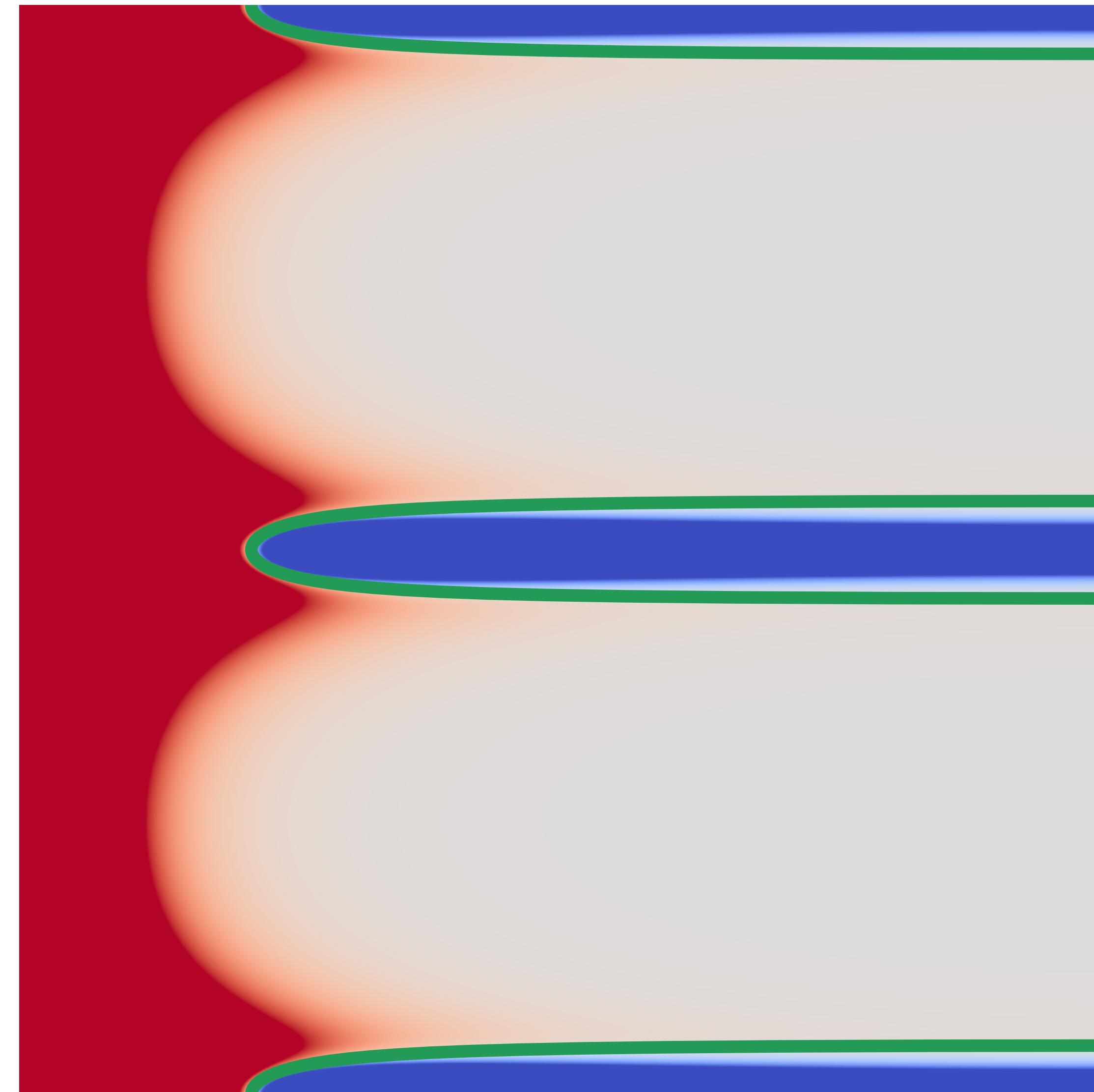


Positivization requires analytic integrability

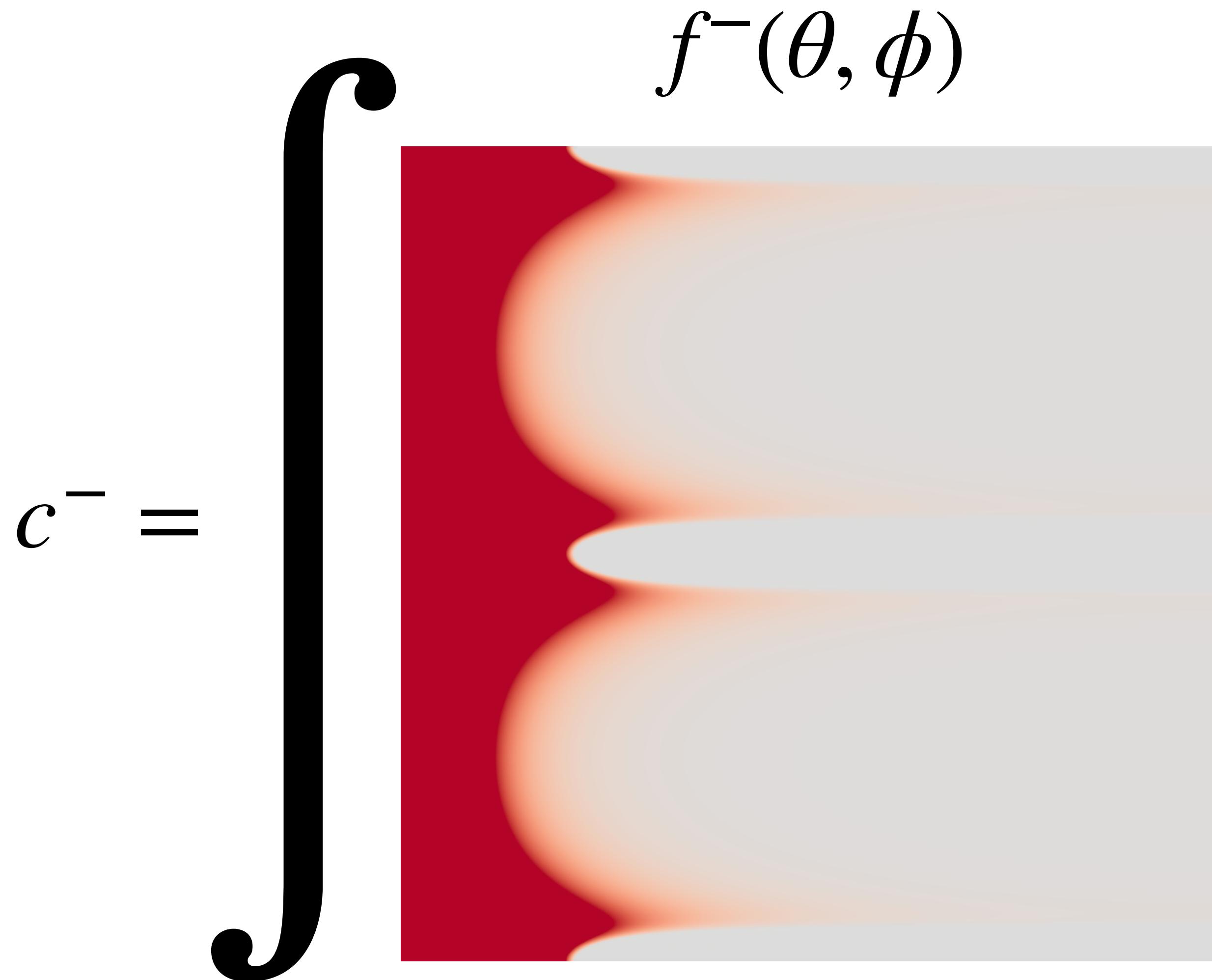
$$c^- p^-(\theta, \phi) = f^-(\theta, \phi)$$

$$c^- = \int f^-(\theta, \phi)$$


No analytic root locations – no positivization :(



No analytic integrability – no positivization :(



Positivization **is inapplicable** to several BRDF derivatives :(

- Directional roughness of anisotropic **Beckmann and GGX**
- Directional exponent of **Ashikhmin-Shirley**
- Width of **Burley's BSSRDF**
- Weights of mixture BRDFs
 - All layered BRDFs (**Disney Principled, Autodesk Standard Surface, etc.**)
 - **Oren-Nayar**
 - **Microcylinder BRDF**
- and many others...

Our product and **mixture** decomposition can handle these!

- Directional roughness of anisotropic **Beckmann and GGX**
- Directional exponent of **Ashikhmin-Shirley**
- Width of **Burley's BSSRDF**

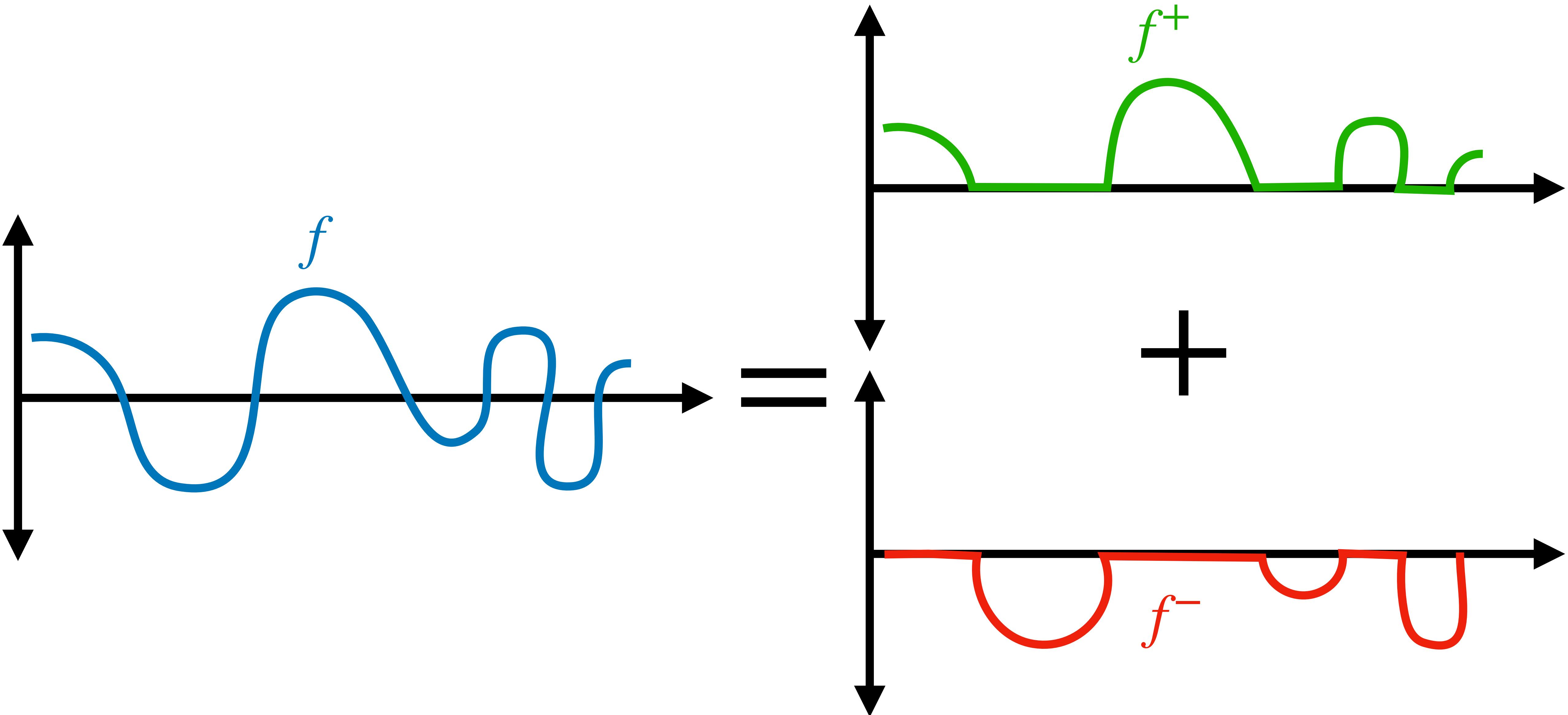
Product decomposition

- Weights of mixture BRDFs
 - All layered BRDFs (**Disney Principled, Autodesk Standard Surface, etc.**)
 - **Oren-Nayar**
 - **Microcylinder BRDF**
 - and many others...

