Automatic Defect Inspection of Patterned TFT-LCD Panels
Using 1-D Fourier Reconstruction and Wavelet Decomposition

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Abstract

Large-sized flat-panel displays have become increasingly important for use in PC monitors and TVs. In this paper we have considered the problem of automatic visual inspection of micro-defects including pinholes, scratches and particles in patterned Thin Film Transistor-Liquid Crystal Display (TFT-LCD) panel surfaces. For large-sized TFT-LCD panel inspection, high resolution line scan is demanded. We propose a global 1-D Fourier-based image reconstruction scheme that directly works on the 1-D line images, instead of the traditional 2-D area images. The proposed method fully utilizes the inherent geometric structure of a TFT-LCD panel. It first eliminates the frequency components that represent the periodic pattern of a TFT-LCD line image in the 1-D Fourier spectrum, and then back-transforms the 1-D Fourier-domain image to the 1-D spatial domain image using the inverse Fourier transform. The Fourier reconstruction process can effectively remove the patterned background and distinctly preserve local anomalies in the resulting 1-D image. Wavelet decomposition is further applied to remove uneven illumination in the filtered image so that defects can be easily segmented with simple statistical control limits. Experimental results on a number of micro-defects embedded in TFT-LCD panels show that the proposed method can reliably detect various ill-defined defects without designing and measuring the quantitative features of individual defect types.

Keywords: Defect detection; TFT-LCD; Line scan; 1-D Fourier transforms; Machine vision
1. Introduction

Flat-panel displays have become increasingly important in recent years. They can be used as small viewfinders for handheld devices such as cellular phones and PDAs, and as larger monitors for notebook and personal computers. Thin Film Transistor-Liquid Crystal Displays (TFT-LCDs) are particularly attractive due to their full-color display capabilities, low power consumption and light weight. To ensure the display quality and improve the yield of LCD flat panels, the inspection of defects in TFT-LCD panels becomes a critical task in LCD manufacturing.

Surface defects of a TFT-LCD panel not only cause visual failure but also cause electrical failure to operate the panel. Appearance defects in TFT-LCD panels can be roughly classified into two categories: macro and micro defects (Nakashima 1994). Macro defects include unevenness of color on a TFT-LCD panel, stains on a panel, and misalignment of a panel. Micro defects include pinholes, scratches and particles in a TFT-LCD panel. The macro defects appear as high contrast regions with irregular sizes and shapes. They are generally large in size and, therefore, can be detected by human inspectors. However, micro defects are generally very small in size, and cannot be easily found by human personnel or detected with electrical methods.

There are currently several electrical and optical based inspection techniques available for TFT-LCD defect inspection (Kido 1992, Chen et al. 2000, Hawthorne 2000, Lin et al. 2001). Kido (1993) and Kido et al. (1995) presented an optical charge sensing technique for the inspection of active-matrix LCD panels. Charge sensing writes a charge to the storage capacitor at the individual TFT array pixel. The charge
is read back and compared with the write charge to determine if the pixel is defective (Hawthorne 2000). Henly and Addiego (1991) used an electro-optic modulator to generate voltage images. The voltage-imaging technique measures the characteristics of a TFT-LCD array by directly measuring the actual voltage distribution on the TFT pixel and not at the storage capacitor. The advantage of the optical charge and voltage imaging approaches is that each pixel on the TFT panel can be individually tested. The main disadvantage of such approaches is that the probes used for voltage measurement must be separately designed for each panel configuration. With the fast development of LCD technologies, the new-generation polysilicon TFT-LCD displays with integrated drivers (Nakamura et al. 2002, Nishibe 2002) do not need storage capacitors. The charge-sensing and voltage-imaging methods will become useless since they cannot test arrays without storage capacitors.

