# An High Equipped Power Constrained Algorithm For OLEDS Based On MSR

Dayaker S (Pg Scholar)<sup>1</sup> Yakub M.Tech Assistant Professor<sup>2</sup>

Department of ECE, HARSHITH GROUP OF INSTITUTIONS, JNTU (HYD)

**ABSTRACT:** This paper presents a power-constrained contrast enhancement algorithmic program for organic semiconductor diode display based on multi scale retinex (MSR). In general, MSR, that is the key element of the planned algorithmic program, consists of power controllable log operation and sub band wise gain management. First, we decompose associate input image to MSRs of various sub-bands, and figure a correct gain for every MSR. Second, we have a tendency to apply a coarse-to-fine power management mechanism, that re computes the MSRs and gains. This step iterates till the target power saving is accurately accomplished. With video sequences, the contrast levels of adjacent pictures are determined systematically using temporal coherence so as to avoid unsteady artifacts. Finally, we gift many improvement skills for data processing. Experimental results show that the planned algorithmic program provides better visual quality than previous strategies, and a consistent power-saving magnitude relation while not unsteady artifacts, even for video sequences

*Index Terms* — Power consumption, contrast enhancement, OLED, multi-scale retinex.

# I. INTRODUCTION

MODERN show panels will be classified into emissive and non-emissive displays. The cathode-ray tube(CRT), the plasma board (PDP) and also the organic light-weight emitting diode (OLED) square measure representative emissive displays that don't need external light sources, whereas the thin-film transistor liquid show (TFT-LCD) is non-emissive. In general, Emissive displays have many blessings over r non-emissive ones [1], [2]. First, since associate degree emissive display can close up individual pixels, it will specific complete darkness and win a high distinction quantitative relation. Second, emissive shows consume less power than non-emissive ones as a result of every component in associate degree emissive display will be severally driven and the power consumption of the pixel is proportional to its intensity level. Note that non-emissive displays ought to activate their backlight despite component intensity. Thus, the OLED is thought to be the foremost promising candidate for the next-generation show [3], [4], which can replace the TFT-LCD displays presently dominating the business market. Even if the OLED is especially used for small panels in mobile devices, its mass-production technology is being developed quickly. Therefore large-size OLED

panels may soon be adopted in a very wider vary of devices like high definition TV (HDTV) and radical video [3].Note that show modules consume most of the facility in digital media devices [5]. Therefore techniques to attenuate power consumption within the show square measure inevitably needed. Several image process techniques for power saving in show panels are projected, on the far side circuit-level power savings. Unfortunately, such techniques specialize in reducing backlight intensity for TFT-LCDs whereas conserving a similar level of perceived quality [6]–[10]. In order that they can't be applied to power saving in emissive show devices like OLED. Lee et al. projected a powerconstrained contrast enhancement algorithmic rule (PCCE) for emissive displays based on bar graph deed (HE) [11], [12] They developed a power-consumption model for OLED displays and formulated an objective operate that consists of the HE term and the power term. By minimizing the target operate based mostly on the bell-shaped optimization theory; they tried to simultaneously achieve distinction improvement and power savings. However, with HE-based distinction improvement there's associate degree inherent risk of overstretching. Also, since their methodology depends certain parameters it cannot mechanically, on