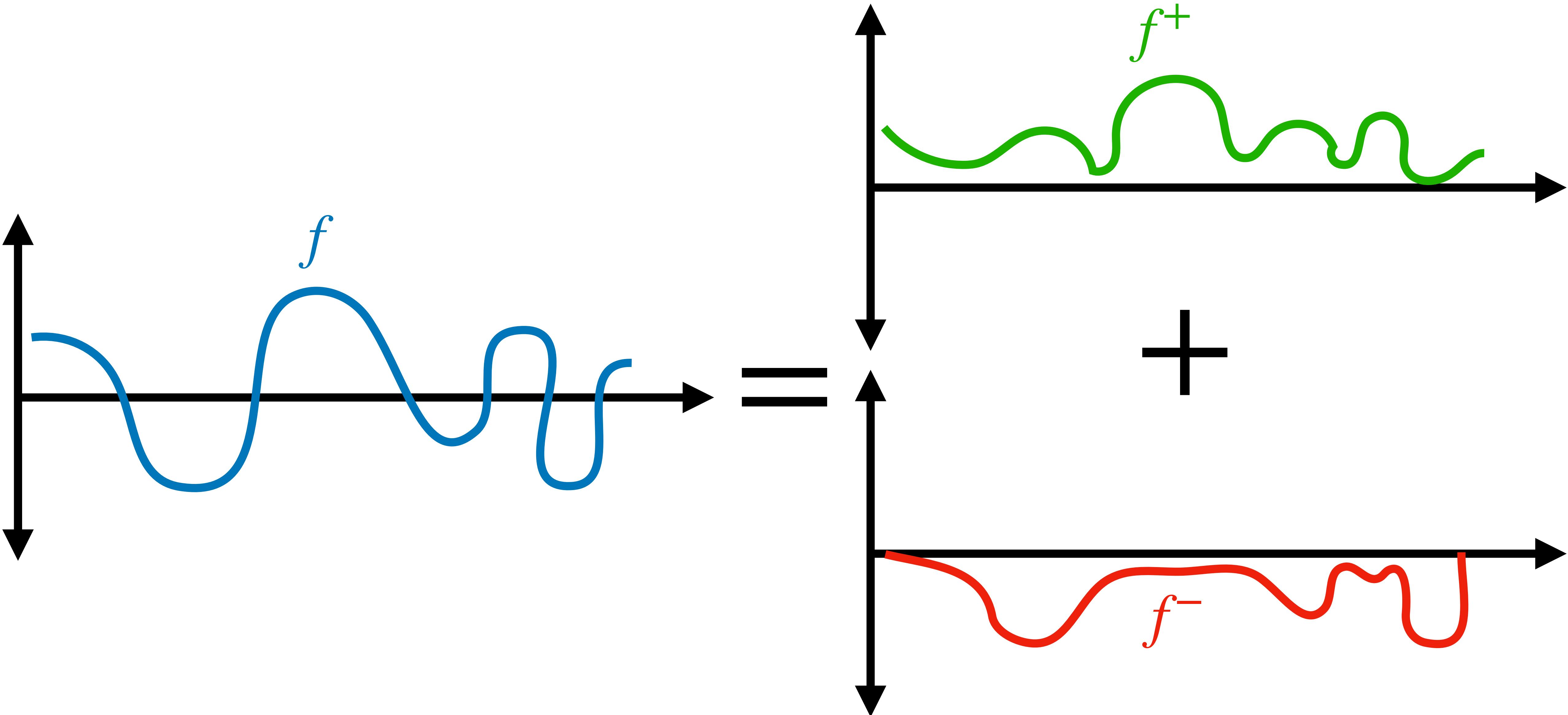
Mixture decomposition

**Key idea: let positive and negative parts
overlap!**

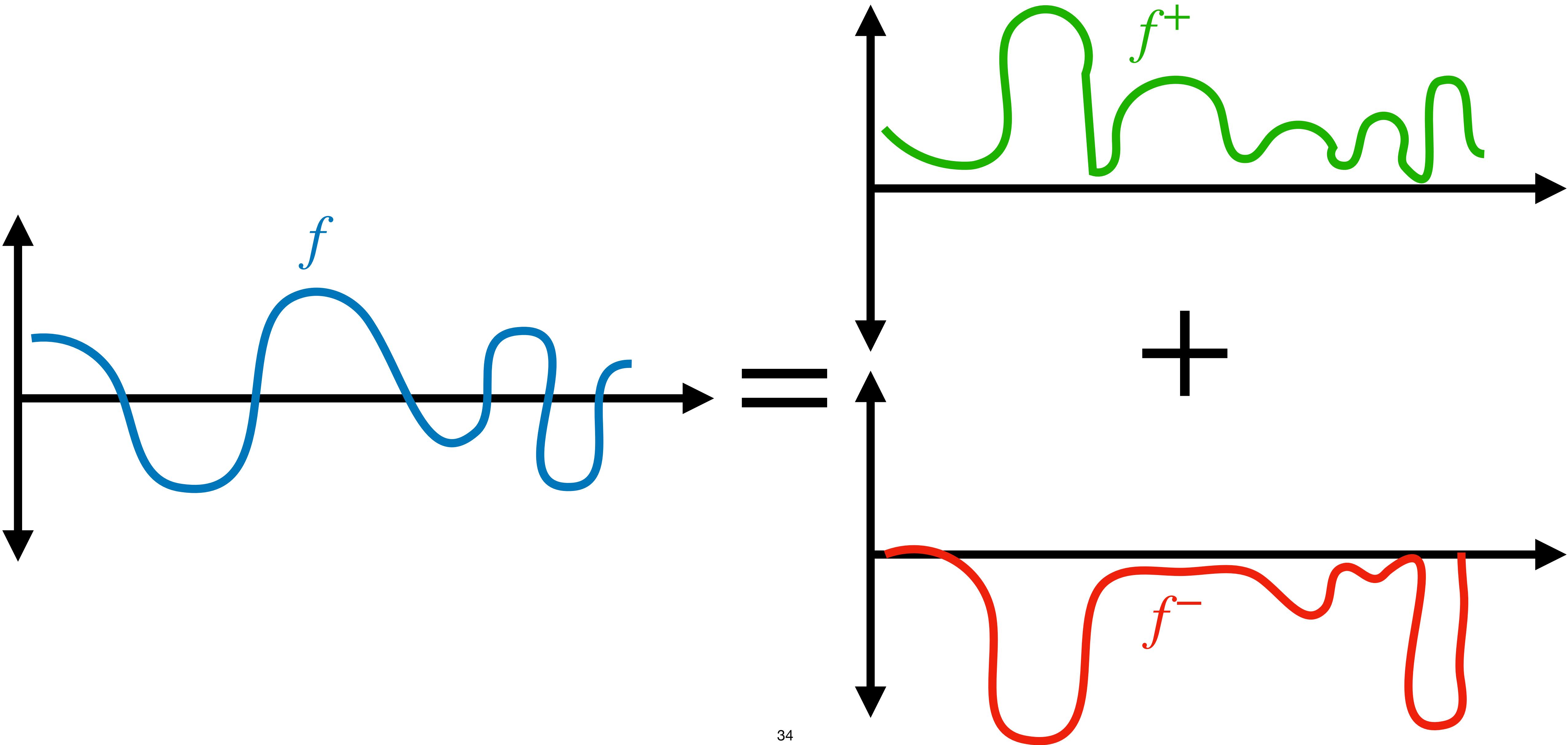
Recall: Positivization has non-overlapping support



