

## A SHARP ESTIMATE ON THE NORM OF THE MARTINGALE TRANSFORM

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### 1. Introduction

The boundedness of singular integral operators in  $L^2(w)$  for  $w \in A_2$  has been known for a long time, by the Hunt-Muckenhaupt-Wheeden Theorem. What is as of yet unknown is the sharp bound of these operators in terms of the  $A_2$  norm of  $w$ . In his thesis, S. Buckley proved that the Hardy-Littlewood maximal operator is bounded linearly in  $\|w\|_{A_2}$ , the square function operator bound is no more than  $\|w\|_{A_2}^{3/2}$  and that the Hilbert transform norm is no worse than quadratic. It is easily seen that each of these operators can not have a better bound than linear in  $\|w\|_{A_2}$  (just look at power weights). In this paper, following the methods of [6], we show that the bound for the martingale transform, which is a dyadic analog of singular integral operators, is linear. We also give a simple proof that the dyadic square function is bounded linearly.

### 2. Notation

In what follows,  $h_I$  will denote the normalized Haar function for the dyadic interval  $I$ , i.e.,  $h_I = \frac{\chi_{I_l} - \chi_{I_r}}{\sqrt{|I|}}$ , where  $I_l, I_r$  denote the left and right children of  $I$  respectively. The weight  $w$  and its inverse  $w^{-1}$  will be dyadic  $A_2$  weights on  $[0, 1]$ .  $w$  will be normalized to have  $\int_{[0,1]} w(x) = 1$ . Let  $(f)_I$  denote  $\frac{1}{|I|} \int_I f(x) dx$  for  $I$  a dyadic interval. Sometimes the parentheses are omitted when it is clear which function we are averaging.  $w_I^{-1}$  will denote  $\frac{1}{|I|} \int_I \frac{1}{w(x)} dx$ . Let  $\mu(I) = w_I w_I^{-1}$  and  $\|w\|_{A_2} = \sup_{I \in \mathcal{D}} \mu(I)$ .

We will write  $w$  in the form  $w(x) = \prod (1 + c_I h_I)$  where  $c_I = \frac{w_{I_l} - w_{I_r}}{2w_I} \sqrt{|I|}$ . Similarly, we define  $d_I = \frac{w_{I_l}^{-1} - w_{I_r}^{-1}}{2w_I^{-1}} \sqrt{|I|}$ .

$\langle \cdot, \cdot \rangle_\mu$  will denote the inner product in  $L^2(d\mu)$ . If the subscript is omitted, the measure is  $dx$ .

The family of operators which we are concerned with are the martingale transforms

$$T_r f = \sum_{I \in \mathcal{D}[0,1]} r(I) \langle f, h_I \rangle h_I$$

where  $r(I)$  assumes the values  $+1$  and  $-1$  only.

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### 3. The main theorem

**Theorem 3.1.**

$$\|T_r f\|_{L^2(w)} \leq c \|w\|_{A_2} \|f\|_{L^2(w)},$$

for  $w \in A_2$  and  $f \in L^2(w)$ .

We will prove this Theorem in section 5.

### 4. Useful lemmas and theorems

The following theorem can be found in [6], Section 2.

**Theorem 4.1.** *Let  $\alpha_I \geq 0$ . Then*

$$\sum_{I \in D[0,1]} (fw^{1/2})_I^2 \alpha_I \leq 4c \|f\|_2^2$$

for all  $f \in L^2(dx)$  iff

$$\frac{1}{|J|} \sum_{I \subset J} (w_I)^2 \alpha_I \leq cw_J$$

for all  $J \in D[0,1]$ .

**Theorem 4.2.** *Let  $\alpha_I \geq 0$ . If*

$$\frac{1}{|J|} \int_J \left( \sum_{I \subset J} \alpha_I w_I \chi_I(x) \right)^2 w^{-1} dx \leq c_1 w_J$$

for all  $J \in D$  and

$$\frac{1}{|J|} \int_J \left( \sum_{I \subset J} \alpha_I w_I^{-1} \chi_I(x) \right)^2 w dx \leq c_1 w_J^{-1}$$

for all  $J \in D$ , then

$$\sum_{I \in D} \alpha_I (fw^{1/2})_I (gw^{-1/2})_I |I| \leq c_2 \|f\|_{L^2} \|g\|_{L^2},$$

for all  $f, g \in L^2(dx)$ , where  $c_2 \leq c\sqrt{c_1}$  and  $c$  is an absolute constant.