systematically and accurately provide the power-saving level desired by a user. On the opposite hand, numerous distinction improvement techniques have been developed for dozens of years [13]–[20]. Mutually of them retinex may be a well-known non-linear improvement method used for distinction improvement further as dynamic range compression. The retinex theory was projected by Land and McCann [13], and Jobson et al. custom-made their theory to single scale retinex (SSR) [14] and multiscale retinex (MSR) [15].Retinex theory assumes that the human sensory system has three freelance ways that to understand short, medium, and long wavelengths within the actinic radiation spectrum. supported the retinextheory, SSR utilizes mathematician low pass filter (LPF) and log operation to intensify a selected wavelength vary of the image, associate degreed MSR offers an output image because the weighted total of the retinex output pictures by victimization many linear LPFs having different support regions. This paper proposes a power-constrained distinction improvement algorithmic rule employing a sub-band rotten MSR (SD-MSR) [16] for OLED show. First, we have a tendency to designed modified log operate for dynamic power saving. Second we have a tendency to propose a coarseto-fine power management mechanism based on SD-MSR, that together achieves distinction enhancement and dynamic vary compression victimization associate degree adaptation weighting strategy correct for associate degree input image. Finally, we have a tendency to gift power management theme for a continuing power reduction ratio in video sequences by victimization temporal coherence in video sequences. Experimental results show that the projected algorithmic rule provides higher visual quality than previous ways, and a consistent power-saving quantitative relation while not flicker artifacts for video sequences.

## II.SUB-BANDDECOMPOSEDMULTI-SCALERETINEX

MSR is an extended SSR with multiple kernel windows of different sizes. MSR output is a weighted sum of several different SSR outputs . The MSR output for a single spectral component can be represented as

$$R^{MSR}(x, y) = \sum_{n=1}^{N} w_n \cdot R_n(x, y)$$
(1)

where

$$R_n(x,y) = logI(x,y) - log(F_n(x,y) * I(x,y))$$
(2)

Here  $R_n(x, y)$ , denotes a retinex output associated with the n-the scale for an input image I(x, y). Note that gain wn is determined so that it can satisfy the condition of  $\sum w_n = 1$ . The symbol "\*" in Eq. (2) denotes the convolution operation and Nis the number of scales.  $F_n(x, y)$  Denotes a surround function and is given by

$$F_n(x, y) = K_n e^{(x^2 + y^2)} / \sigma_n^2$$
(3)

Where  $K_n$  is determined so that  $F_n(x, y)$  can satisfy  $\sum \sum F_n(x, y) = 1. \ \sigma_n^2$ denotes the variance of the Gaussian kernel at then-the sub-band. Under the condition  $\sigma_n > \sigma_{n-1}$  every SSR, we can derive successive frequency sub-bands. Note that a small is suitable for enhancing fine details, whereas a Largent is suitable for improving tonality. Thus, it is important to select an appropriate value of an in the MSR. Based on this rationale, Jang et al. proposed an SD-MSR that consists modified logarithmic function, sub-band of a decomposition, space varying sub-band gain, and an automatic gain/offset control [16] (see Fig. 1). The modified log (mlog) is defined as

 $\begin{aligned} mlog(I(x,y)) &= \\ \begin{cases} w_L \log(I(x,y)+1) & I(x,y) \leq \tau \\ -w_H \log(D-I(x,y)) + logD) & I(x,y) > \tau \end{aligned}$ 

Where  $\tau$  is a user-defined threshold and D denotes an image dynamic range. For example, D is 256 for an 8-bit image



Figure1 :block diagram of conventional SD-MSR

 $w_L$  And  $w_L$  denote weighting parameters according to and are defined as

$$w_{L} = \frac{\frac{\tau}{D-1} log D}{log(\tau+1)}, \ w_{L} = \frac{(1-\frac{\tau}{D-1}) log D}{log(D-1)}$$
(5)

s a result, the mlog function of Eq. (4) enhances the contrasts of dark regions as well as bright regions. In this way, we can enhance image details both in highlights and shadows. Another feature of SD-MSR is to decompose the modified retinex outputs into nearly non-

overlapping spectral bands. The following equation accomplishes this sub-band decomposition:

$$\bar{R}_1 = R_1 \qquad n = 1$$
  
$$\bar{R}_n = R_n - R_{n-1} \qquad 2 \le n < N \tag{6}$$

As n increases, Rn corresponds to the low frequency region n more and more. Here, Rn is computed by replacing the log of Eq. (2) With the mlog of Eq. (4) Next, the space vary in g sub-band gain at then-the sub-band is defined as

$$g_n(x,y) = \left(\frac{1}{NR_n(x,y) + \varepsilon_g}\right)^{1 - \frac{\sigma_n}{\sigma max + \varepsilon\sigma}}$$
(7)