A decomposition with overlapping support



Another decomposition with overlapping support



Our Mixture Decomposition

$$g = \beta g_s + (1 - \beta) g_d$$

Our Mixture Decomposition

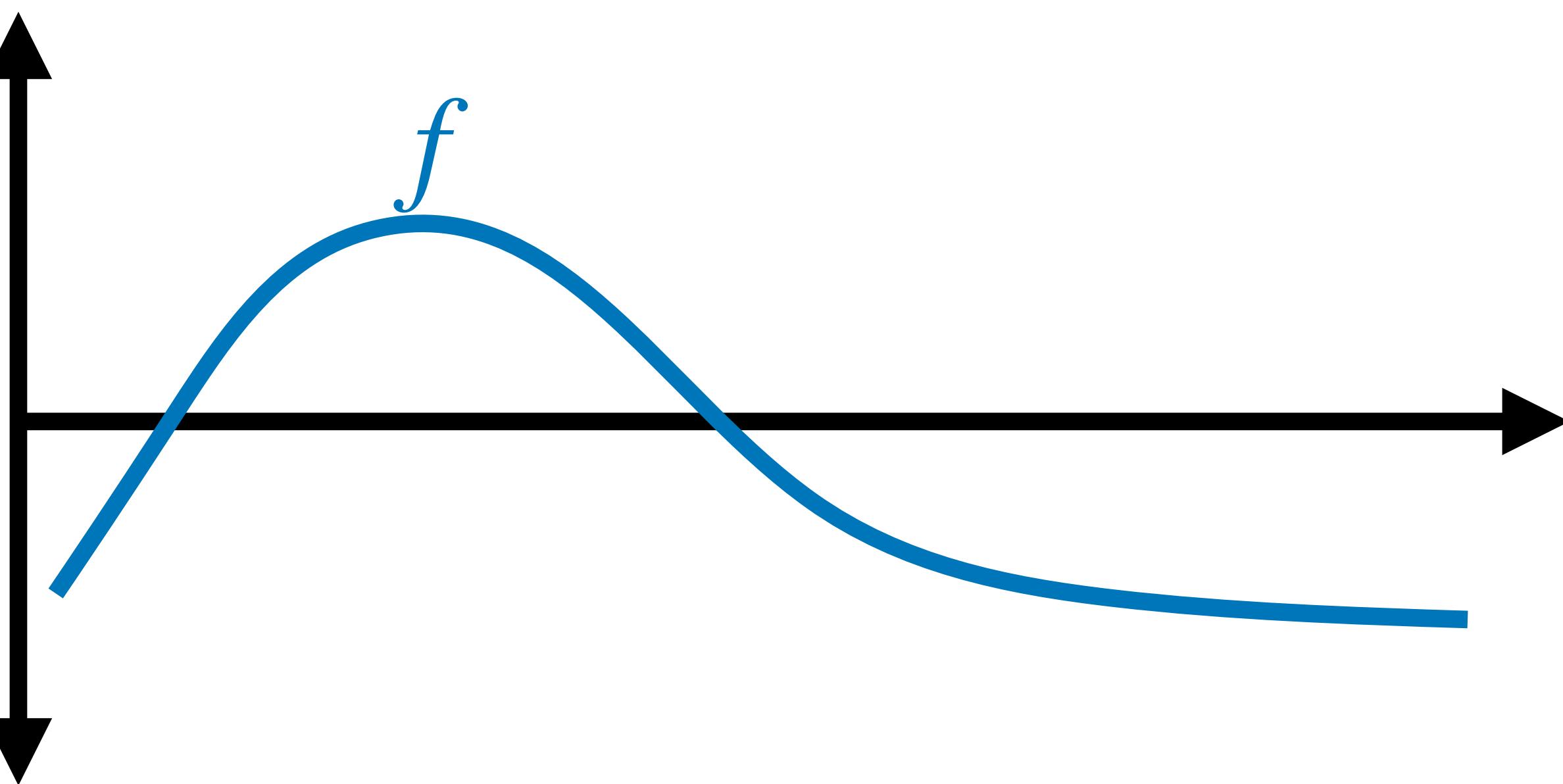
$$g = \beta g_s + (1 - \beta) g_d$$

$$\partial_\beta g = f = g_s - g_d$$

Our Mixture Decomposition

$$g = \beta g_s + (1 - \beta) g_d$$

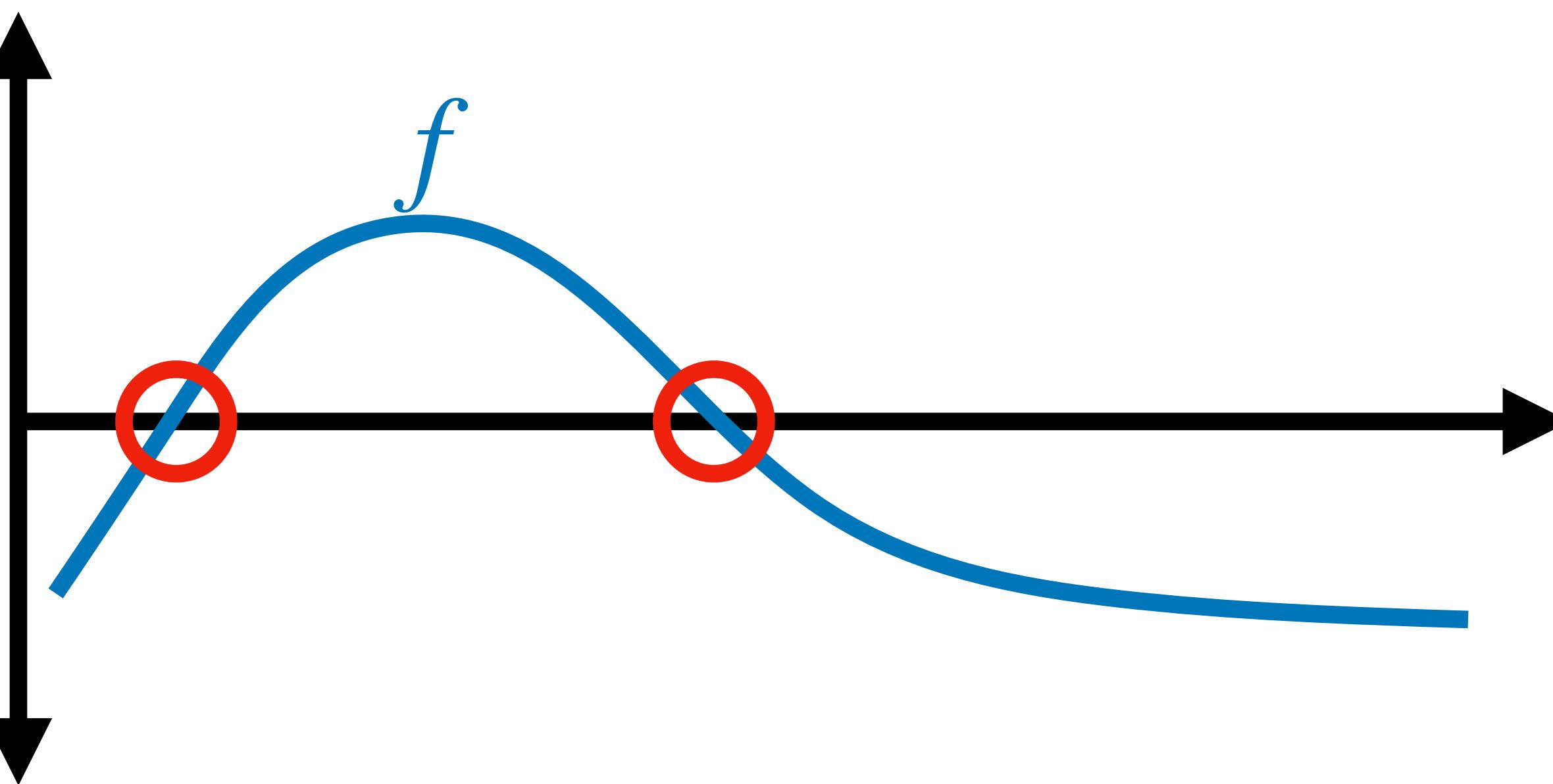
$$\partial_\beta g = f = g_s - g_d$$



Our Mixture Decomposition

$$g = \beta g_s + (1 - \beta) g_d$$

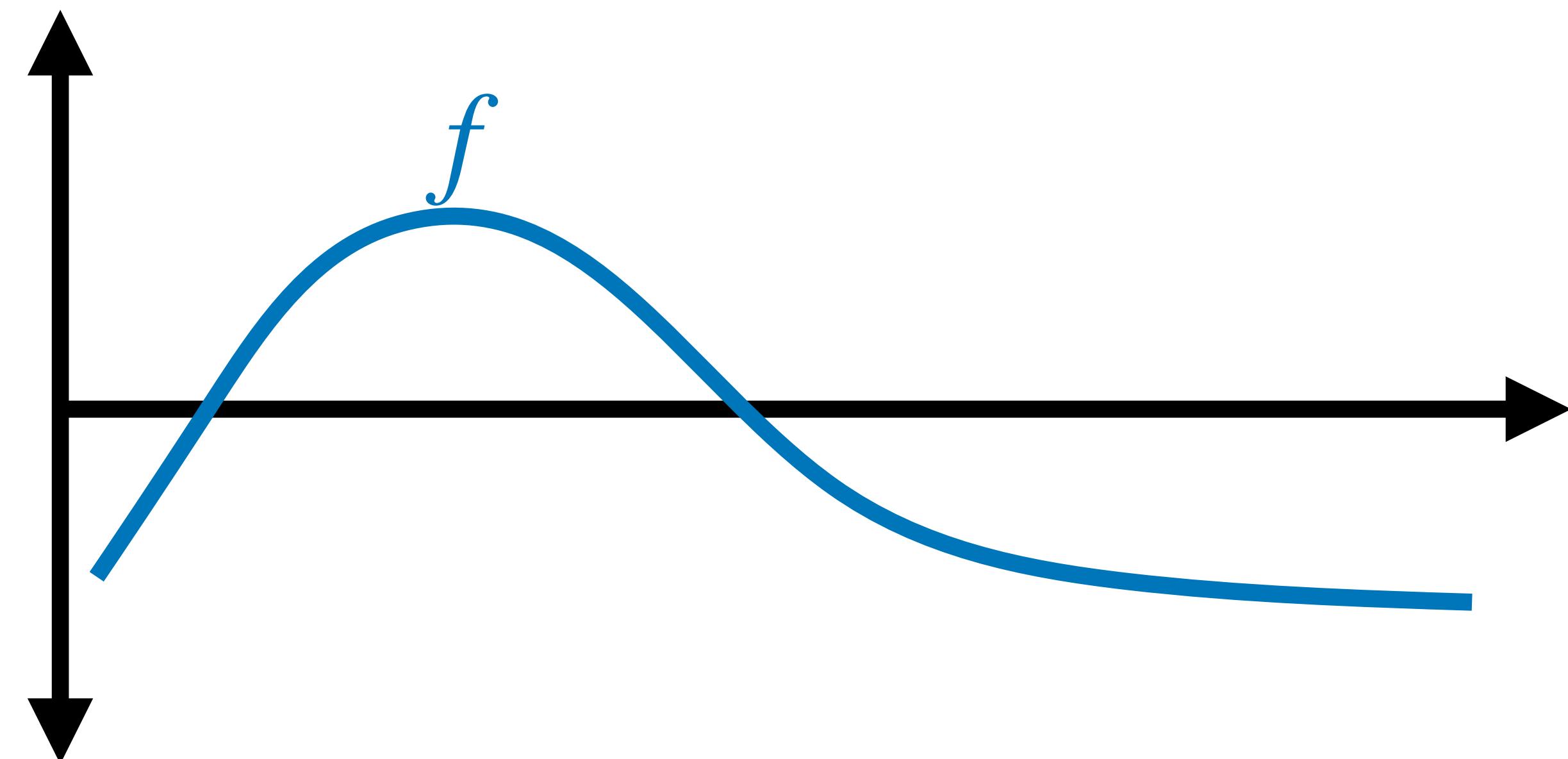