*Proof.* From [6], section 4, we know that if

$$\frac{1}{|J|} \int_J \left( \sum_{I \subset J} \alpha_I w_I \chi_I(x) \right)^2 v(x) dx \leq w_J$$

for all  $J \in D$ , then

$$\sum_{I \in D_1} \alpha_I (fw^{1/2})_I (gv^{1/2})_I |I| \leq c(\|f\|_{L^2}^2 + \|g\|_{L^2}^2),$$

where  $D_1 = \{I \in D : \frac{(fw^{1/2})_I^2}{w_I} \geq \frac{(gv^{1/2})_I^2}{v_I}\}$ .

To prove Theorem 4.2, we use this theorem with  $v = \frac{w^{-1}}{c_1}$ . Then, since in the conclusion  $v$  appears with a square root only, the bound for the conclusion

would be of the order  $\sqrt{c_1}$ . Furthermore, imposing that the condition be fulfilled for both  $w$  and  $w^{-1}$ , we can sum over all dyadic intervals, since for all of them either

$$\frac{(fw^{1/2})_I^2}{w_I} \geq \frac{(g(w^{-1})^{1/2})_I^2}{w_I^{-1}},$$

or

$$\frac{(fw^{1/2})_I^2}{w_I} \leq \frac{(g(w^{-1})^{1/2})_I^2}{w_I^{-1}}$$

is true. Finally, letting  $g^* = (\frac{\|g\|_{L^2}}{\|f\|_{L^2}})^{1/2}g$  and  $f^* = (\frac{\|g\|_{L^2}}{\|f\|_{L^2}})^{-1/2}f$ , we get

$$\sum_{I \in D_1} \alpha_I (f^* w^{1/2})_I (g^* (w^{-1})^{1/2})_I |I| \leq 2c (\|f^*\|_{L^2} \|g^*\|_{L^2}),$$

since this normalization does not change the right hand side at all.  $\square$

The following estimate is not used in the proof of the main theorem, but may be interesting in its own right. It gives the (unweighted) carleson norm of the sequence  $\{c_I\}$ . See [7], theorem 4.1 .

**Lemma 4.3.** *For  $w \in A_2$  where  $w(x) = C \prod (1 + c_I h_I)$ , the following is true:*

$$\sum_{I \subset J} c_I^2 \leq 2 \log(\|w\|_{A_2}) |J|,$$

and this estimate is sharp.

*Proof.* By the definition of  $w$ ,

$$\begin{aligned} (1 - c_I^2/|I|) &= 1 - \left(\frac{w_{I_l} - w_{I_r}}{2w_I}\right)^2 = \\ &= \frac{4(1/2(w_{I_l} + w_{I_r}))^2 - (w_{I_l} - w_{I_r})^2}{(2w_I)^2} = \frac{w_{I_l} w_{I_r}}{(w_I)^2}, \end{aligned}$$

and therefore

$$\frac{w_{I_l} w_{I_r}}{(w_I)^2} \leq e^{-c_I^2/|I|}.$$

Taking square roots, this becomes  $w_I \geq e^{c_I^2/2|I|} (w_{I_l} w_{I_r})^{1/2}$ .

Since  $(w_I^{-1})^2 \geq (w_{I_l}^{-1} w_{I_r}^{-1})$ , we can multiply these two equations to get

$$\mu(I) \geq e^{c_I^2/2|I|} (\mu(I_l) \mu(I_r))^{1/2}.$$

Now use the same inequality to replace  $\mu(I_l)$  and  $\mu(I_r)$ , and repeat this process. After  $n$  steps we get

$$\mu(I) \geq \exp \left( \sum_{K \subset I, |K| \geq 2^{-n}|I|} \frac{c_K^2 |K|}{2|K||I|} \right) \left( \prod_{|K|=2^{-n}|I|} \mu(K) \right)^{2^{-n}}.$$

(The  $|K|/|I|$  in the exp comes from the repeated square roots that we took.) Realizing that  $\mu(K) \geq 1$  always, we get

$$\mu(I) \geq \exp \left( \sum_{K \subset I, |K| \geq 2^{-n}|I|} \frac{c_K^2}{2|I|} \right).$$

Taking the limit as  $n \rightarrow \infty$ , and taking logarithms, we get the lemma.