A few vision-based techniques were developed for LCD defect inspection. The processing algorithm is typically either a computationally intensive pattern matching, or a more straightforward differencing method (Hawthorne 2000). Nakashima (1994) presented an inspection system based on image subtraction and optical Fourier filtering for detecting defects in an LCD color filter panel. The image subtraction method was utilized to detect white and black defects such as black matrix holes and particles, and the optical Fourier filtering was applied for grain defects. Sokolov and Treskunov (1992) developed an automated visual system for final inspection of LCDs. They compared brightness distributions between a reference LCD image and a test image to detect anomalies in the surface. The existing vision-based techniques generally need a pre-stored reference image for comparison. This requires large data storage for the
reference image and precise environmental controls such as alignment and lighting for the TFT-LCD panel under inspection. Oh et al. (2004) studied the detection of line-type defects in TFT-LCD panels. In a low-resolution image, a directional filter bank (DFB) that finds directional information is used to identify horizontal or vertical line-shaped defects. In a high resolution image, a multi-level thresholding technique based on a statistical approach is employed to detect abnormal line defects that are brighter or darker than the surrounding pixels. Their approach is only applicable to line-shaped defects.

Most of the vision-based techniques mentioned previously are based on the analysis of grey levels in two-dimensional area images. Since there is an ongoing strong demand for the TFT-LCDs used in large-screen PC monitors and TVs, large-sized TFT-LCD panel manufacturing is getting important. For automatic visual inspection of TFT-LCD panels, very high-resolution image acquisition will be mandatory. High resolution line scan, therefore, will be demanded for the inspection of large-sized TFT-LCD panels. In this study, we propose an automatic visual inspection scheme that identifies defects directly from one-dimensional (1-D) line images of TFT-LCD panels. The proposed algorithms especially focus on the detection of micro-defects including pinholes, scratches and particles in TFT-LCD panel surfaces.

In the LCD manufacturing process, perpendicular data and gate conductive lines are patterned onto the glass substrate. A TFT-LCD panel comprises horizontal gate lines on one plane, and vertical data lines on the other plane. A thin film transistor is
located at each intersection of the data and gate lines (Pratt and Hawthorne 1998). Figure 1(a) shows the schema of a single pixel of a TFT-LCD panel (Kido 1993), and Figure 1(b) demonstrates the magnified two-dimensional (2-D) image of a TFT-LCD panel at the resolution of 60 pixels/mm. The geometric structure of a TFT-LCD panel surface involves repetitive, equally-spaced horizontal and vertical lines. Under a given resolution, the horizontal line patterns are repeated for every fixed number of the scan lines. The repeated cycle corresponds to the vertical spacing between two adjacent gate lines in a TFT-LCD panel surface. For each horizontal scan line, the distance between two adjacent peaks in the grey-level line image corresponds to the spacing between two neighbouring vertical data lines. Each scanned line profile exhibits periodicity and regularity properties. The problem of defect detection in TFT-LCD panel surfaces can now be considered as the identification of anomalies in a periodic 1-D grey-level signal. Note that the periodic pattern for each input line image is different to some extent, as seen in Figure 2.

The proposed method does not rely on the design of quantitative features to describe local anomalies, nor does it require a reference template for comparison. It alleviates all limitations of the feature extraction and template matching approaches. It is based on an image reconstruction scheme that directly works on the 1-D input images using the 1-D Fourier transforms. The Fourier spectrum is ideally suited for describing the periodicity of the TFT-LCD image signal in question. The proposed 1-D Fourier-based image reconstruction scheme fully utilizes the inherent geometric structure of TFT-LCD panels. It first eliminates the frequency components that represent the periodic peak pattern of the 1-D grey level profile of a TFT-LCD line.
image in the 1-D Fourier spectrum, and then back-transforms the 1-D Fourier-domain image using the 1-D inverse Fourier transform. The Fourier reconstruction process can effectively remove the background of periodic peaks and distinctly preserve local anomalies in the resulting 1-D image. For a faultless TFT-LCD line image, the reconstruction process will result in a flat horizontal line. Conversely, high fluctuation of a line will be generated in the reconstruction for a defective TFT-LCD line image. The simple statistical process control limits can then be used to set up the thresholds for distinguishing between defective segments and uniform background in each reconstructed 1-D line image.