Where

$$\sigma_{max} = \max_{n \in [1,2,3..N]} \sigma_n$$

$$NR_n(x, y) = \frac{|\bar{R}_n(x,y)|}{\bar{R}_{nmax}}$$
(8)

In a high spectral band of small, they make the gain difference between pixels larger, especially for the pixels with low  $NR_n(x, y)$ . This is because this spectral band has large high-frequency components representing image details. Meanwhile, they lower the gain difference between pixels in a high spectral band of large n to maintain the characteristics of a natural scene. Thus, using Eq. (7), the final enhanced imagelis output as follows

$$I' = \sum_{n=1}^{N} g_n \,\widetilde{R_n} \tag{9}$$

## **III.THE PROPOSED ALGORITHM**

We propose a power governable distinction enhancement algorithm for OLED show primarily based on SD-MSR. Fig. 2 describe the projected formula that consists of three stages. the primary stage coarsely reduces the facility of Associate in Nursing input image nearer to the target power with distinction improvement, and the second stage finely controls the image power such that it's terribly near the target power. If the input could be a video sequence, the ultimate stage adjusts the facility of every image so that it is like those of its neighbors by considering the temporal coherence of the input video sequence. The projected formula is differentiated from previous methods in the following 3 aspects. First, we tend to control the target power level mechanically. Second, we tend to avoid theflickering development by keeping the facility levels of adjacent images constant for video sequences. Third, we tend to

come through time period process of the projected formula on a all-purpose graphics process unit (GPU) even for full HD video sequences



Figure 2: block diagram of proposed algorithm

Image nearer to the target power with distinction improvement and the second stage finely controls the image power such that it's terribly near the target power. If the input may be a video sequence, the ultimate stage adjusts the ability of every image so that it's the same as those of its neighbors by considering the temporal coherence of the input video sequence. The projected algorithmic program is differentiated from previous methods within the following 3 aspects. First, we have a tendency to control the target power level mechanically. Second, we have a tendency to avoid the flickering development by keeping the ability levels of adjacent images constant for video sequences. Third, we have a tendency to bring home the bacon real-time process of the projected algorithmic program on a general purpose graphics process unit (GPU) even for full HD video sequences.

#### A. Power Modeling in OLED Display

Before presenting an in depth clarification of the projected algorithmic program, we want to model power for associate OLED show. Dong et al. conferred a pelbased power model that estimates the ability consumption of OLED modules supported the red greenblue (RGB) specification of every pixel [21]. the ability consumption of associate OLED show with K pixels, i.e., P is

$$P_{OLED} = C + \sum_{i=1}^{K} (f_R(R_i) + f_G(G_i) + f_B(B_i)) \quad (10)$$

Also, we consider only the Y-componentbecause it dominates the entire overall power. Note that the Y-component indicates the luminance component in YUVcolor format. So we use the Y-component power consumption(YP) of an OLED display with Kpixels [11] as

$$Y_P = \sum_{i=1}^{K} Y_i^{\gamma} \tag{11}$$

Where  $\gamma$  is a parameter for gamma correction for a given display device

#### **B.** The Proposed Algorithm

This section details the proposed algorithm.1) Coarse Control Stage: The mlog of conventional SD-MSR plays a role in enhancing the contrasts of highlights and shadow regions. In other words, contrast in the dark region becomes high by increasing the intensity level of the pixels in the region, and contrast in the bright region also becomes high by decreasing the intensity level of the pixels in the region. However, the increase of the intensity values in the shadow region results in the increase in power consumption for the OLED display. So, for low power consumption as well as contrast enhancement, even in the shadow region, we redefine a so-called power-constrained log (plog) from them log of Eq. (4) as follows

$$plog(I(x,y)) = \begin{cases} \frac{\tau log D log(\alpha I(x,y)+1)}{(D-1) log(\alpha \tau+1)} & I(x,y) \le \tau\\ m log(I(x,y)) & I(x,y) > \tau \end{cases}$$
(12)