To see that the bound is indeed sharp, consider  $w = \exp(\sum_{i \in \mathbb{N}} b_i \Phi_i)$  where  $\Phi_i$  are Rademacher functions. Then  $\sum_{I \subset J} c_I^2 = \sum_{i \geq j} \tanh^2(b_i) |J|$ , while  $\|w\|_{A_2} = \prod_{i \in \mathbb{N}} (1 + \sinh^2(b_i))$ .  $\square$

**Theorem 4.4.** *Let  $Sf$  be the dyadic square function of  $f$ . Then*

$$\|Sf\|_{L^2(w)} \leq c \|w\|_{A_2} \|f\|_{L^2(w)},$$

and this estimate is sharp.

*Note.* This estimate has also been proven, using a different method, in [5].

*Proof.* Let

$$f_I = \frac{h_I}{(w_I)^{1/2}},$$

and

$$g_I = h_I (w_I)^{1/2} w^{-1}.$$

Then  $\langle f_I, g_J \rangle_w = 1$  if  $I = J$  and 0 otherwise. Let  $f \in L^2(w)$ .

$$\|Sf\|_{L^2(w)} = \left( \sum_{I \in D} \langle f, h_I \rangle_w^2 \right)^{1/2} = \left( \sum_{I \in D} \langle f, g_I \rangle_w^2 \right)^{1/2}.$$

In that notation, the theorem to prove becomes

$$\sum_{I \in D} \langle f, g_I \rangle_w^2 \leq \|w\|_{A_2}^2 \|f\|_{L^2(w)}^2.$$

We will bound this by first showing that

$$\sum_{I \in D} \langle f, f_I \rangle_w^2 \leq \|w\|_{A_2}^2 \|f\|_{L^2(w)}^2,$$

from which the desired inequality can be extracted by algebraic manipulations.

By the theorem on the lower bound of the square function in [3] we know the following:

$$\|f\|_{L^2(w)} \leq c \|w\|_{A_2}^{1/2} \left( \sum_{I \in D} \langle f, g_I \rangle_w^2 \right)^{1/2}.$$

This allows us to compute the norm of the sequence valued operator  $(Jg)_I = \langle g, f_I \rangle_w$  from  $L^2(w)$  to  $l^2$ :

$$\begin{aligned} & \sup_{\|g\|_{L^2(w)}=1} \sup_{\|\{k_I\}\|_{l^2}=1} \sum_{I \in D} k_I \int g f_I w = \\ & \sup_{\|g\|_{L^2(w)}=1} \sup_{\|\{k_I\}\|_{l^2}=1} \int g \sum_{I \in D} k_I f_I w \leq \end{aligned}$$

by Cauchy-Schwarz, and the above inequality,

$$\begin{aligned} & \sup_{\|\{k_I\}\|_{l^2}=1} \sup_{\|g\|_{L^2(w)}=1} \left\| \sum_{I \in D} k_I f_I(x) \right\|_{L^2(w)} \|g\|_{L^2(w)} \leq \\ & \|w\|_{A_2}^{1/2} \sup_{\|\{k_I\}\|_{l^2}=1} \left[ \sum_{J \in D} \left\langle \left( \sum_{I \in D} k_I f_I \right), g_J \right\rangle_w^2 \right]^{1/2} = \\ & \|w\|_{A_2}^{1/2} \sup_{\|\{k_I\}\|_{l^2}=1} \left( \sum_{I \in D} k_I^2 \right)^{1/2} = \|w\|_{A_2}^{1/2}. \end{aligned}$$

This means that

$$\sum_{I \in D} \langle g, f_I \rangle_w^2 \leq \|w\|_{A_2} \|g\|_{L^2(w)}^2,$$

for every weight in  $A_2$ . In particular, this is true for the weight  $w^{-1}$ :

$$\sum_{I \in D} \left\langle g, \frac{h_I}{(w_I^{-1})^{1/2}} \right\rangle_{w^{-1}}^2 \leq \|w\|_{A_2} \|g\|_{L^2(w^{-1})}^2.$$

Let  $g^* = gw^{-1}$ . Then this becomes

$$\sum_{I \in D} \left\langle g^*, \frac{h_I}{(w_I^{-1})^{1/2}} \right\rangle_{dx}^2 \leq \|w\|_{A_2} \|g^*\|_{L^2(w)}^2.$$