$$\partial_\beta g = f = g_s - g_d$$



Our Mixture Decomposition

$$g = \beta g_s + (1 - \beta) g_d$$

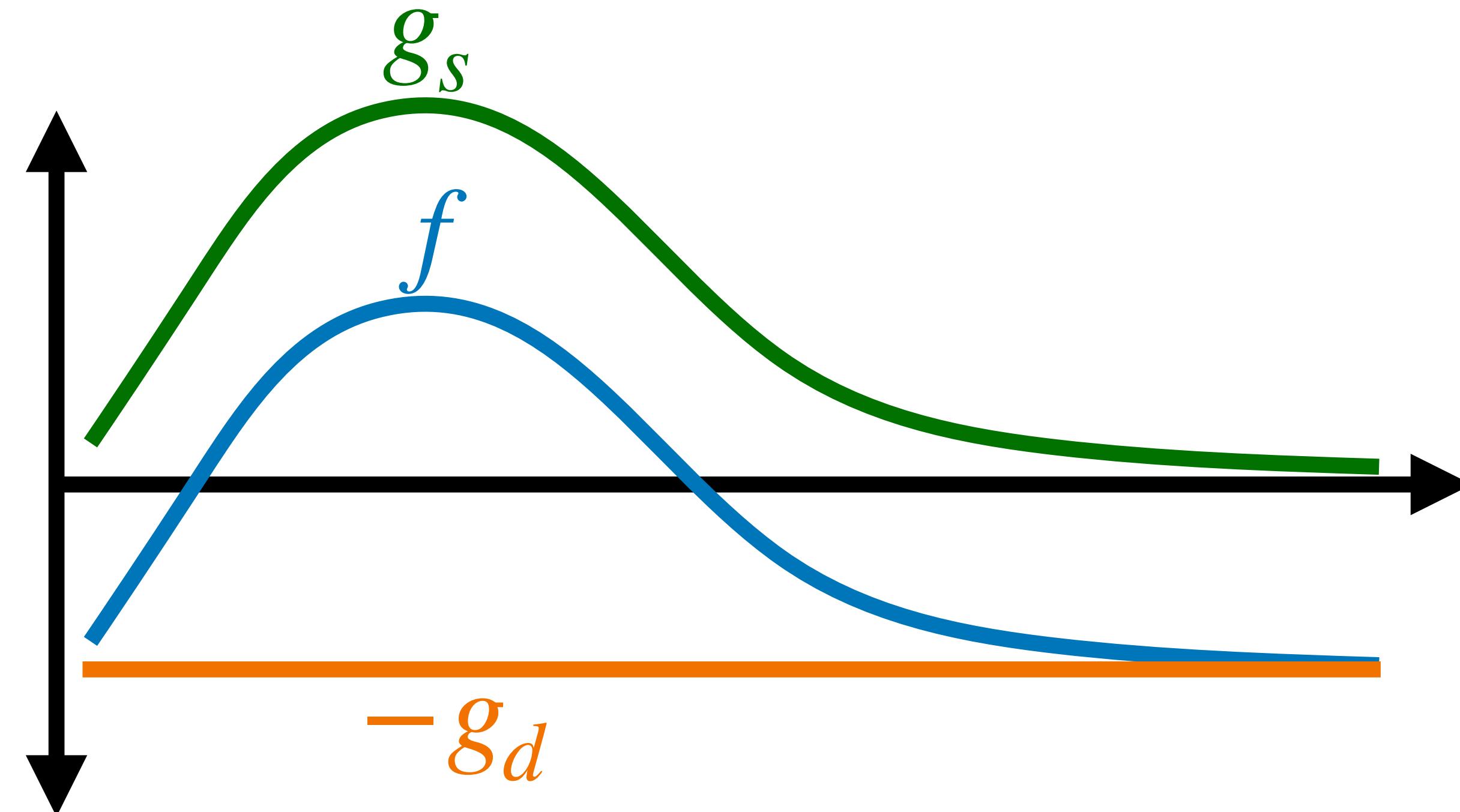
$$\partial_\beta g = f = g_s - g_d$$



Our Mixture Decomposition

$$g = \beta g_s + (1 - \beta)g_d$$

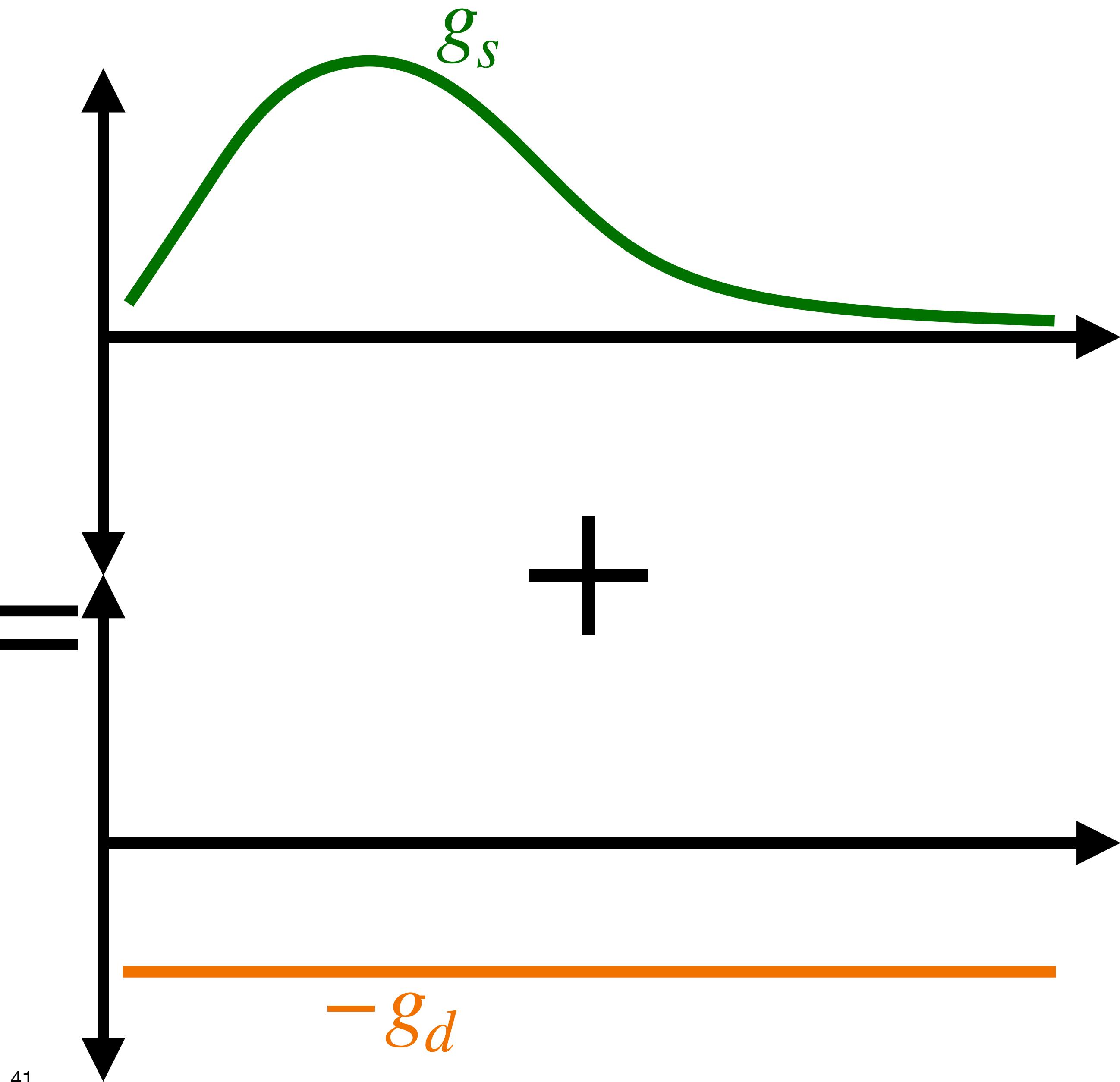
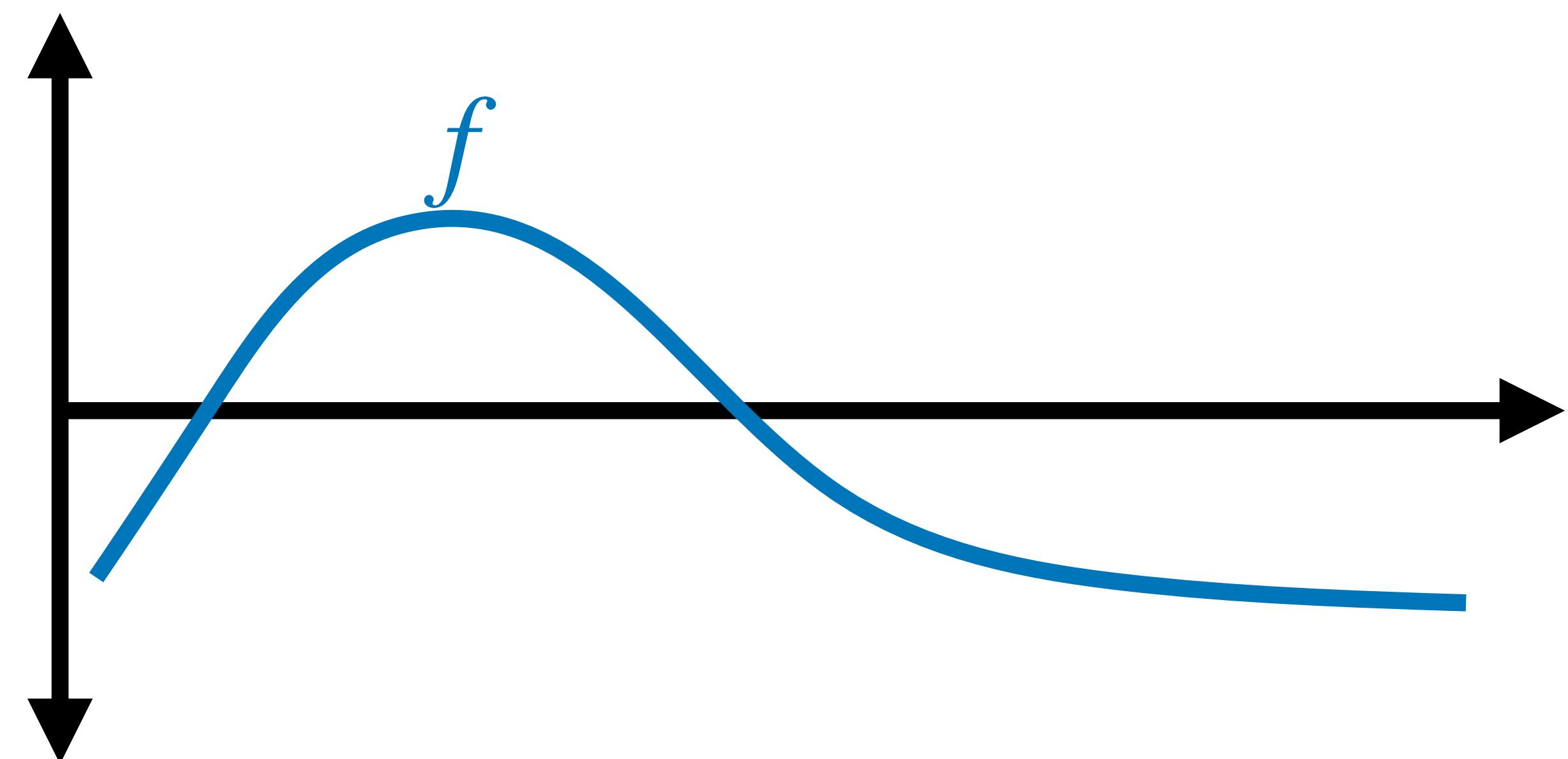
$$\partial_\beta g = f = g_s - g_d$$



Our Mixture Decomposition

$$g = \beta g_s + (1 - \beta)g_d$$

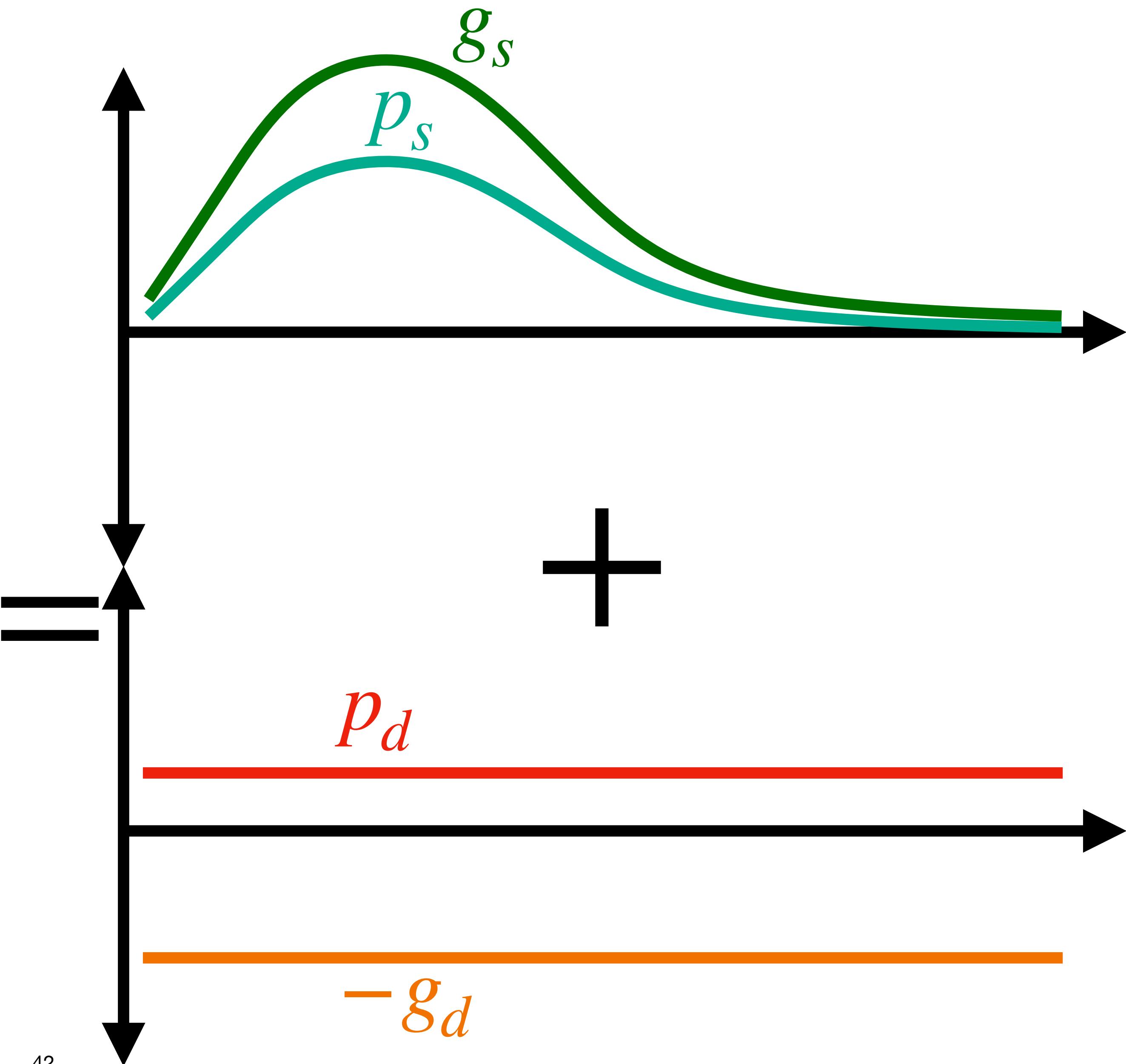
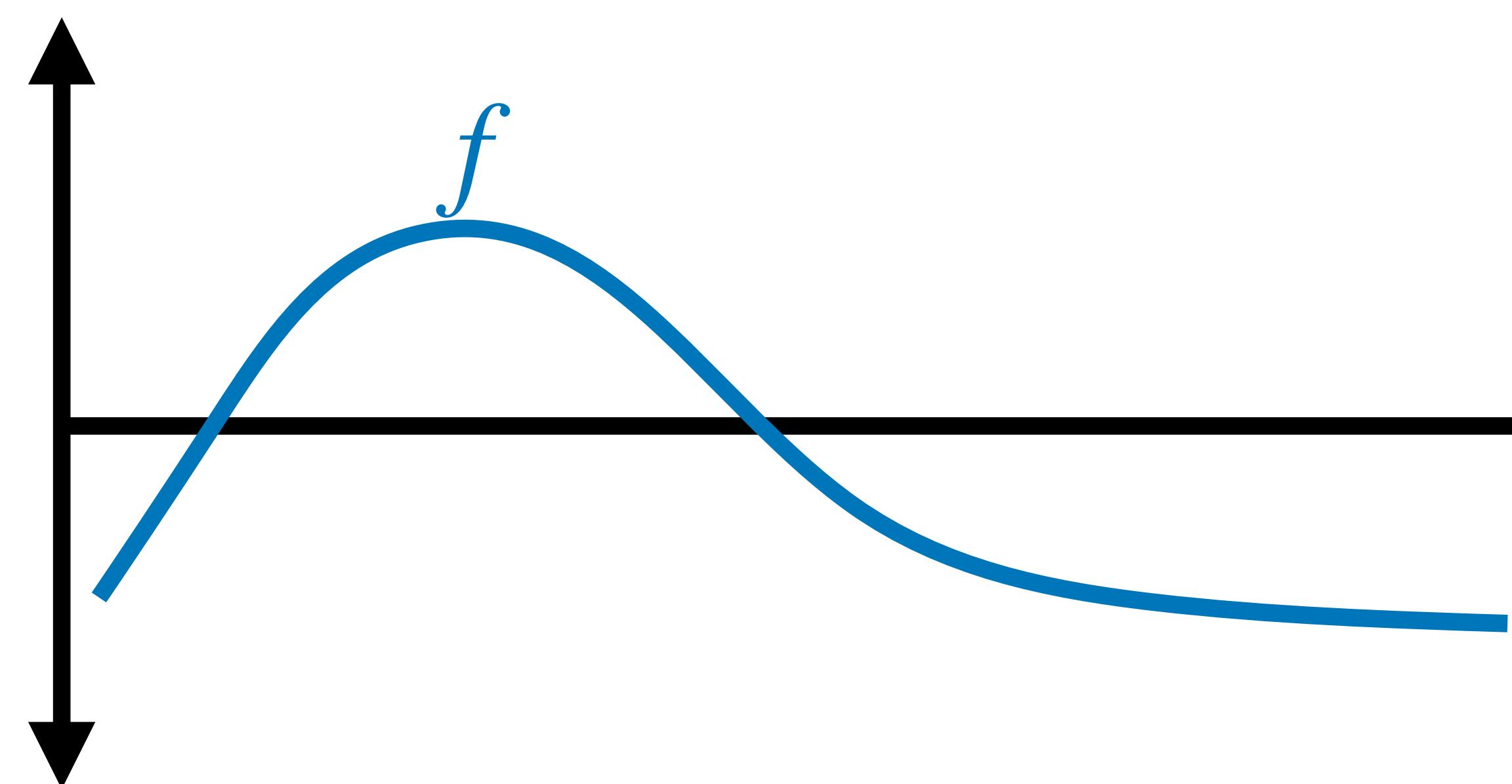
$$\partial_\beta g = f = g_s - g_d$$