This paper is organized as follows. Section 2 first discusses the 1-D Fourier transform and the properties of one-dimensional TFT-LCD line images in the Fourier spectra. The removal of repetitive background patterns in any scan lines of a TFT-LCD panel is then analyzed. Section 3 presents the experimental results from a variety of TFT-LCD micro-defects including pinholes, scratches and particles. Section 4 has conclusions of the paper.

2. The Fourier Reconstruction in 1-D Line Images

2.1. The 1-D Fourier transform

The present study uses machine vision to tackle the problem of detecting micro-defects embedded in TFT-LCD panel surfaces. The line images, instead of the area images, of a TFT-LCD panel obtained from a line scan system is directly used as the input for inspection in the manufacture of LCDs. Each scanned line image shows a periodic grey-level profile corresponding to the uniformly-spaced conductive data lines.
Figures 2(a-f) show a cycle of six line images for a faultless TFT-LCD panel at the resolution of 60 pixels/mm. The line images in Figures 2(a) and (b) are the grey-level profiles associated with horizontal gate lines in the TFT-LCD panel. No significant periodic peaks are observed in these two line images since they do not cross over conductive data lines. Figures 2(c-f) are the scanned lines that cross over equally-spaced data lines. Visible periodic peaks are observed in the profiles, in which each peak represents a data line. For the sake of comparison, Figures 3(a1-f1), 3(a2-f2) and 3(a3-f3) illustrate, respectively, six consecutive line images that cross over a pinhole, a scratch and a particle in TFT-LCD panels. The corresponding 2-D grey-level images of the three defect samples can be found in the experimental section of this paper. Defects of pinhole, scratch and particle obviously break the regularity of periodic structures and appear as local anomalies in the 1-D grey-level profiles.

In this study, the 1-D Fourier-based image reconstruction technique is directly applied to the 1-D grey-level line image to remove periodic peaks that represent data-line and gate-line patterns in TFT-LCD panel surfaces. The Fourier transform can characterize the line image of a periodic pattern in terms of frequency components. The periodically occurring features of data lines, gate lines, and TFTs of a TFT-LCD panel can be observed from the magnitude of frequency components. The global periodic pattern of a scan line is easily distinguishable as concentration of high-energy bursts in the Fourier spectrum.

Let \( f_y(x) \) be the grey level at pixel location \( x \) of a line image \( y \), and \( N \) the length of a scan line. The one-dimensional discrete Fourier transform of \( f_y(x) \) is
given by

\[ F_y(u) = \frac{1}{N} \sum_{x=0}^{N-1} f_y(x) \cdot \exp[-j2\pi ux / N] \]  

(1)

for frequency variable \( u = 0, 1, 2, \ldots, N - 1 \). The Fourier transform is generally complex; that is

\[ F_y(u) = R_y(u) + jI_y(u) \]  

(2)

where \( R_y(u) \) and \( I_y(u) \) are the real and imaginary parts of \( F_y(u) \), i.e.,

\[ R_y(u) = \frac{1}{N} \sum_{x=0}^{N-1} f_y(x) \cdot \cos(2\pi ux / N) \]  

(3)

\[ I_y(u) = \frac{1}{N} \sum_{x=0}^{N-1} f_y(x) \cdot \sin(2\pi ux / N) \]  

(4)

The power spectrum \( P_y(u) \) of \( F_y(u) \) is defined by

\[ P_y(u) = |F_y(u)|^2 = R_y^2(u) + I_y^2(u) \]  

(5)

### 2.2 Removal of periodic patterns

An arbitrary line image scanned from a faultless TFT-LCD panel surface consists
of periodic peaks in the grey-level profile. Each peak represents the location of an equally-spaced conductive data line on the plane. Let $\Delta x$ be the known spacing in pixel between two adjacent data lines (i.e., the distance between two neighbouring peaks) in the line image. $\Delta x$ is the period of a 1-D spatial domain image. The relationship between the frequency components in the Fourier domain image and spatial characteristics of the 1-D grey-level profile in the spatial domain image is given by

$$\Delta u = \frac{N}{\Delta x}$$  \hspace{1cm} (6)$$

where $N$ is the length of a scanned line image, and $\Delta u$ is the spacing between two adjacent peaks in the power spectrum. For illustration, Figures 4(a) and (b) depict $\Delta x$ and $\Delta u$ in a faultless line image of the TFT-LCD panel and the corresponding power spectrum, respectively.