Therefore, the *plog* of Eq. (12) has the effect of controlling the increase in power consumption while partially lowering the contrast in the dark region. From Eq. (7) and MSRs computed by *plog*, i.e.,  $\{Rn\}$ , we can derive the following output image

$$\tilde{R}_t = \sum_{n=1}^N g_n \dot{R}_n \tag{13}$$

On the other hand, basin YP on Eq. (11), the power reduction ratio of an input image and its output image is defined as follows

$$p_t = 1 - \frac{YP(\hat{\hat{R}}_t)}{YP(I)} \tag{14}$$

**IV. SIMULATION RESULTS** 

In this paper,  $\overline{R}_n$  can be computed with Eq. (15) as in [16].

$$f(X) = X^{N} = \frac{X - m}{M - m}(L - 1) + l$$
(15)

Let  $\delta_t$  denote the difference between  $p_t$  and P as in Eq. (16)

$$\delta_t = P - p_t \tag{16}$$

Eq. (17) because such a condition indicates an excess of power reduction over P.

$$\tau_{t+1} = \tau_t + \frac{(D - \tau_t)}{2}$$
 (17)

we increase  $\tau$  relatively small as in Eq. (18) because  $\delta Ct$  weakly over runs P

$$\tau_{t+1} = \tau_t + \frac{(D - \tau_t)}{4}$$
(18)

So we approach P by decreasing  $\tau$  relatively small as in Eq. (19).

$$\tau_{t+1} = \tau_t + {\tau_t / 4}$$
(19)

So we rapidly approach P by decreasing  $\tau$  significantly.

$$\tau_{t+1} = \tau_t + \frac{\tau_t}{2}$$
 (20)

On the other hand, the low-frequency region is rarely related to image details, but it dominates image power as a whole. So we try to approach P by finely controlling the proportion of the lowest-band MSR which may have most of the image power. In detail, we control the gain of RN as follows:

$$R^{*} = \sum_{n=1}^{N-1} g_n \bar{R}_n + (g_N + \lambda) \bar{R}_N \qquad (21)$$

Where  $\lambda$  indicates a control parameter for the lowestband MSR. $\lambda$ , which is updated according to Eq. (22) enables the FCS to approach the target power with little change of contrast

$$\lambda_{t+1} = \lambda_t - \delta_t^F \tag{22}$$



Figure 3: Profile of the proposed algorithm



Figure 5: Error analysis w.r.t. no of iterations

The performance of the proposed algorithm can be evaluated by choosing the two images from Kodak Lossless True Color Image Suite1(caps and beach) and a **CONCLUSION** 

This paper proposes Associate in Nursing SD-MSR-based image processing algorithm for fine power management in OLED displays. We designed a powerconstrained log perform for effective power saving in regions. victimization dark the power-constrained log function for SD-MSR Associate in Nursing Associate in Nursing adjective weight strategy proper for an input image, we tend to projected a coarse-to-fine power control mechanism for still pictures. Finally, we tend to given a power management theme for a constant power reduction ratio in video sequences by victimization temporal coherence in video sequences. Experimental results showed that the proposed algorithm provides higher visual quality than previous works, and a consistent power-saving magnitude relation while not the flickering artifact even for video



Figure 4: Output results of the proposed algorithm, Linea algorithm and PCCE algorithm

high dynamic range (HDR) test image memorial. Also, we employed six common intermediate format (CIF) video sequences: container(500 frames),football(90 frames), Paris(300 frames), foreman (300 frames), bus(500 frames),Stefan(90 frames) and five 720p sequences: big ship (60 frames), crew (60 frames), jets (60 frames), night(60 frames), raven (60 frames), and four 1080p sequences: crowd run(500 frames), park joy(60 frames), toys and calendar(60 frames) traffic (60 frames). We processed only the luminance components in the experiments. More specifically, given a color image, we converted it to the YUV color space and then process only the Y-component without modifying the Uand V-components.

sequences. Specifically, the proposed algorithm provides at most twelve months and on average13% higher edge-preserving ratios than the state-of-the-art algorithm (i.e., PCCE [11]). In addition, we tend to tried the possibility of real-time operation by accomplishing Associate in nursing entire execution time of nine ms per 1080p image.