Our Mixture Decomposition

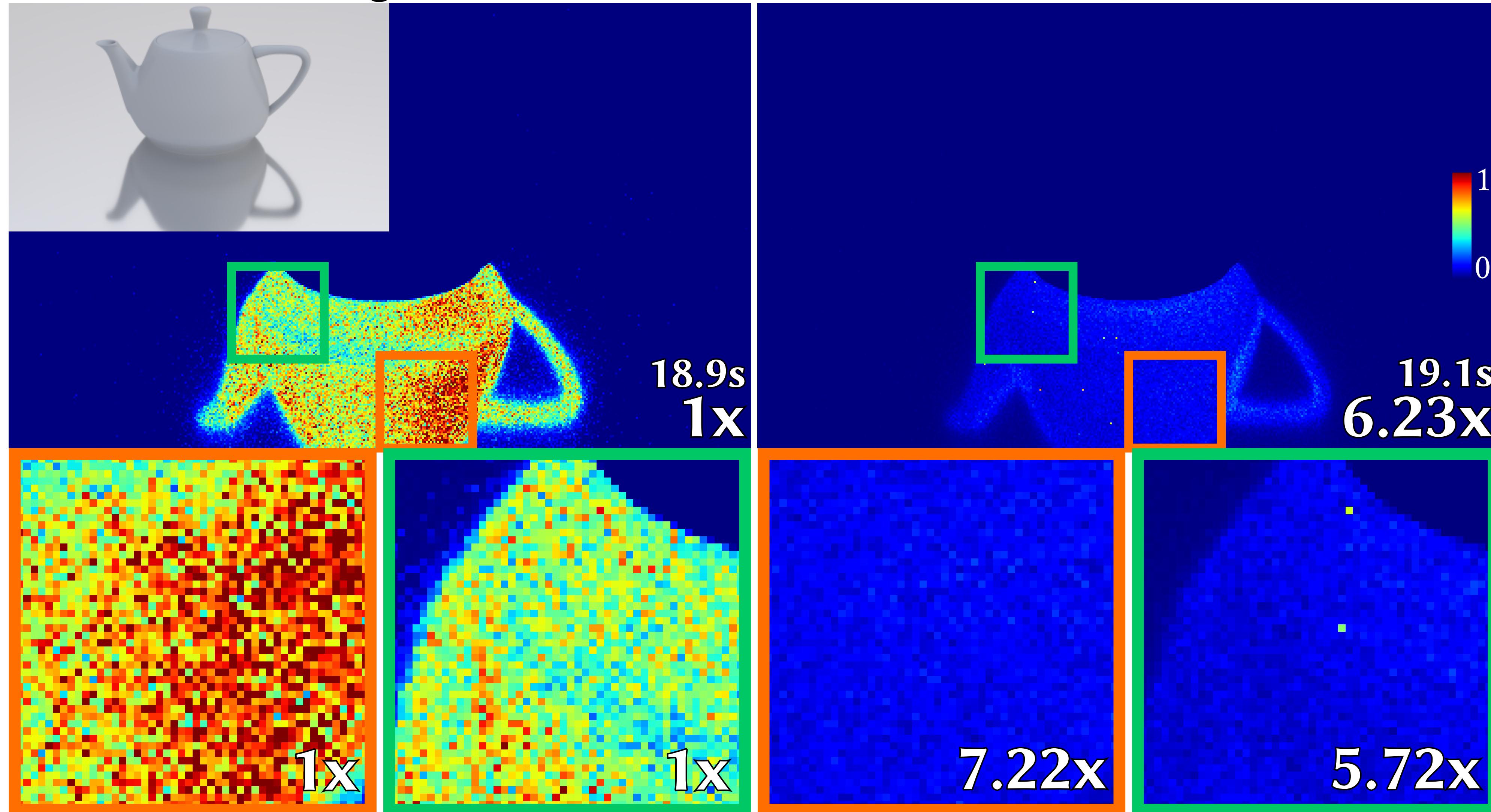
$$g = \beta g_s + (1 - \beta)g_d$$

$$\partial_\beta g = f = g_s - g_d$$



Mixture decomposition reduces variance of glossy reflections

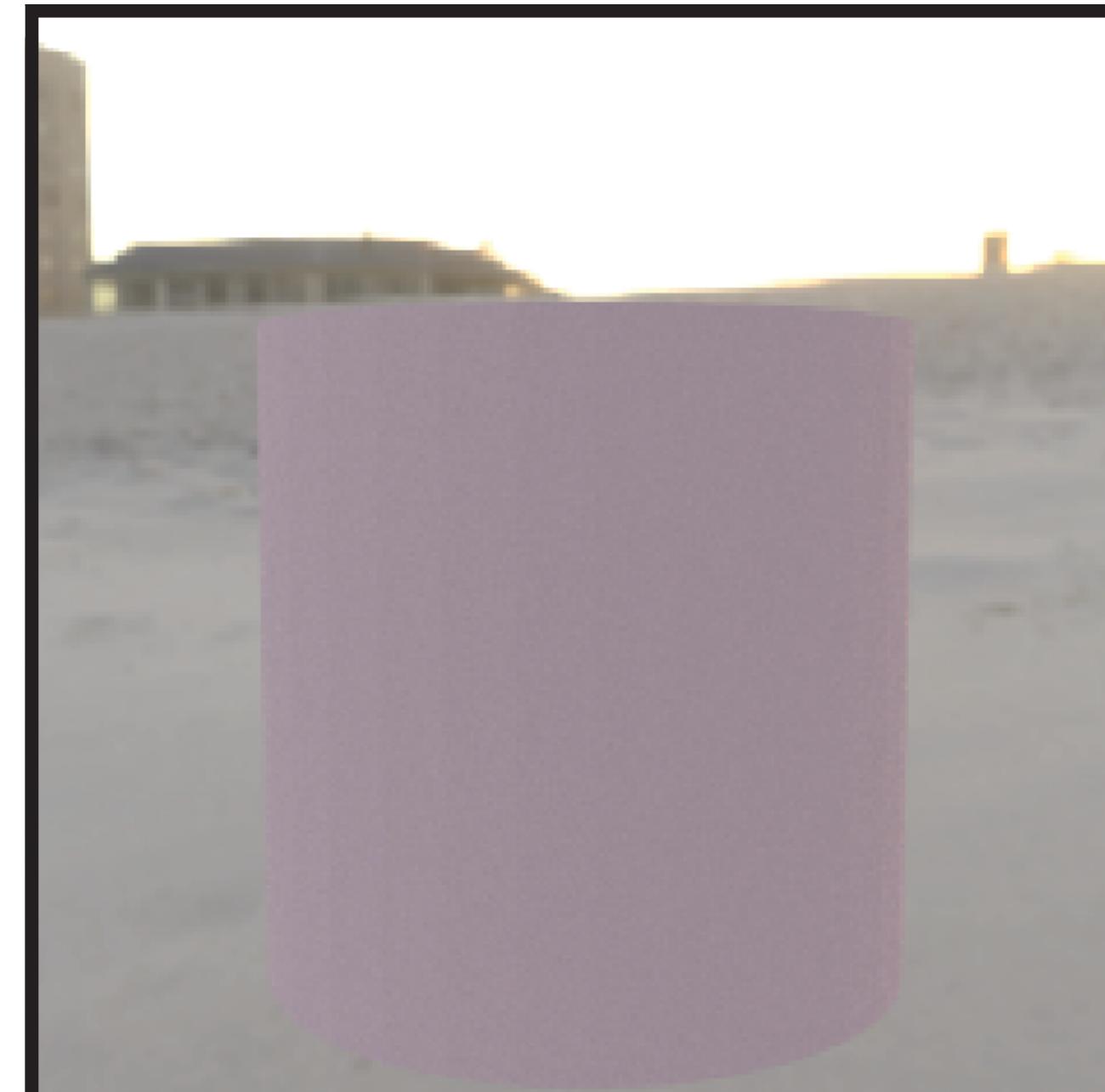
Forward Rendering



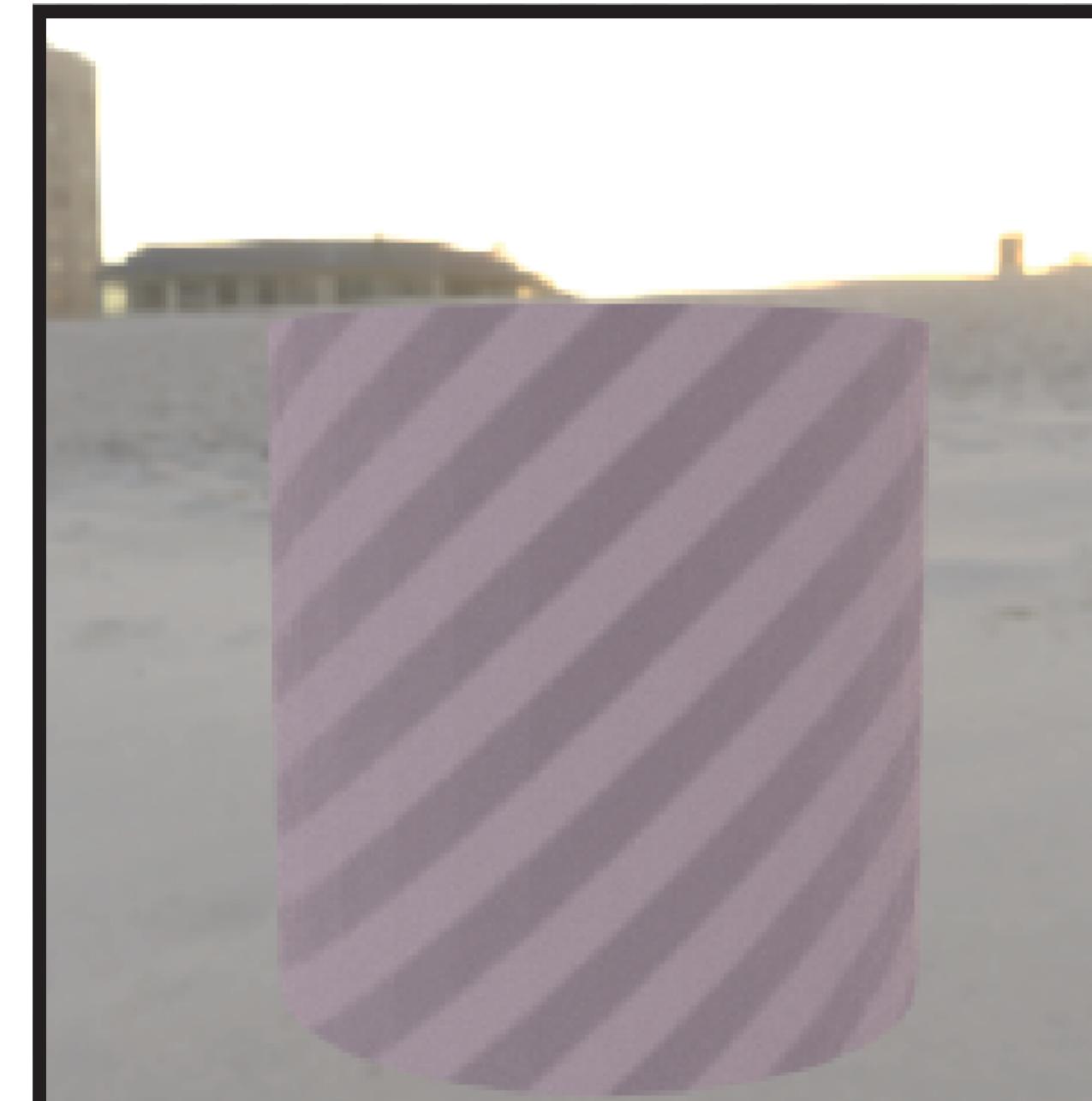
(a) BRDF Sampling

(b) Our Mixture Decomposition w/ MIS

