We first define the function h and put the number of digits high, to have better computer precision

>
$$h(x) := K \frac{(x \cdot \log(x) + (1 K x) \cdot \log(1 K x))}{\log(2)};$$

 $h(0) := 0;$
 $h(1) := 0;$

$$h := x \mapsto \mathsf{K} \frac{x \cdot \log(x) + (1 \mathsf{K} x) \cdot \log(1 \mathsf{K} x)}{\log(2)}$$

$$h(0) := 0$$

$$h(1) := 0$$
(1)

The number of digits can be adjusted. (it has been done with larger accuracy as well, but some outcomes become unconveniently lengthy)

>
$$Digits := 20$$
;
$$Digits := 20$$
(2)

We compute the values for b,a and c which are related with the atomic distribution which shows that the improvement is bounded (by c).

> with(RootFinding):

$$B := NextZero(x \mapsto h(x) * (2 \mathsf{K} \ h(x)) \mathsf{K} \ h(2 * x \mathsf{K} \ x^2), 0.14);$$

$$B := 0.32945473850303697239$$
 (3)

> evalf[10](B);> $A := \frac{(1 \text{ K } h(B))}{(2 \text{ K } h(B))};$

$$A := \frac{1 \,\mathsf{K} \, \frac{0.63379181713619122525}{\ln(2)}}{2 \,\mathsf{K} \, \frac{0.63379181713619122525}{\ln(2)}}$$
 (5)

 \rightarrow evalf [10](A);

 $\succ C := A + (1 \mathsf{K} \ A) \cdot B;$

$$C := \frac{0.67054526149696302761 \left(1 \, \text{K} \, \frac{0.63379181713619122525}{\ln(2)} \right)}{2 \, \text{K} \, \frac{0.63379181713619122525}{\ln(2)}}$$
(7)

+ 0.32945473850303697239

 \rightarrow evalf [10](C);

Next, we consider the ratio's of the derivatives to compute the optimal choice of alpha.

>
$$gI(x) := diff((1K(CKx)/(1Kx))^2 * h(2*xKx^2)K(1K(CKx)/(1Kx)) * h(x), x);$$

$$gI := x \rightarrow \frac{\partial}{\partial x} \left(\left(1 \,\mathsf{K} \, \frac{C \,\mathsf{K} \, x}{1 \,\mathsf{K} \, x} \right)^2 h(2 \, x \,\mathsf{K} \, x^2) \,\mathsf{K} \, \left(1 \,\mathsf{K} \, \frac{C \,\mathsf{K} \, x}{1 \,\mathsf{K} \, x} \right) h(x) \right) \tag{9}$$

$$g2(x) := diff((1K 2*(CK x)/(1K x))K (1K (CK x)/(1K x))*h(x), x);$$

$$g2 := x \rightarrow \frac{\partial}{\partial x} \left(1K \frac{2CK 2x}{1K x}K \left(1K \frac{CK x}{1K x}\right)h(x)\right)$$

$$(10)$$

> alpha :=
$$K \frac{eval(g1(x), x=B)}{(eval(g2(x), x=B)K eval(g1(x), x=B))};$$

$$\alpha := K \left[\frac{1}{\ln(2)} \left(1.3761289611414332582 \right) \right]$$
 (11)

$$\mathsf{K} \; \frac{1.4913236397607875851 \left(1 \, \mathsf{K} \; \frac{0.63379181713619122525}{\ln(2)} \right)}{2 \, \mathsf{K} \; \frac{0.63379181713619122525}{\ln(2)}} \right) \bigg)$$

$$\mathsf{K} \ \frac{1}{\ln(2)} \left[0.27111744475595941547 \left[1 \right] \right]$$

$$\mathsf{K} \ \frac{1}{\ln(2)} \left(0.63379181713619122525 \left(1.4913236397607875851 \right) \right)$$

$$\mathsf{K} = \frac{1.4913236397607875851 \left(1 \, \mathsf{K} \cdot \frac{0.63379181713619122525}{\ln(2)} \right)}{2 \, \mathsf{K} \cdot \frac{0.63379181713619122525}{\ln(2)}} \right) \bigg)$$

$$\mathsf{K} \ \frac{1}{\ln(2)} \left(0.71065222421202958606 \left(1 \right) \right)$$