In order to eliminate the periodic peaks that represent the data lines in the 1-D spatial domain image, the corresponding frequency components in the Fourier domain image are set to zero, and back-transformed to the spatial domain image using the 1-D inverse Fourier transform. Let $U$ be a set of frequency peaks in the Fourier power spectrum, and

$$U = \{u_1^*, u_2^*, ..., u_n^*\}$$

where $u_i^* = \frac{N}{2} \pm (i \cdot \Delta u)$, for $i = 1, 2, ..., n$ and $n = \text{INT}\left[\frac{N}{2 \cdot \Delta u}\right]$. The frequency
components $u_i^*$'s and their neighbours defined by $r$ are set to zero, i.e.,

$$F_y(u) = 0, \text{ for } u_i^* - r \leq u \leq u_i^* + r; \quad i = 1, 2, ..., n$$  \hspace{1cm} (7)

In this study, the $r$-distant neighbours are given by 3 at the image resolution of 60 pixels/mm. With the newly assigned values of $F_y(u)$, the 1-D spatial domain image can be reconstructed using the 1-D inverse Fourier transform:

$$\hat{f}_y(x) = \sum_{u=0}^{N-1} F_y(u) \cdot \exp[j2\piux/N]$$  \hspace{1cm} (8)

Figures 5(e) and (f) present, respectively, the reconstructed results of the faultless line image in Figure 5(a), and the defective line image of a pinhole in Figure 5(b). These results exhibit that the periodic peaks representing the data lines in the TFT-LCD panel are well eliminated, while the anomaly associated with the pinhole defect is distinctly preserved.

Both reconstructed line images in Figures 5(e) and (f) present minor variations in the grey-level profiles. Given the Fourier spectrum of a 1-D grey-level image, the frequency components in the neighbourhood of the center of the Fourier spectrum are low frequency bands that represent the coarse approximation of the original signal, and those apart from the center are high frequency bands that represent the details. Figure 5(d) indicates that the frequency components corresponding to a local anomaly in the spatial domain are around the center of the Fourier spectrum. Since a defect results
from the local anomaly and the background contains a global repetitive pattern, the Fourier transform will inherently present the details of the global background pattern in high frequency bands and the unique local anomaly in low frequency bands in the Fourier spectrum. To smooth further the resulting 1-D grey-level profile from eqs. (7) and (8), the high frequency components outside the region defined by the parameter \( w \) (see Figure 4(b) for the notation) are also set to zero, i.e.,

\[
F_y(u) = 0, \text{ for } 0 \leq u \leq w \quad \text{and} \quad N - w \leq u \leq N - 1
\]  

Eqs. (7) and (9) are simultaneously applied for back-transforming the Fourier domain image to the spatial domain image using eq. (8). Figures 5(g) and (h) illustrate the final reconstruction results of Figures 5(a) and (b) by implementing both eqs. (7) and (9), where \( r = 3 \) is applied in eq. (7), and \( w = \frac{3}{8}N \) in eq. (9). The results visibly reveal that all periodic peaks are effectively removed in both faultless and defective line images, while high fluctuation of grey levels associated with the defect is very well preserved.