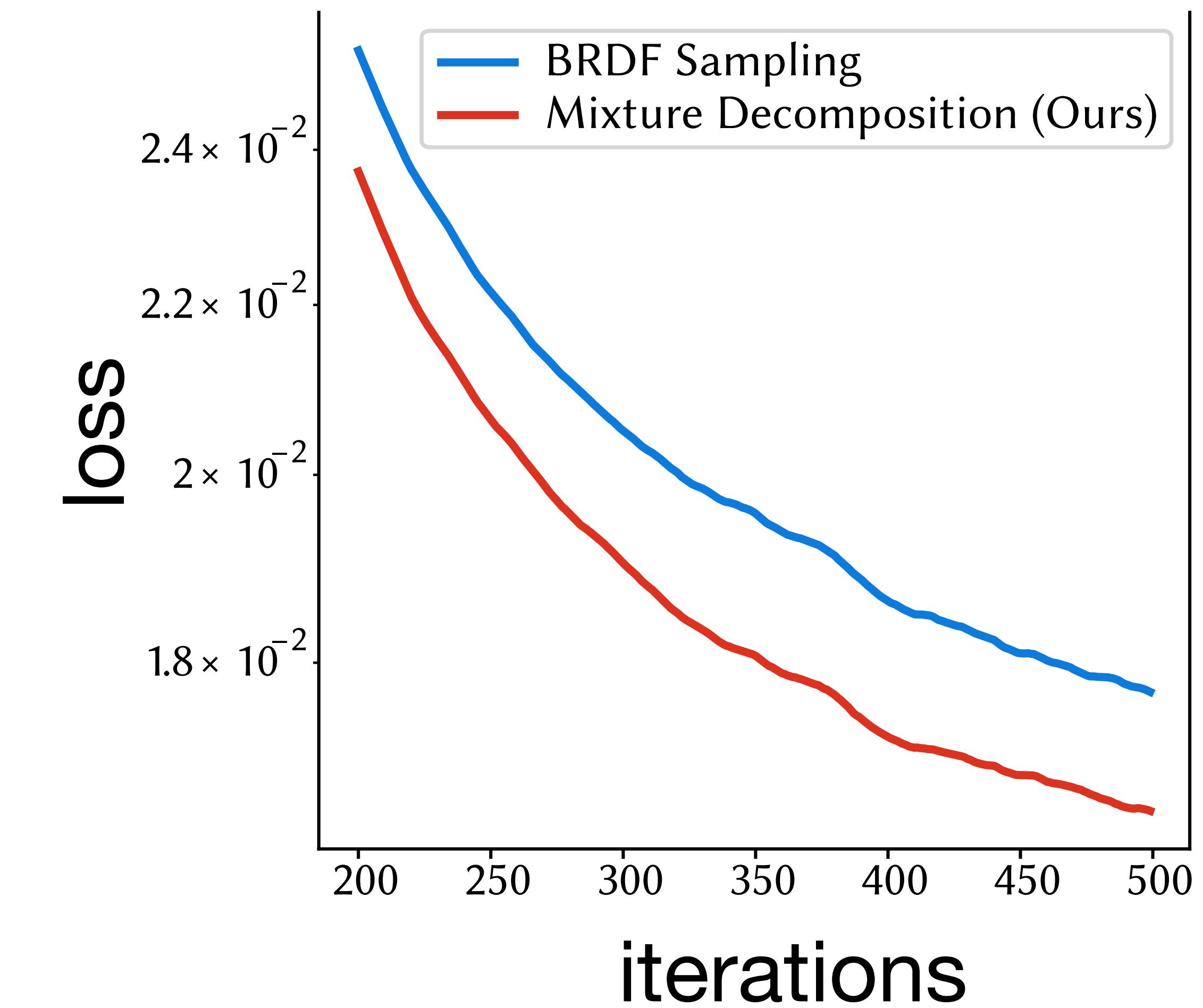
Mixture decomposition improves inverse rendering



initialization



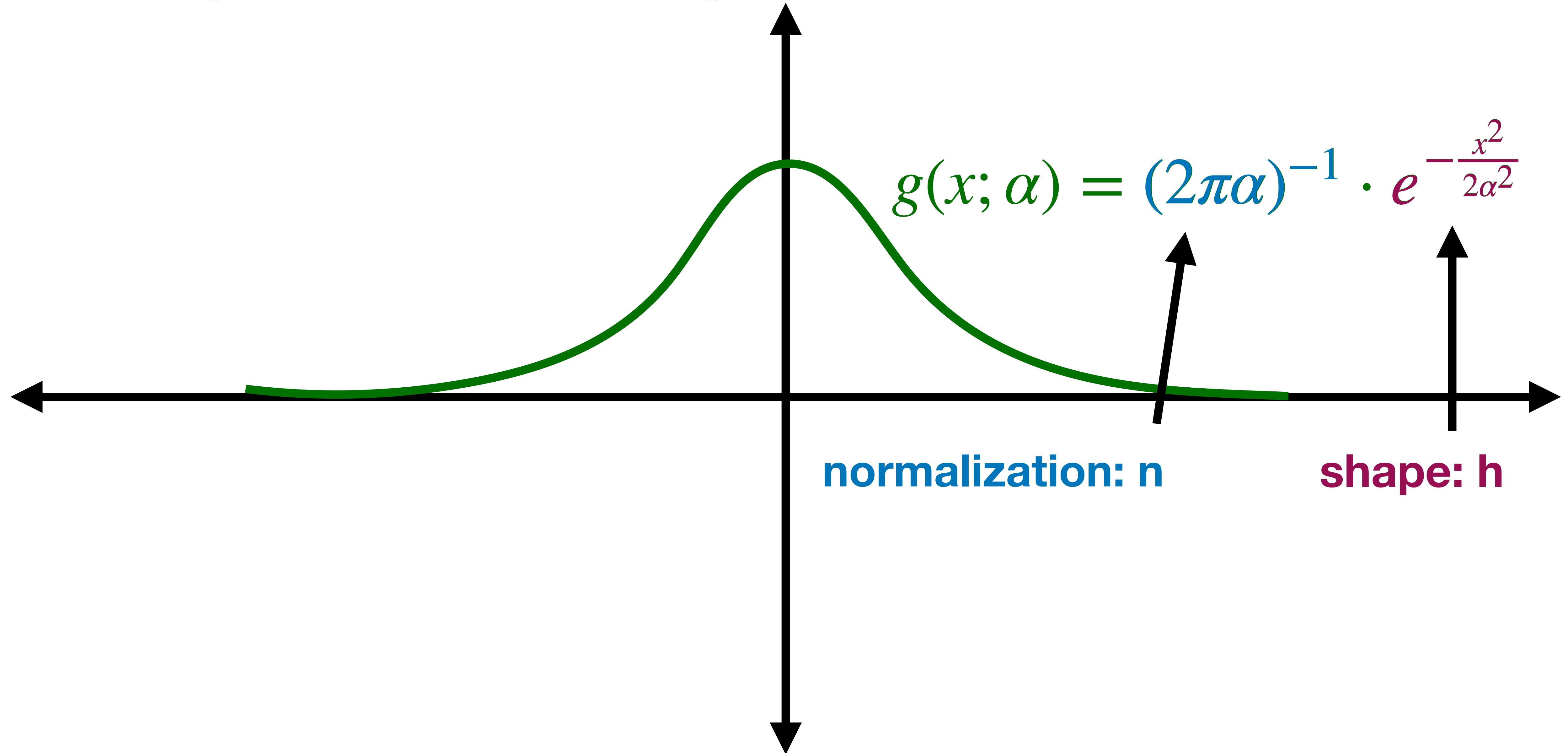
recovery



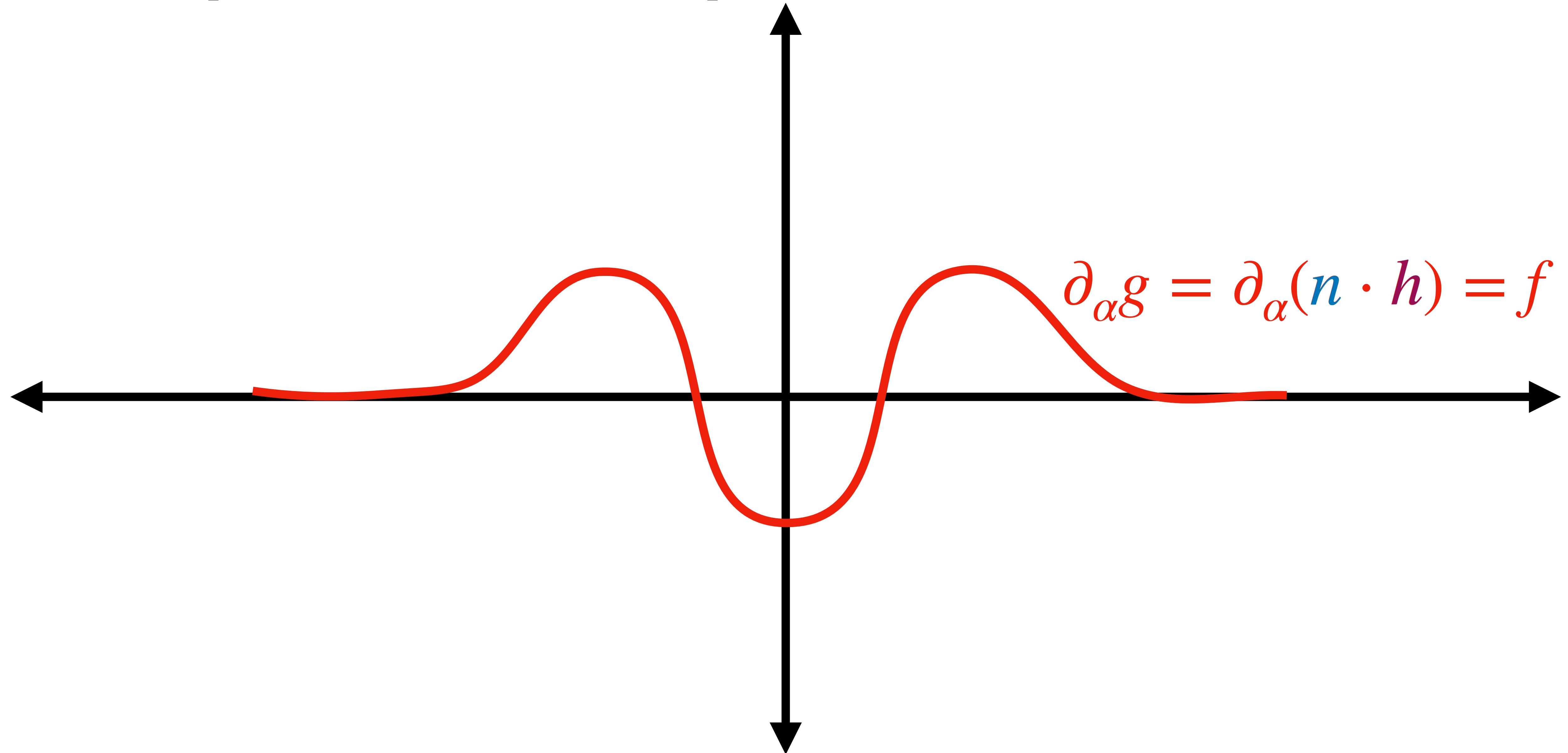
- Weights of mixture BRDFs
 - All layered BRDFs (**Disney Principled, Autodesk Standard Surface, etc.**)
 - **Oren-Nayar**
 - **Microcylinder BRDF**
 - and many others...