The left hand side can be rewritten as

$$\sum_{I \in D} \left\langle g^*, h_I (w_I)^{1/2} w w^{-1} \right\rangle_{dx}^2 \frac{1}{\mu(I)}.$$

So finally, we have the inequality

$$\sum_{I \in D} \langle g^*, g_I \rangle_w^2 \leq \|w\|_{A_2}^2 \|g^*\|_{L^2(w)}^2,$$

which is what we needed to prove.  $\square$

*Remark 4.5.* We have the following inequality:

$$\frac{1}{|J|} \sum_{I \subset J} (w_I^{-1})^2 d_I^2 w_I \leq c \|w\|_{A_2}^2 w_J^{-1}.$$

This inequality follows by realizing that the left hand side is bounded by  $\|S(w^{-1}\chi_J)\|_{L^2(w)}^2$ , which by Theorem 4.4 is bounded by  $c \|w\|_{A_2}^2 \|w^{-1}\chi_J\|_{L^2(w)}^2$ .

**Lemma 4.6.**

$$\sum_{I \subset J} \left( \frac{c_I}{\sqrt{|I|}} \right)^2 w_I |I| \leq c \|w\|_{A_2} w_J |J|,$$

and this estimate is sharp.

*Proof.* By [1], 2.2 we have the following two estimates:

$$\sum_{I \subset J} (w_I)^p \left( \frac{w_{I_l} - w_{I_r}}{w_I} \right)^2 |I| \leq \frac{c}{p(p-1)} \|w\|_{B_p}^p (w_J)^p |J|,$$

for  $w \in B_p$ ,  $p > 1$ , and

$$\sum_{I \subset J} (w_I)^q \left( \frac{w_{I_l} - w_{I_r}}{w_I} \right)^2 |I| \leq \frac{c}{q(1-q)} (w_J)^q |J|,$$

for  $0 < q < 1$ . ( $c$  in both estimates is an absolute constant.)

Using Hölder's inequality on  $(w_I)^{(1+\epsilon)/2}$  and  $(w_I)^{(1-\epsilon)/2}$  we get

$$\sum_{I \subset J} \left( \frac{c_I}{\sqrt{|I|}} \right)^2 w_I |I| \leq \left( \sum_{I \subset J} \left( \frac{c_I}{\sqrt{|I|}} \right)^2 (w_I)^{1+\epsilon} |I| \right)^{1/2} \left( \sum_{I \subset J} \left( \frac{c_I}{\sqrt{|I|}} \right)^2 (w_I)^{1-\epsilon} |I| \right)^{1/2},$$

where we choose  $\epsilon$  to be  $c/\|w\|_{A_2}$  for  $c$  a small constant.

If  $c$  is chosen smaller than a dimensional constant, this  $\epsilon$  is one for which the reverse Hölder inequality holds for  $w$ , i.e.,  $(\frac{1}{|I|} \int_I w^{1+\epsilon})^{\frac{1}{1+\epsilon}} \leq 2 \frac{1}{|I|} \int_I w$  as can be seen by carefully reading the proof of the reverse Hölder inequality in [2]. Therefore  $w$  is in  $B_{1+\epsilon}$  with norm 2.

Letting  $p = 1+\epsilon$  and  $q = 1-\epsilon$  in the above estimates, and writing  $\frac{w_{I_l} - w_{I_r}}{2w_I} \sqrt{|I|}$  as  $c_I$  we get

$$\left( \sum_{I \subset J} \left( \frac{c_I}{\sqrt{|I|}} \right)^2 (w_I)^{1+\epsilon} |I| \right)^{1/2} \leq \left( \frac{c}{\epsilon(1+\epsilon)} \right)^{1/2} 2^{\frac{1+\epsilon}{2}} (w_J)^{\frac{1+\epsilon}{2}} |J|^{1/2}$$

and

$$\left( \sum_{I \subset J} \left( \frac{c_I}{\sqrt{|I|}} \right)^2 (w_I)^{1-\epsilon} |I| \right)^{1/2} \leq \left( \frac{c}{\epsilon(1-\epsilon)} \right)^{1/2} (w_J)^{\frac{1-\epsilon}{2}} |J|^{1/2},$$

which, when combined, yields the required estimate. (Remember that  $\epsilon = c/\|w\|_{A_2}$ .)