Observing closely the resulting grey-level profiles in Figures 5(g) and (h), we find that the reconstructed line image yields significant grey-level variations in both ends of the profile, which may cause false alarms in defect detection of TFT-LCD panels. This is because that the implicit \( N \)-point periodicity of the DFT gives rise to boundary discontinuities that result in substantial high-frequency transform content (Gonzalez and Woods 2002). It causes the end points to take on erroneous values. In order to tackle
the problem of boundary discontinuities, each original line image is extended in both ends by duplicating their neighbouring grey-level segments of length $\Delta x$ (i.e., the spacing between two conductive data lines). That is,

$$f_y(i-\Delta x) = f_y(i), \quad i = 0, 1, 2, \ldots, \Delta x - 1$$

and

$$f_y(j+\Delta x) = f_y(j), \quad j = N - \Delta x, N - \Delta x + 1, \ldots, N - 1$$

The newly extended grey-level line image is applied to the Fourier reconstruction, and only the resulting image in the range $[0, N-1]$ is used as the effective profile for inspection. Figures 5(i) and (j) demonstrate that the extension procedure above can significantly improve the effect of boundary discontinuities.

**2.3. Removal of uneven illumination**

The proposed Fourier reconstruction procedure has shown its effectiveness on the removal of regularly-patterned background and the preservation of local anomalies in TFT-LCD panel images. However, as observed in Figure 5(i), the reconstructed line image is not a horizontal flat profile due to the unevenness of illumination on the TFT-LCD panel surfaces. This may cause the difficulty in the discrimination between uneven lighting and anomalies using simple statistical process control limits as the segmentation thresholds.

In this study, the unevenness of illumination is removed using the wavelet
transform. The reconstructed 1-D grey-level profile $\hat{f}_y(x)$ is decomposed into smooth and detailed parts using the Haar wavelet frame. The discrete wavelet frame decomposition with a Haar basis proceeds as follows.

$$\hat{s}_y(x) = \frac{1}{2} \sum_i L(i) \cdot \hat{f}_y(x+i)$$

(10)

$$\hat{d}_y(x) = \frac{1}{2} \sum_i H(i) \cdot \hat{f}_y(x+i)$$

(11)

where $L(i)$ and $H(i)$ are the lowpass and highpass coefficients, respectively, of the Haar basis (Burrus et al. 1998). $\hat{s}_y(x)$ is the decomposed smooth version of $\hat{f}_y(x)$, which is an approximation of the original grey-level profile. $\hat{d}_y(x)$ is the decomposed detail version of $\hat{f}_y(x)$, which entails only the fine detail of the original grey-level profile. Note that both $\hat{s}_y(x)$ and $\hat{d}_y(x)$ have the same length of the original grey-level profile $\hat{f}_y(x)$ because the filtered 1-D images are not subsampled. Figures 6 and 7 demonstrate the results of the wavelet frame decomposition from faultless and defective line images, respectively. It is apparent from Figures 6(c) and 7(c) that the smooth line image $\hat{s}_y(x)$ cannot remove the unevenness of illumination, and the detailed line image $\hat{d}_y(x)$ becomes flat for a faultless segment of a TFT-LCD line image, while the higher frequency oscillations corresponding to the defect segment is well signified in $\hat{d}_y(x)$. 
A simple dual thresholding based on the statistical process control limits can be therefore used to separate the defective segments from the uniform background in the decomposed detail \( \hat{d}_y(x) \). The upper and lower control limits for intensity variation in the decomposed detail image \( \hat{d}_y(x) \) of a filtered line image \( \hat{f}_y(x) \) are given by

\[
\mu_{\hat{d}_y} \pm K \cdot \sigma_{\hat{d}_y}
\]

(12)

where

\[
\mu_{\hat{d}_y} = \frac{1}{N} \sum_{x=0}^{N-1} \hat{d}_y(x);
\]

\[
\sigma_{\hat{d}_y} = \left\{ \frac{1}{N - 1} \sum_{x=0}^{N-1} \left[ \hat{d}_y(x) - \mu_{\hat{d}_y} \right]^2 \right\}^{1/2};
\]

\( K \) = a control constant.

Note that the mean \( \mu_{\hat{d}_y} \) and the standard deviation \( \sigma_{\hat{d}_y} \) of the decomposed detail \( \hat{d}_y(x) \) are calculated for each individual scan line \( y \), i.e., each input line image has its own control limits. In this study, we have used \( K = 3 \) for all scan lines to follow the typical 3-sigma standard in the statistical process control. Points in a uniform segment will lie within the control limits, and points in a defective segment will fall outside the control limits.