Our mixture decomposition

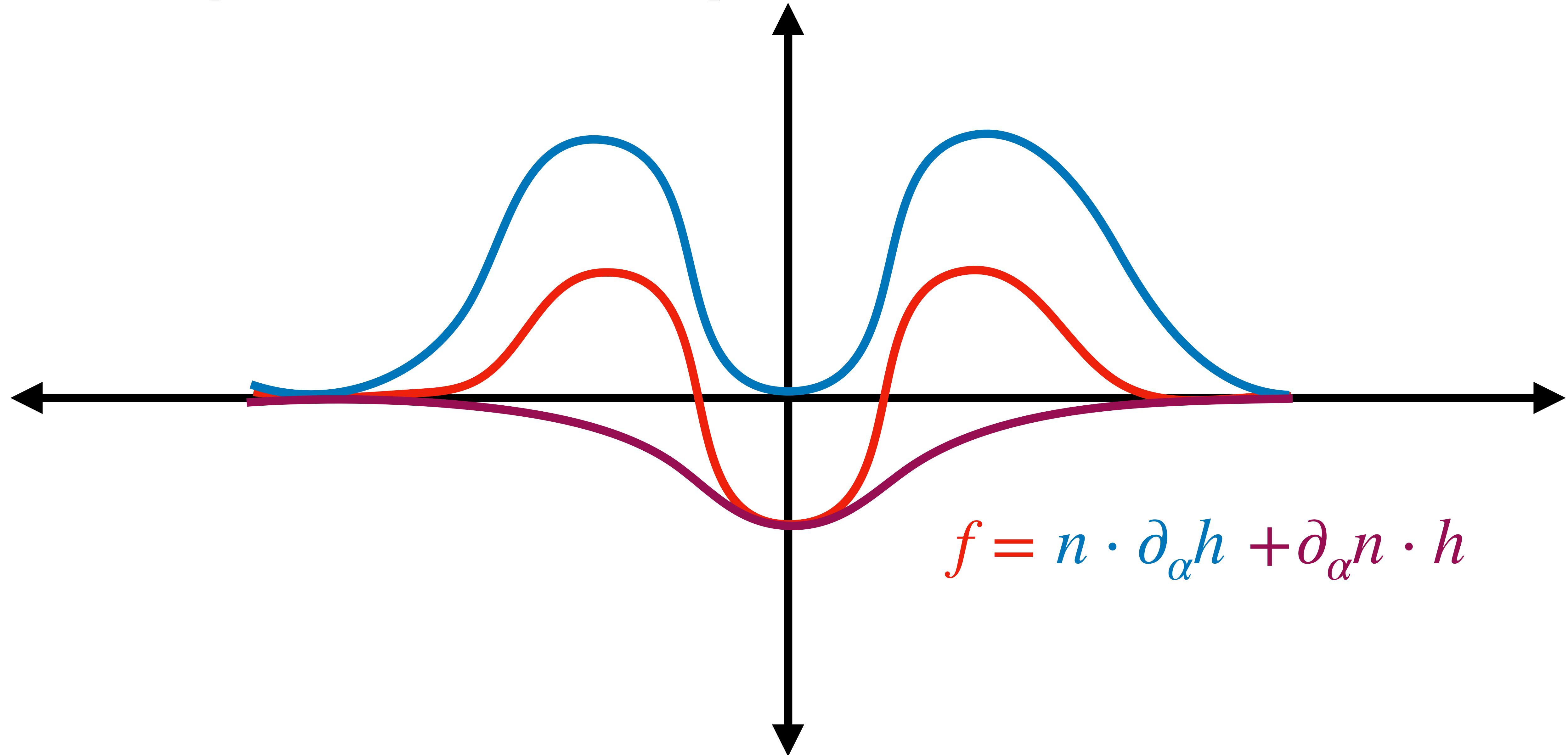
Our product decomposition



Our product decomposition

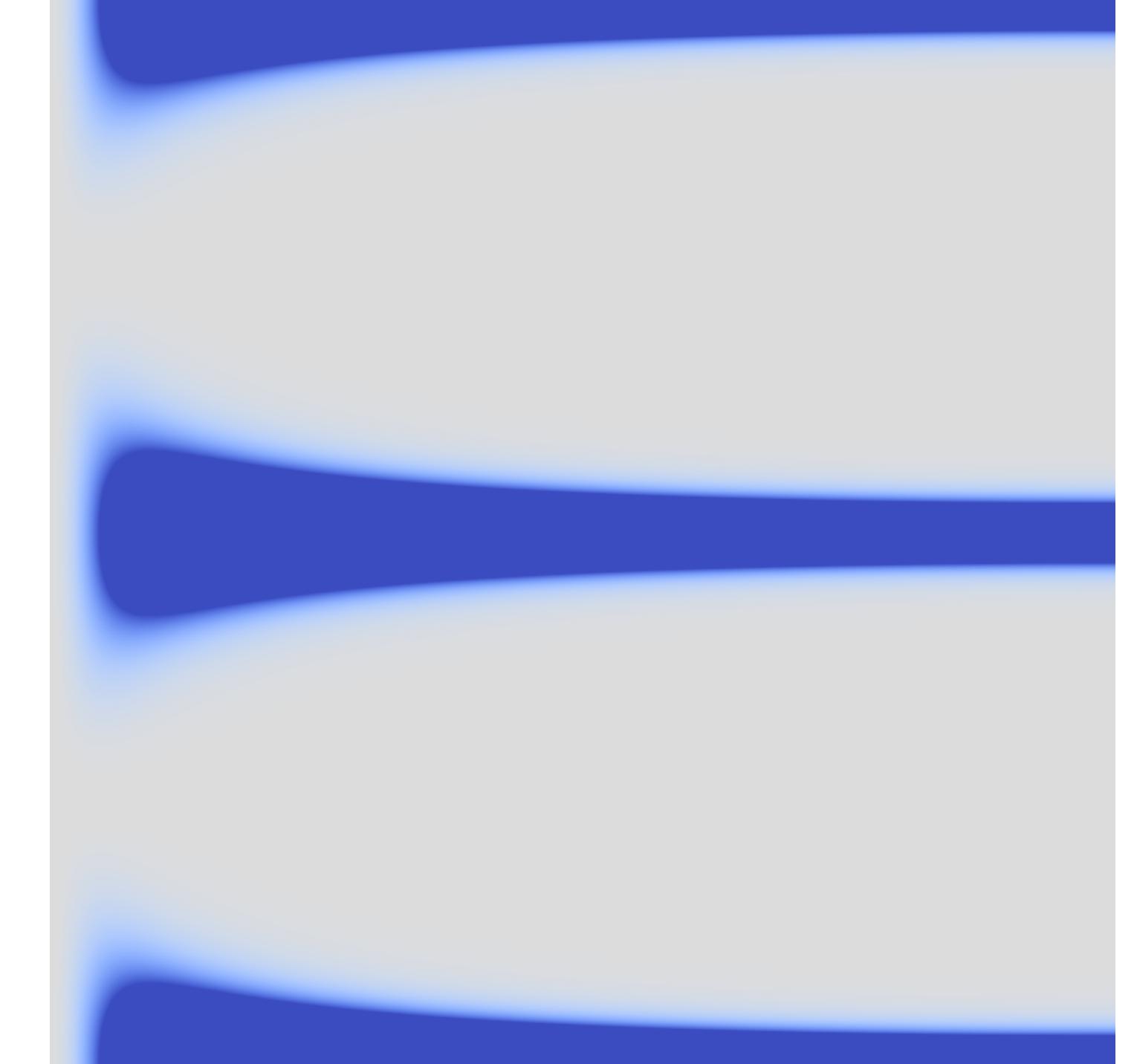
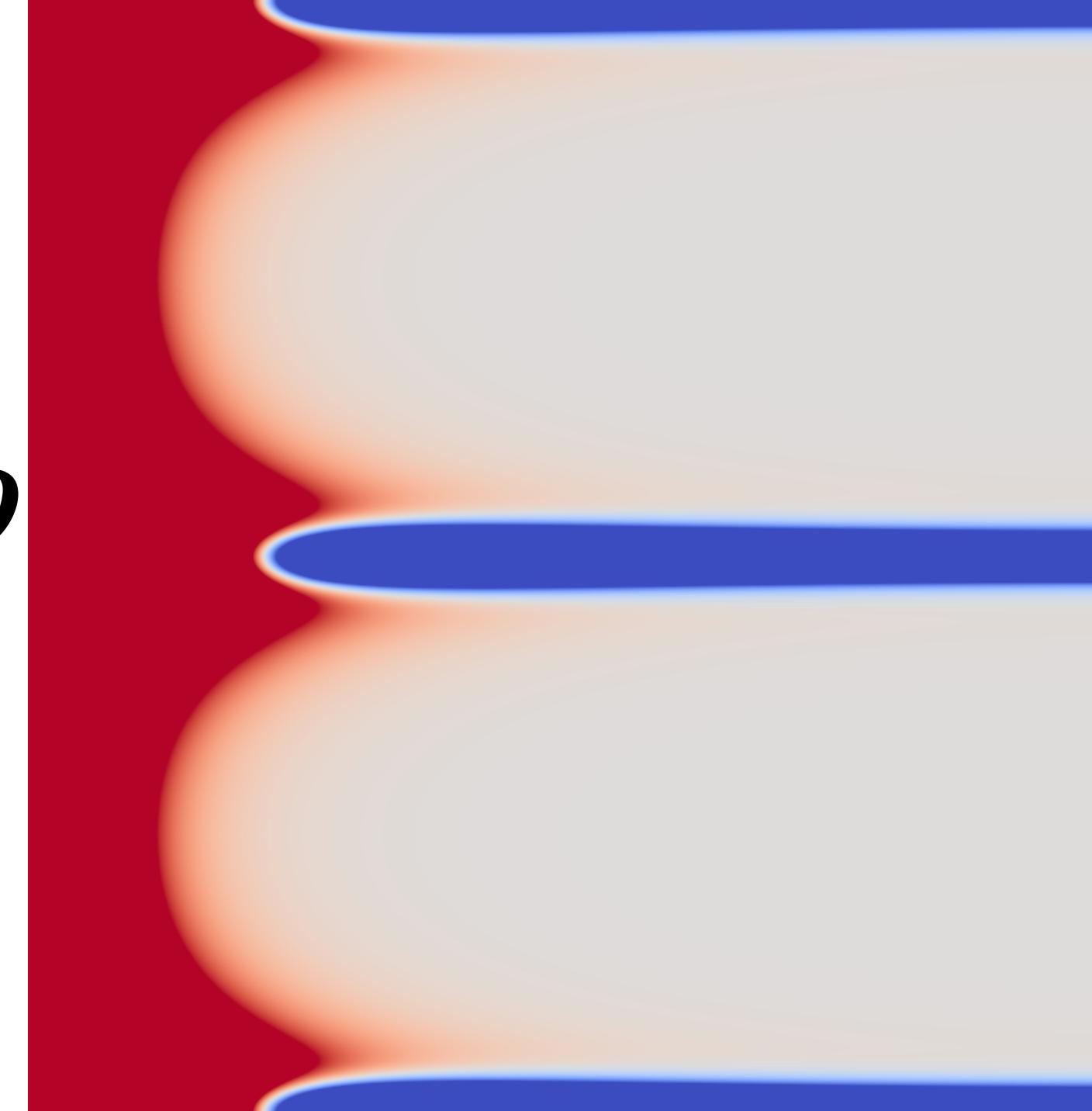