$$\frac{2.9826472795215751702\,\mathsf{K}}{2.9826472795215751702\,\left(1\,\mathsf{K}\,\,\frac{0.63379181713619122525}{\ln(2)}\right)}{2\,\mathsf{K}\,\,\frac{0.63379181713619122525}{\ln(2)}}$$

$$K \frac{1}{\ln(2)} \left[1.3761289611414332582 \right[1$$

$$\mathsf{K} = \frac{1.4913236397607875851 \left(1 \, \mathsf{K} \, \frac{0.63379181713619122525}{\ln(2)} \right)}{2 \, \mathsf{K} \, \frac{0.63379181713619122525}{\ln(2)}}$$

$$+ \frac{1}{\ln(2)} \left(0.27111744475595941547 \left(1 \right) \right)$$

> *evalf* [10](alpha)

We verify the statement for probability distributions for which the support contains 3 elements; a_1,a_2 and 1, which have probabilities respectively equal to p_1, p_2 and $1-p_1-p_2$. As such we can compute E[H(p)], E[H(p+q-pq)] and E'[H(min(2*p,1/2)]

>
$$H(p_1, a_1, p_2, a_2) := p_1 \cdot h(a_1) + p_2 \cdot h(a_2);$$

 $H := (p_1, a_1, p_2, a_2) \mapsto p_1 \cdot h(a_1) + p_2 \cdot h(a_2)$ (13)

► $Hpq(p_1, a_1, p_2, a_2) := p_1^2 \cdot h((1 \times a_1)^2) + p_2^2 \cdot h((1 \times a_2)^2) + 2 \cdot p_1 \cdot p_2 \cdot h((1 \times a_1)^2) + 2 \cdot p_1 \cdot p_2 \cdot h((1 \times a_2)^2) + 2 \cdot p_1 \cdot p_2 \cdot h((1 \times a_2)^2) + 2 \cdot p_1 \cdot p_2 \cdot h((1 \times a_2)^2) + 2 \cdot p_1 \cdot p_2 \cdot h((1 \times a_2)^2) + 2 \cdot p_1 \cdot p_2 \cdot h((1 \times a_2)^2) + 2 \cdot p_1 \cdot p_2 \cdot h((1 \times a_2)^2) + 2 \cdot p_1 \cdot p_2 \cdot h((1 \times a_2)^2) + 2 \cdot p_2 \cdot h((1 \times a_2)^2) + 2 \cdot p_1 \cdot p_2 \cdot h((1 \times a_2)^2) + 2 \cdot h((1 \times a_2)^2) + 2 \cdot h((1 \times a_2)^2) + 2 \cdot h$

$$Hpq := (p_1, a_1, p_2, a_2) \mapsto p_1^2 \cdot h((1 \times a_1)^2) + p_2^2 \cdot h((1 \times a_2)^2) + 2 \cdot p_1 \cdot p_2$$

$$\cdot h((1 \times a_1) \cdot (1 \times a_2))$$
(14)

>
$$Hpr(p_1, a_1, p_2, a_2) := (p_1 + \min((p_2 \mathsf{K} \ (1 \mathsf{K} \ p_1 \mathsf{K} \ p_2)), 0)) \cdot h\left(\min\left(2 \cdot a_1, \frac{1}{2}\right)\right) + \max((p_2 \mathsf{K} \ (1 \mathsf{K} \ p_1 \mathsf{K} \ p_2)), 0) \cdot h\left(\min\left(2 \cdot a_2, \frac{1}{2}\right)\right);$$

$$Hpr := (p_1, a_1, p_2, a_2) \mapsto (p_1 + \min(2 \cdot p_2 \mathsf{K} \ 1 + p_1, 0)) \cdot h\left(\min\left(2 \cdot a_1, \frac{1}{2}\right)\right) + \max(2 \cdot p_2 \mathsf{K} \ 1 + p_1, 0) \cdot h\left(\min\left(2 \cdot a_2, \frac{1}{2}\right)\right)$$

$$(15)$$

We now consider the optimization problem for the two considered cases; hereby we solve the minimization problem with two different substitutions (once with $E[H(p)] \le C$ and once with E[H(p)] = C)

and also plot the 2D graph to see that the inequality is indeed true for the mentioned value C.