With video sequences, the contrast levels of adjacent pictures are determined systematically using temporal coherence so as to avoid unsteady artifacts. Finally, we gift many improvement skills for data processing. Experimental results show that the planned algorithmic program provides better visual quality than previous strategies, and a consistent power-saving magnitude relation while not unsteady artifacts, even for video sequences

# TABULAR COLUMNS

C	OMPARIS EN	ON IN T HANCE	TERMS ( MENT N	OF SHA IETRIC	RPNESS	
NAM E	P=10%			P=30%		
	Linear	PCC E	Prop osed	Line ar	PCC E	Pro pose d
Bikes	4.9529	5.19 46	5.79 83	14.8 588	15.57 96	17.3 672
Buildi ngs	5.8787	6.11 45	6.72 97	17.6 361	18.39 93	20.2 419
Cemet ery	4.5727	4.80 38	5.38 14	13.7 181	14.39 61	16.1 338
House	2.4153	2.81 92	2.86 18	7.24 59	8.311 8	8.49 86
Light house	3.0829	3.41 77	3.58 55	9.24 87	10.01 78	10.6 978
Monar ch	4.0227	4.52 32	5.03 17	12.0 681	13.39 54	15.0 005
Ocean	1.9568	2.23 76	2.46 88	5.87 05	6.639 8	7.36 65
Parrot	2.0217	2.23 56	2.38 93	6.06 52	6.585 2	7.08 10
Plane	1.9168	2.00 40	2.15 97	5.75 05	5.987 6	6.44 92
Rapids	3.7163	3.97 52	4.31 99	11.1 489	11.81 97	12.8 169
Statue	2.5265	2.86 98	3.07 07	7.57 94	8.486 7	9.23 62
Woma n	3.4289	3.76 62	3.97 99	10.2 868	11.14 70	11.8 233

COMPARISON IN TERMS OF THE EME VALUE										
NAME	P=10%			P=30%						
	Linea	PCC	Propo	Line	PCC	Prop				
	r	E	sed	ar	E	osed				
Bikes	9.609	18.2	63.59	9.60	14.9	63.9				
	4	726	24	94	232	642				
Buildin	5.167	7.39	20.42	5.16	6.89	20.8				
gs	4	84	97	74	14	775				
Cemete	6.331	15.6	5017	6.33	12.2	48.9				
ry	5	733	56	15	051	967				
House	4042	15.3	27.31	4.42	13.1	26.4				
	13	661	82	16	372	888				
Light	6.649	14.0	31.02	6.64	11.2	29.5				
house	0	490	58	90	048	062				
Monarc	4.906	35.8	57.95	4.90	25.8	56.7				
h	2	532	89	62	058	810				
Ocean	4.854	29.9	59.24	4.85	25.6	58.9				
	0	294	86	40	026	096				
Parrot	7.936	18.0	29.45	7.93	11.0	25.8				
	1	644	59	61	465	077				
Plane	6.621	10.6	49.34	6.62	10.0	48.9				
	6	665	52	16	051	671				
Rapids	7.543	20.9	57.99	7.54	11.8	56.5				
	4	534	40	34	584	294				
Statue	7.741	11.1	79.87	7.74	9.56	80.3				
	6	254	40	16	45	500				
Woman	7.055	30.0	80.52	7.05	19.3	80.4				
	7	266	96	57	998	561				

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