3. Experimental Results

In this section, we present experimental results from a variety of micro-defects including pinholes, scratches and particles in TFT-LCD panel surfaces to evaluate the
efficacy of the proposed 1-D Fourier reconstruction scheme. To visualize the detection results, the test images are displayed in two dimensions of size 256×256 pixels. That is, a 2-D test image comprises 256 scan lines, each of the length 256 pixels. The 3-sigma control limits were applied to all test images. Points falling within and outside the control limits are respectively represented by white and black intensities so that the detected defects can be easily observed and verified in the binarized image. In order to remove noise in the resulting binary image, a defect is assumed at least of the size 2×2 pixels. Therefore, a true defect must cross two consecutive scan lines at the same location in each of the two scan lines, i.e., a point at location \( x \) of a scan line \( y \) is classified as a part of the defect only if both \( \hat{d}_y(x) \) and \( \hat{d}_{y-1}(x) \) fall outside the control limits. This restriction can effectively remove isolated noisy points in the filtered image.

Figures 8(a-c) show, respectively, three defective 2-D images containing pinhole, scratch and particle blemishes in TFT-LCD panel surfaces at a fine resolution of 60 pixels/mm. Figure 8(d) illustrates a faultless 2-D image at the same resolution. The micro-defects of pinholes, scratches and particles can only be detected in images of fine resolution. They become blurred or invisible at resolutions under 50 pixels/mm.

Figures 9(a-d) present the detection results of the four test images in Figures 8(a-d), respectively. The 3-sigma standard is used for every scan line in Figures 8(a-d) to set up the control limits. Each one of the 2-D binary images in Figures 9(a-d) is the result of merging all the 256 filtered line images in their scanning sequence. In the defective images of Figures 8(a-c), pinhole, scratch and particles are well segmented. For the
faultless TFT-LCD surface image of Figure 8(d), the resulting binary image is uniformly white and no defect is claimed.

In order to demonstrate further the robustness of the proposed method under varying illumination, Figures 10(a1-e1) show five versions of a pinhole image (with some particles) under illuminations of 500, 1000, 1500, 2000 and 3000 Lux, respectively. Note that Figures 10(a1) and (e1) are two highly underexposed and overexposed images for inspection. Figures 10(a2-e2) depict the grey-level profiles of a scan line that crosses over the same region of the pinhole in Figures 10(a1-e1). The detection results as binary images are presented in Figures 10(a3-e3). They show that the defects of a pinhole and small particles are reliably detected for the test images under the illumination in the range 1000-2000 Lux. The pinhole and some particles can also be detected in the overly dark and bright images of Figures 10(a1) and (e1), except that their detected sizes are smaller and scattered.

In the proposed 1-D Fourier reconstruction scheme for TFT-LCD panel inspection, there are two parameters that affect the reconstruction results: the number of neighbouring points $r$ of each peak $u^*_j$ (see eq. (7)), and the high frequency region $w$ (see eq. (9)) in the Fourier spectrum. Empirical study has shown that $r = 3$ and $w = \frac{3}{8}N$ give the best reconstruction results for the inspection images at the resolution of 60 pixels/mm. In order to show that the proposed parameter values for $r$ and $w$ are not sensitive to varying line image length $N$, Figures 11-13 show additional test images of pinhole, particles, scratch, and a faultless surface in the lengths
of \( N = 256, 320 \) and \( 400 \) pixels. It is apparent from the experiments that the resulting binary images are uniform white for a faultless TFT-LCD panel image, and all defects are consistently detected with the same parameter setup, regardless of changes in the length \( N \). The parameters \( r \) and \( w \) used for the removal of periodic patterns in line images of TFT-LCD panel surfaces, however, may have to readjust for TFT-LCD panels at different image resolutions. This is because the structural pattern of a TFT-LCD panel at different resolutions, both coarser and finer, may yield distinct periodic grey-level profiles.