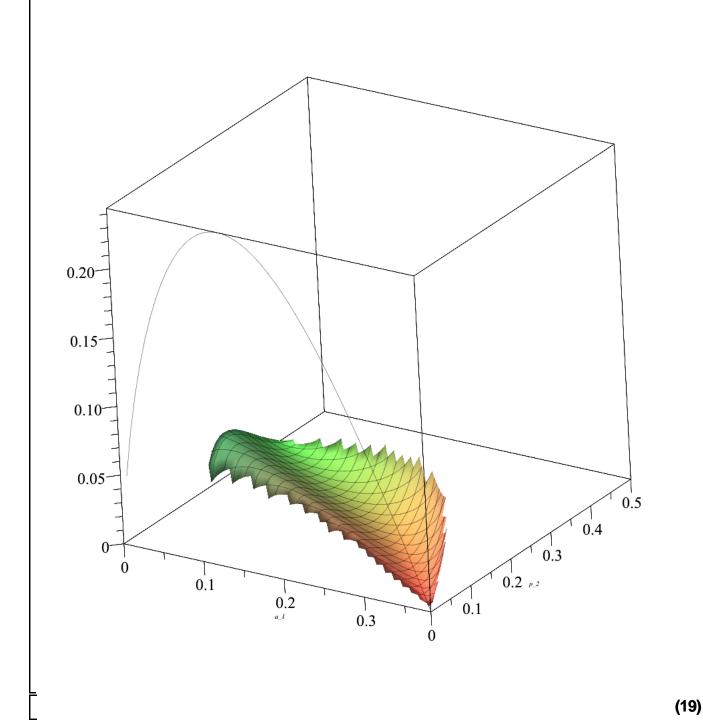
Case 1: (a_1,a_2,1) getting weights (1-2p_2,p_2.p_2), In this case the difference is zero and attained by the earlier mentioned atomic distribution since a 1=a 2)

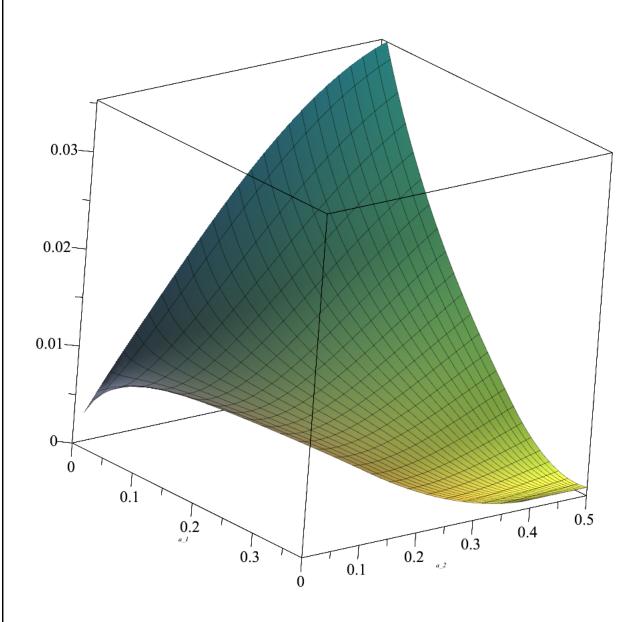
> with(Optimization):
Minimize((1 K alpha)·Hpq(1K 2 p_2, a_1, p_2, a_2) + alpha· Hpr(1K 2 p_2, a_1, p_2, a_2)
K H(1K 2 p_2, a_1, p_2, a_2), { p_2 ≤ 1, a_1 ≤ C, a_2 ≤ 0.5, (1 K 2 p_2)·a_1 + p_2
·a_2 + p_2 ≤ C, a_1 ≤ a_2}, assume = nonnegative)
[K 4.801 ×
$$10^{\text{K}}$$
 [a_1 = 0.32945473850294784466, a_2 = 0.32945473850294784466, p_2
= 0.078877292706045659566]]