To see that this estimate is sharp, let  $w = \prod_{i \in \mathbb{N}} (1 + a2^{-i/2} h_{I_i})$  where  $h_{I_i}$  is the Haar function whose support is of the form  $[0, 2^{-i}]$ . Then  $\|w\|_{A_2}$  is comparable to  $\frac{1}{1-a}$ , and  $\sum_{I \subset J} \left( \frac{c_I}{\sqrt{|I|}} \right)^2 w_I |I| = \sum_{i \geq j} a^2 \left( \frac{1+a}{2} \right)^i$ .  $\square$

**Lemma 4.7.**  $\sum_{K \subset I} \frac{|c_K|}{\sqrt{|K|}} \frac{|d_K|}{\sqrt{|K|}} \mu(K) |K| \leq C \|w\|_{A_2} |I|$ .

*Proof.* By Lemma 5.3 from [6], we have

$$\frac{1}{|J|} \sum_{I \subset J} v_I w_I \left| \frac{v_{I_i} - v_{I_r}}{v_I} \frac{w_{I_i} - w_{I_r}}{w_I} \right| |I| \leq C \sqrt{v_J w_J},$$

provided  $v_I w_I \leq 1$  for all  $I \subset J$ .

Letting  $v_I = \frac{w_I^{-1}}{\|w\|_{A_2}}$ , and clearing the denominators, we get the above lemma.  $\square$

**Lemma 4.8.**  $\frac{1}{|J|} \sum_{I \subset J} \frac{|c_I|}{\sqrt{|I|}} \frac{|d_I|}{\sqrt{|I|}} w_I |I| \leq C \|w\|_{A_2} w_J.$

*Proof.* This lemma will be proven by the method of Bellman functions.

Let  $B(x, y) = x \left( \frac{-4A}{xy} - \frac{xy}{4A} + 4A + 1 \right)$ . Then  $B(x, y)$  has the following properties, which can be checked by elementary calculus:

1.  $0 \leq B(x, y) \leq 5Ax$  on  $\{x, y > 0; 1 \leq xy \leq A\}$
2.  $\begin{pmatrix} -B_{xx} & -B_{xy} - \frac{1}{y} \\ -B_{xy} - \frac{1}{y} & -B_{yy} \end{pmatrix}$  is positive semidefinite on  $\{x, y > 0; 1 \leq xy \leq 2A\}$ .
3.  $\begin{pmatrix} -B_{xx} & -B_{xy} + \frac{1}{y} \\ -B_{xy} + \frac{1}{y} & -B_{yy} \end{pmatrix}$  is positive semidefinite on  $\{x, y > 0; 1 \leq xy \leq 2A\}$ .

Let us show that these conditions imply the following discrete condition:

$$(*) \quad B(x, y) - \frac{B(x_-, y_-) + B(x_+, y_+)}{2} \geq C \left| (x_- - x_+)(y_- - y_+) \frac{1}{y} \right|,$$

where  $x = \frac{(x_- + x_+)}{2}$  and  $y = \frac{(y_- + y_+)}{2}$ , and  $(x, y), (x_-, y_-), (x_+, y_+)$  are in the domain  $\{x, y > 0; 1 \leq xy \leq A\}$ . Let  $x(t) = \frac{x_-(1+t) + x_+(1-t)}{2}$  and  $y(t) = \frac{y_-(1+t) + y_+(1-t)}{2}$  for  $t \in [-1, 1]$ . Note that  $y(t) \leq \frac{2y_- + 2y_+}{2} = 2y$ . Denote by  $b(t)$  the function  $B(x(t), y(t))$ . Note that  $(x(t), y(t))$  are in the domain  $\{x, y > 0; 1 \leq xy \leq 2A\}$ . Then

$$\begin{aligned} -b''(t) &= (x'(t), y'(t))(-d^2 B)(x'(t), y'(t))^t \\ &\geq (x'(t), y'(t)) \begin{pmatrix} 0 & 1/y(t) \\ 1/y(t) & 0 \end{pmatrix} (x'(t), y'(t))^t, \end{aligned}$$

and

$$\begin{aligned} -b''(t) &= (x'(t), y'(t))(-d^2 B)(x'(t), y'(t))^t \\ &\geq (x'(t), y'(t)) \begin{pmatrix} 0 & -1/y(t) \\ -1/y(t) & 0 \end{pmatrix} (x'(t), y'(t))^t. \end{aligned}$$