Our product decomposition



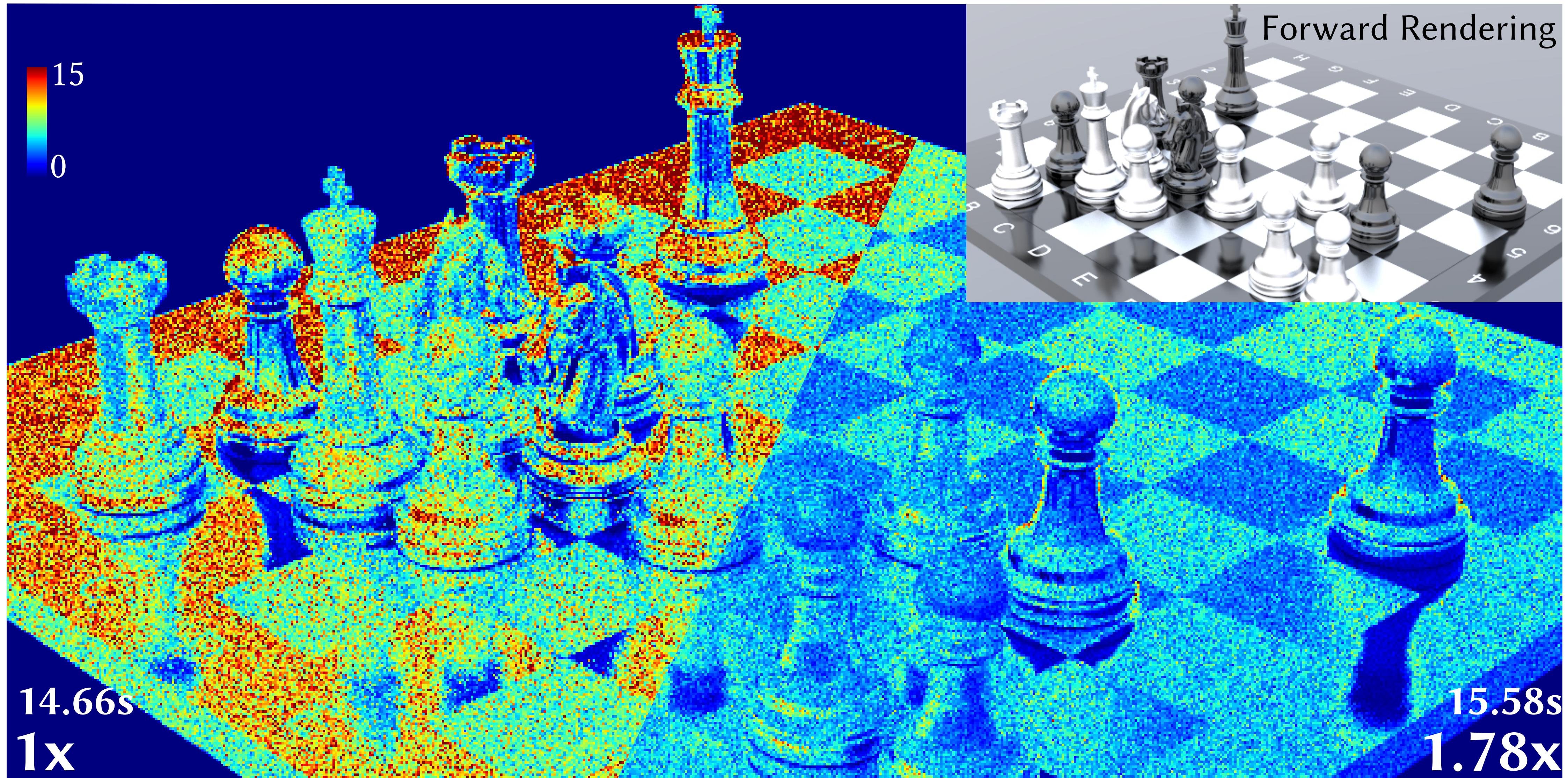
Product decomposition of anisotropic GGX derivative

θ



$$f = n \cdot \partial_\alpha h + \partial_\alpha n \cdot h$$

Product decomposition under global illumination

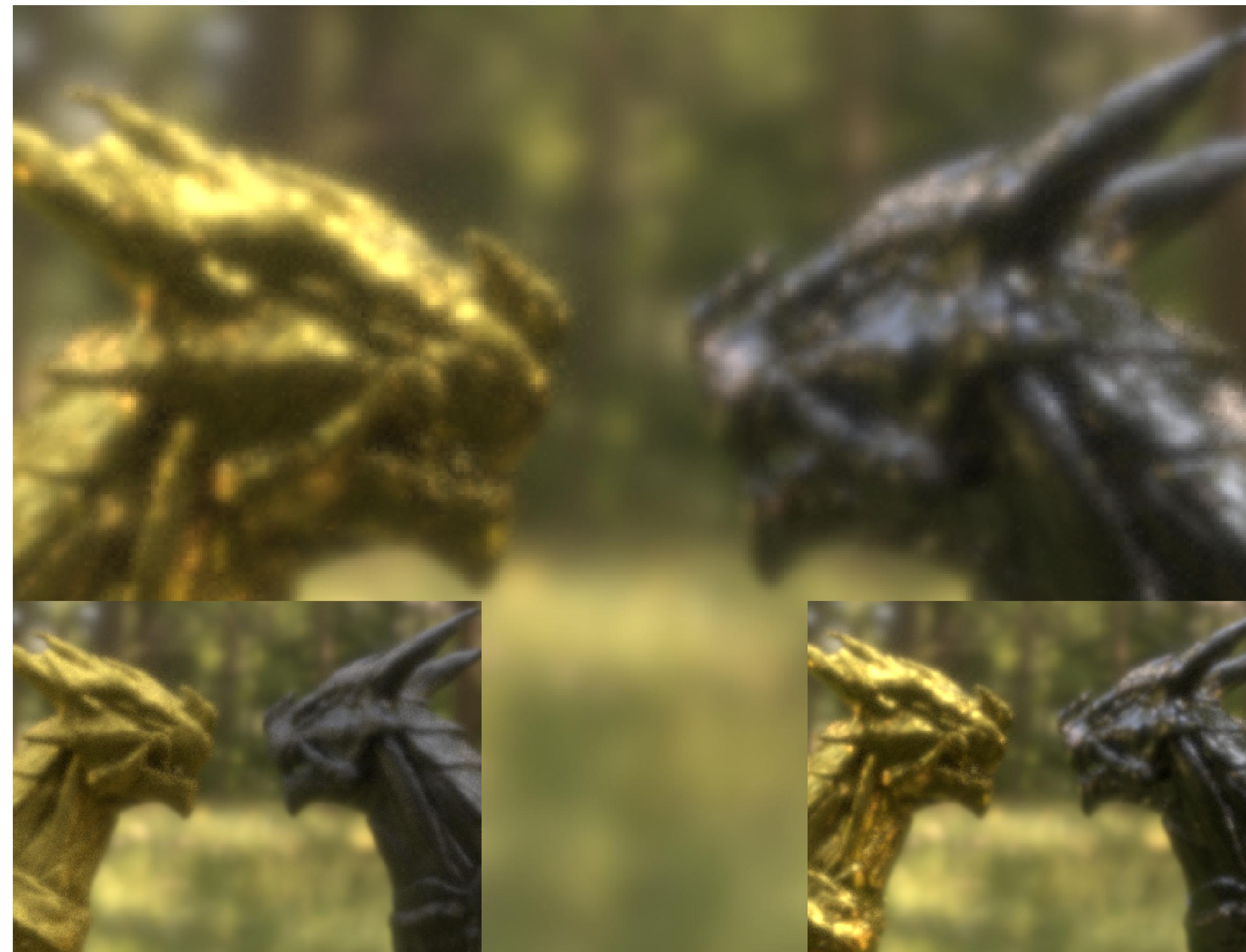


(a) BRDF Sampling

(b) Our Product Decomposition with MIS

Product decomposition improves inverse rendering

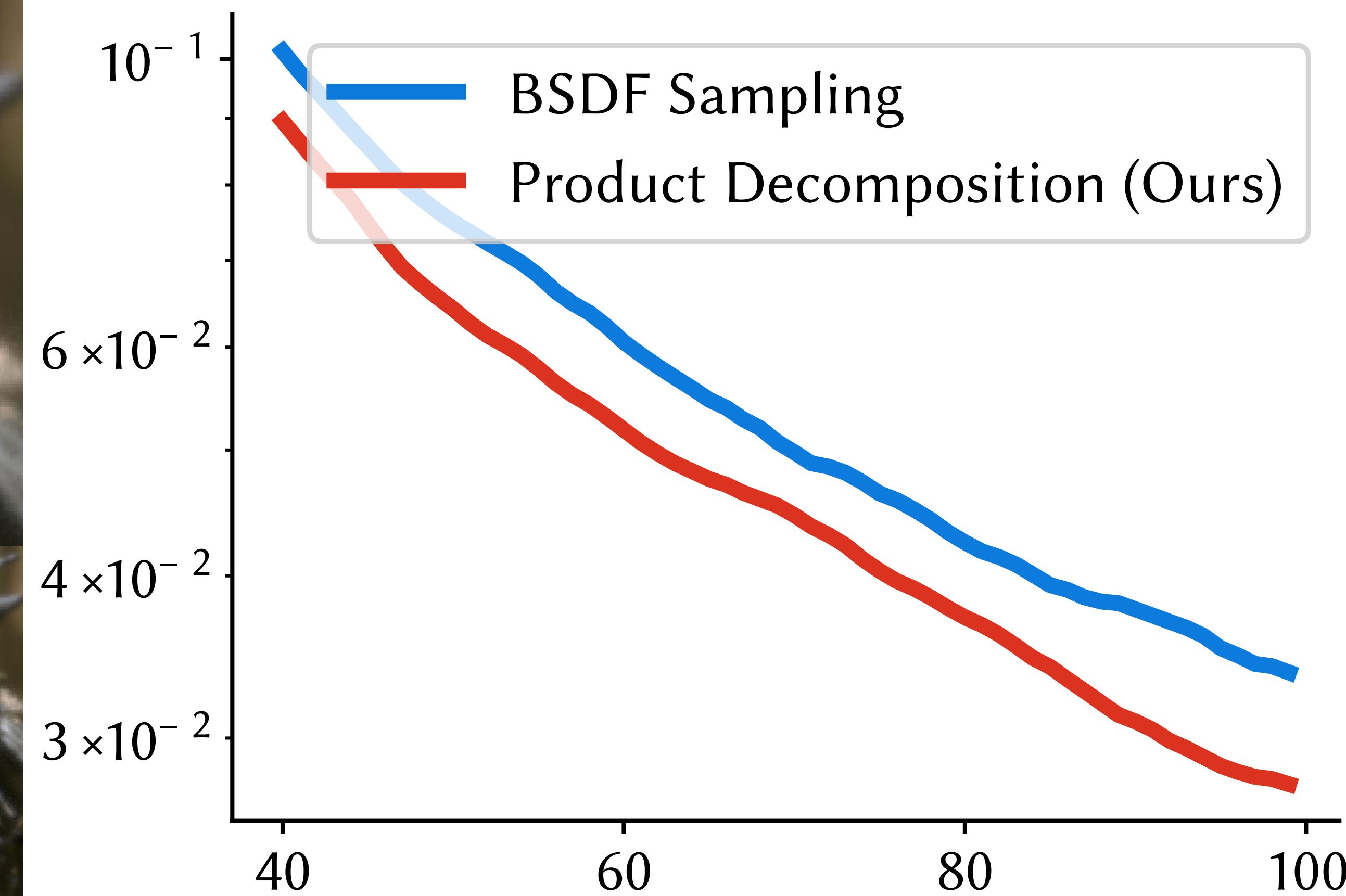
(a) Forward Rendering of Target



Initialization

Our Recovery Rendering

(b) α_x Recovery Loss (L1) over Iterations



- Directional roughness of anisotropic **Beckmann and GGX**
- Directional exponent of **Ashikhmin-Shirley**
- Width of **Burley's BSSRDF**

Our product decomposition

We now have good importance sampling techniques for BRDF in differentiable rendering