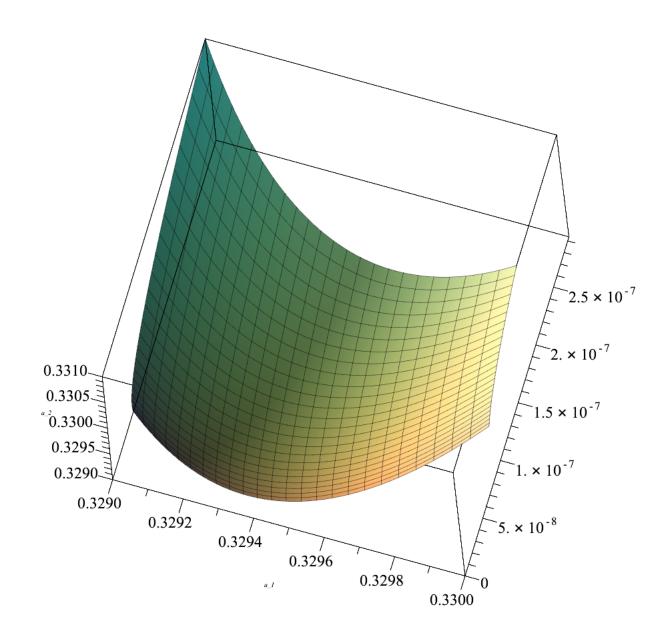
>
$$F(p_1, a_1, p_2, a_2) := (1 \text{ K alpha}) \cdot Hpq(p_1, a_1, p_2, a_2) + \text{alpha} \cdot Hpr(p_1, a_1, p_2, a_2) \times H(p_1, a_1, p_2, a_2)$$

 $F := (p_1, a_1, p_2, a_2) \mapsto (1 \text{ K } \alpha) \cdot Hpq(p_1, a_1, p_2, a_2) + \alpha \cdot Hpr(p_1, a_1, p_2, a_2)$ (17)
 $K \cdot H(p_1, a_1, p_2, a_2)$

> with(Optimization): $Minimize \Big(F \Big(1 \,\mathsf{K} \, \, 2 \cdot \frac{(C \,\mathsf{K} \, \, a_1)}{(1 + a_2 \,\mathsf{K} \, \, 2 \cdot a_1)}, \, a_1, \, \frac{(C \,\mathsf{K} \, \, a_1)}{(1 + a_2 \,\mathsf{K} \, \, 2 \cdot a_1)}, \, a_2 \Big), \, \{0 \leq a_1, \, a_1 \leq C, \, a_1 \leq a_2, \, a_2 \leq 0.5 \} \Big);$ [K 1.8 × 10^{K20}, [$a_1 = 0.32945473850303617469, \, a_2 = 0.32945473850303807804$]] (18)







Case 2: (a_1,a_2) getting weights (1-p_2,p_2),In this case the difference is even strict. (the plot shows that the initial point is indeed a global maximum)

```
with (Optimization):

Minimize((1 K alpha)·Hpq(1 K p_2, a_1, p_2, a_2)) + alpha·Hpr(1 K p_2, a_1, p_2, a_2)

K H(1 K p_2, a_1, p_2, a_2), { p_2 \le 1, a_1 \le C, a_2 \le 0.5, (1 K p_2)·a_1 + p_2·a_2
\le C, a_1 \le a_2}, assume = nonnegative);

Warning, no iterations performed as initial point satisfies first-order conditions

[0.00086730175254384197, [a_1 = 0.38234553336670272114, a_2 = 0.38234553336670272114, a_3 = 0.38234553336670272114
```

 $= 0.38234553336670272114, p_2 = 1.0]]$

 $\begin{array}{ll} \color{red} \Longrightarrow & \textit{with}(\textit{Optimization}): \\ & \textit{Minimize}\Big(F\Big(1\,\mathsf{K}\,\,\frac{(C\,\mathsf{K}\,\,a_1)}{(\,a_2\,\mathsf{K}\,\,a_1)},\,a_1,\,\frac{(C\,\mathsf{K}\,\,a_1)}{(\,a_2\,\mathsf{K}\,\,a_1)},\,a_2\Big),\,\{0\leq a_1,\,a_1\leq C,\,C\leq a_2,\,a_2\leq 0.5\}\Big); \end{array}$

Warning, no iterations performed as initial point satisfies firstorder conditions