Therefore, evaluating  $x'(t)$  and  $y'(t)$  we have

$$-b''(t) \geq \frac{2}{y(t)} \left| \frac{(x_- - x_+)(y_- - y_+)}{2} \right| \geq \frac{1}{4y} |(x_- - x_+)(y_- - y_+)|.$$

By the definition of  $b$ ,

$$\begin{aligned} B(x, y) - \frac{B(x_-, y_-) + B(x_+, y_+)}{2} &= b(0) - \frac{b(1) + b(-1)}{2} \\ &= -1/2 \int_{-1}^1 b''(t)(1 - |t|)dt \geq C \left| (x_- - x_+)(y_- - y_+) \frac{1}{y} \right|. \end{aligned}$$

Now we are ready to run the usual Bellman function argument to prove the lemma.

Let  $x = w_J, y = w_J^{-1}, x_- = w_{J_l}, x_+ = w_{J_r}, y_- = w_{J_l}^{-1}, y_+ = w_{J_r}^{-1}$  and  $A = \|w\|_{A_2}$ . These  $x, y, x_+,$  etc. satisfy the conditions for (\*). Therefore, we have

$$c|w_{J_l} - w_{J_r}| |w_{J_l}^{-1} - w_{J_r}^{-1}| \frac{1}{w_J^{-1}} + \frac{B(w_{J_l}, w_{J_l}^{-1}) + B(w_{J_r}, w_{J_r}^{-1})}{2} \leq B(w_J, w_J^{-1}).$$

Now  $x_- y_-$  and  $x_+ y_+ \leq A = \|w\|_{A_2}$  again, so we can use estimate (\*) on the Bellman functions on the left side, too. This process can be repeated as often as we want. After  $n$  iterations, we have the following formula:

$$c \sum_{I \subset J, |I| \geq 2^{-n}|J|} \frac{|c_I|}{\sqrt{|I|}} \frac{|d_I|}{\sqrt{|I|}} w_I \frac{|I|}{|J|} + \text{positive terms} \leq B(w_J, w_J^{-1}) \leq 5 \|w\|_{A_2} w_J.$$

So letting  $n$  go to  $\infty$ , we get the desired estimate.  $\square$

## 5. Dividing the estimate up into 4 sums

To begin with, let us estimate the norm of the martingale transform by duality,

$$\begin{aligned} \|T_r f\|_{L^2(w)} &= \sup_{\|g\|_{L^2(w^{-1})}=1} \int T_r f g \, dx \\ &= \sup_{\|g\|_{L^2(w^{-1})}=1} \int \sum_{I, J} r(I) \langle f, h_I \rangle h_I \langle g, h_J \rangle h_J \, dx. \end{aligned}$$

Since the  $h_I$ 's are orthonormal in  $L^2(dx)$  this becomes

$$\sup_{\|g\|_{L^2(w^{-1})}=1} \sum_I r(I) \langle f, h_I \rangle \langle g, h_I \rangle.$$

Replacing  $f$  by  $f w^{1/2}$ , and  $g$  by  $g w^{-1/2}$  this becomes:

$$\|T_r\|_{L^2(w) \rightarrow L^2(w)} = \sup_{\|g\|_{L^2(dx)}=1} \sup_{\|f\|_{L^2(dx)}=1} \sum_I r(I) \langle f w^{-1/2}, h_I \rangle \langle g w^{1/2}, h_I \rangle.$$

In order to estimate this, it is convenient to express the Haar functions in terms of a different family, which is more suited to working with weights.

The following ‘‘Haar functions’’ for  $L^2(w)$  are normalized and orthogonal in  $L^2(w)$ :

$$h_I^w(x) = \frac{h_I(x) + \gamma_w^I \chi_I}{\delta_w^I},$$



where  $\gamma_w^I = \frac{-c_I}{|I|}$  and  $\delta_w^{I^2} = w_I(1 - c_I^2/|I|) = \frac{w_{I_l}w_{I_r}}{w_I}$  are chosen to make these functions an orthonormal family.

Of course, there is the equivalent family for  $w^{-1}$ , with  $\gamma_{w^{-1}}^I = \frac{-d_I}{|I|}$  and  $\delta_{w^{-1}}^{I^2} = w_I^{-1}(1 - d_I^2/|I|) = \frac{w_{I_l}^{-1}w_{I_r}^{-1}}{w_I^{-1}}$ .

Substituting  $h_I(x) = \delta_w^I h_I^w(x) - \gamma_w^I \chi_I$  (or the equivalent for  $w^{-1}$ ) into our equation, and sorting the different types of terms, we can write what we are trying to estimate as four sums  $I + II + III + IV$

$$\begin{aligned} I &: \sum_{I \in D[0,1]} r(I) \langle fw^{-1/2}, h_I^{w^{-1}} \rangle \delta_{w^{-1}}^I \langle gw^{1/2}, h_I^w \rangle \delta_w^I, \\ II &: - \sum_{I \in D[0,1]} r(I) \langle fw^{-1/2}, \chi_I \rangle \gamma_{w^{-1}}^I \langle gw^{1/2}, h_I^w \rangle \delta_w^I, \\ III &: - \sum_{I \in D[0,1]} r(I) \langle fw^{-1/2}, h_I^{w^{-1}} \rangle \delta_{w^{-1}}^I \langle gw^{1/2}, \chi_I \rangle \gamma_w^I, \\ IV &: \sum_{I \in D[0,1]} r(I) \langle fw^{-1/2}, \chi_I \rangle \gamma_{w^{-1}}^I \langle gw^{1/2}, \chi_I \rangle \gamma_w^I. \end{aligned}$$

We will estimate each sum separately in absolute value.

**5.1. Sum I.** Since  $r(I)$  could be any combination of signs, we will sum in absolute value.

$$I : \sum \left| \langle fw^{-1/2}, h_I^{w^{-1}} \rangle \langle gw^{1/2}, h_I^w \rangle \right| \sqrt{\frac{w_{I_l}w_{I_r}w_{I_l}^{-1}w_{I_r}^{-1}}{w_I w_I^{-1}}}.$$

Since  $w_{I_r}w_{I_l} \leq (w_I)^2$  and  $w_{I_r}^{-1}w_{I_l}^{-1} \leq (w_I^{-1})^2$ , this is bounded by

$$\begin{aligned} \sum \left| \langle fw^{-1/2}, h_I^{w^{-1}} \rangle \langle gw^{1/2}, h_I^w \rangle \right| \sqrt{\mu(I)} \\ \leq \|w\|^{1/2} \sum \left| \langle fw^{-1/2}, h_I^{w^{-1}} \rangle \langle gw^{1/2}, h_I^w \rangle \right|. \end{aligned}$$

Taking the inner product in  $L^2(w)$  and  $L^2(w^{-1})$  instead of in  $L^2(dx)$ , this becomes

$$\|w\|^{1/2} \sum \left| \langle fw^{1/2}, h_I^{w^{-1}} \rangle_{w^{-1}} \langle gw^{-1/2}, h_I^w \rangle_w \right|.$$

Since  $h^w$  and  $h^{w^{-1}}$  are orthonormal in  $L^2(w)$  and  $L^2(w^{-1})$ , Cauchy Schwarz and the Bessel inequality allow us to estimate this by

$$\|w\|^{1/2} \left\| fw^{1/2} \right\|_{L^2(w^{-1})} \left\| gw^{-1/2} \right\|_{L^2(w)} = \|w\|^{1/2} \|f\|_{L^2(dx)} \|g\|_{L^2(dx)}.$$

**5.2. Sum II and Sum III.** Sums II and III are equivalent, so we will show Sum II only.

$$\begin{aligned} & \sum_{I \in D[0,1]} \left| \langle fw^{-1/2}, \chi_I \rangle \gamma_{w^{-1}}^I \langle gw^{1/2}, h_I^w \rangle \delta_w^I \right| \\ &= \sum_{I \in D[0,1]} \left| (fw^{-1/2})_I |I| \frac{|d_I|}{|I|} \langle gw^{1/2}, h_I^w \rangle \sqrt{w_I(1 - c_I^2/|I|)} \right|. \end{aligned}$$

Using the fact that  $(1 - c_I^2/|I|) \leq 1$  and applying the Cauchy-Schwarz inequality, this becomes

$$\left( \sum_{I \in D[0,1]} |(fw^{-1/2})_I|^2 |d_I|^2 w_I \right)^{1/2} \left( \sum_{I \in D[0,1]} \langle gw^{1/2}, h_I^w \rangle^2 \right)^{1/2}.$$

The second of these terms can again be estimated as in Sum I, and is bounded by  $\|g\|_{L^2(dx)}$ . To estimate the first term, we will use Theorem 4.1 with  $w^{-1}$  for  $w$ , and  $\alpha_I = d_I^2 w_I$ , and the inequality from Remark 4.5:

$$\frac{1}{|J|} \sum_{I \subset J} (w_I^{-1})^2 d_I^2 w_I \leq c \|w\|_{A_2}^2 w_J^{-1}.$$

Theorem 4.1 gives us

$$\left( \sum_{I \in D[0,1]} |(fw^{-1/2})_I|^2 |d_I|^2 w_I \right)^{1/2} \leq c \|w\|_{A_2} \|f\|_{L^2(dx)}.$$

Therefore Sum II is bounded by  $c \|w\|_{A_2} \|f\|_{L^2(dx)} \|g\|_{L^2(dx)}$ .

**5.3. Sum IV.**

$$IV : \sum_{I \in D[0,1]} \left| (fw^{-1/2})_I |I| (gw^{1/2})_I |I| \frac{|c_I d_I|}{|I|^2} \right|$$

In order to estimate this sum, we will make use of Theorem 4.2. with  $\alpha_I = \frac{|c_I d_I|}{|I|}$ . Since  $w$  and  $w^{-1}$  have the same  $A_2$  norm, it is sufficient to calculate one of the conditions for the theorem.

$$\begin{aligned} & \frac{1}{|J|} \int_J \left( \sum_{I \subset J} \alpha_I w_I \chi_I(x) \right)^2 w^{-1} dx \\ &= \frac{1}{|J|} \sum_{I, K \subset J} \alpha_I w_I \alpha_K w_K \int_J \chi_I(x) \chi_K(x) w^{-1}(x) dx. \end{aligned}$$

We can break this sum into the case where  $I = K$  or  $I \subset K$ , since  $K \subset I$  is equivalent to the latter. So we get

$$\frac{1}{|J|} \left( \sum_{I \subset J} \alpha_I^2 (w_I)^2 w_I^{-1} |I| + 2 \sum_{I, K, K \subset I \subset J} \alpha_I \alpha_K w_K w_I w_K^{-1} |K| \right).$$

The first sum is

$$\frac{1}{|J|} \sum_{I \subset J} \left( \frac{c_I}{\sqrt{|I|}} \right)^2 \left( \frac{d_I}{\sqrt{|I|}} \right)^2 w_I |I| \mu(I).$$

Because  $w$  is an  $A_\infty$  weight,  $|\frac{d_I}{\sqrt{|I|}}| \leq 1$  (see [4]), and so we can estimate the above by

$$\frac{1}{|J|} \sum_{I \subset J} \left( \frac{c_I}{\sqrt{|I|}} \right)^2 w_I |I| \mu(I)$$

which is  $\leq c \|w\|_{A_2}^2 w_J$  by Remark 4.5.

This leaves us to estimate the second sum

$$\begin{aligned} \frac{2}{|J|} \sum_{I, K, K \subset I \subset J} \frac{|c_I|}{\sqrt{|I|}} \frac{|d_I|}{\sqrt{|I|}} \frac{|c_K|}{\sqrt{|K|}} \frac{|d_K|}{\sqrt{|K|}} w_I \mu(K) |K| \\ = \frac{2}{|J|} \sum_{I \subset J} \frac{|c_I|}{\sqrt{|I|}} \frac{|d_I|}{\sqrt{|I|}} w_I \sum_{K \subset I} \frac{|c_K|}{\sqrt{|K|}} \frac{|d_K|}{\sqrt{|K|}} \mu(K) |K|. \end{aligned}$$

The inside sum, by lemma 4.7 is bounded by  $c \|w\|_{A_2} |I|$ . That leaves us to estimate

$$\|w\|_{A_2} \frac{2}{|J|} \sum_{I \subset J} \frac{|c_I|}{\sqrt{|I|}} \frac{|d_I|}{\sqrt{|I|}} w_I |I|$$

which, by lemma 4.8 is bounded by  $c \|w\|_{A_2}^2 w_J$ . Remembering that the final estimate for sum IV will be the square root of the estimate for the above, we have that sums I to IV are each individually bounded by  $c \|w\|_{A_2} \|f\|_{L^2(dx)} \|g\|_{L^2(dx)}$ , which concludes the proof.

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