4. Conclusions

The present paper has presented a 1-D Fourier image reconstruction scheme for automatic visual inspection of micro-defects in patterned TFT-LCD panel surfaces. The sensed image for inspection is assumed to be obtained from a line scan system. The proposed method can directly work on line images and, therefore, has the merit in inspecting large-sized TFT-LCD panels.

The proposed method neither requires the design of quantitative features for various defects, nor needs a prestored template image for comparison. It is based on a global approach using 1-D Fourier reconstruction. By eliminating the frequency components that correspond to the periodically structural patterns of a TFT-LCD panel, and then back-transforming the remaining frequency components to the spatial domain image, the regularly-patterned background in 1-D grey level profiles can be effectively removed. A line image of a faultless region in the TFT-LCD panel surface will become an approximately uniform grey-level profile, while the line image of a defective
region will be distinctly preserved in the reconstruction. The Haar wavelet frame decomposition is further applied to remove the effect of uneven illumination. Finally, simple 3-sigma control limits are employed to segment the highly deviated elements (i.e. defects) from the uniform background in the decomposed detail part of the Harr wavelet frame.

At the resolution of 60 pixels/mm, the appearances of pinholes, scratches and particles can be visibly observed in the images. With the parameter values \( r = 3 \) (number of neighbouring points of each peak in the Fourier spectrum) and \( w = \frac{3}{8} N \) (the region of high frequency components) at the given resolution of 60 pixels/mm, the proposed method has reliably detected pinholes, scratches and particles. Experimental results also demonstrate that the proposed method is robust to detect the micro-defects under varying illumination conditions. The proposed method in this paper mainly aims at the detection of various micro-defects in TFT-LCD panel surfaces. It will be investigated in further research to classify the types of detected defects for in-process quality control.
References


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Figure 1.  (a) Schema of a single pixel of a TFT panel (see Kido 1993); (b) Surface image of a TFT-LCD panel at the resolution of 60 pixels/mm.
Figure 2. A cycle of six horizontal grey-level profiles for the TFT-LCD panel image in Figure 1 (b).
Figure 3. Six consecutive line images that cross over a defect in the TFT-LCD panel: (a1-f1) pinhole, (a2-f2) scratch and (a3-f3) particle.
Figure 4. The relationship between (a) $\Delta x$ in the spatial domain and (b) $\Delta u$ in the Fourier spectrum domain.
Figure 5. (a) A faultless line image of the TFT-LCD panel; (b) a defective line image of the pinhole; (c), (d) the respective Fourier power spectra of (a) and (b); (e), (f) the Fourier reconstructed images of (a) and (b), respectively, by setting the frequency components around the peaks $u's$ to zero; (g), (h) the final reconstructed images of (a) and (b) by further setting high frequency components to zero; (i), (j) the reconstructed images from the extended lengths of (a) and (b), respectively.
Figure 6. Wavelet frame decomposition for a faultless line image: (a) grey-level profile of a faultless line image \( f(x) \); (b) reconstructed image \( \hat{f}(x) \); (c) decomposed smooth part \( \hat{s}(x) \); (d) decomposed detail part \( \hat{d}(x) \).

Figure 7. Wavelet frame decomposition for a defective line image: (a) grey-level profile of a pinhole; (b) reconstructed image \( \hat{f}(x) \); (c) decomposed smooth part \( \hat{s}(x) \); (d) decomposed detail part \( \hat{d}(x) \).
Figure 8. Three defective images and their faultless version at the resolution of 60 pixels/mm: (a) pinhole, (b) scratch, (c) particles and (d) clear surface.

Figure 9. (a-d) Detection results as merged binary images for Figures 8(a-d), respectively.
Figure 10. Effect of changes in illumination: (a1-e1) A pinhole image under illuminations of 500, 1000, 1500, 2000 and 3000 LUX; (a2-e2) respective grey-level profiles of a scan image in (a1-e1); (a3-e3) detection results as merged binary images for the test images in (a1-e1), respectively.
Figure 11. Detection results from line images of length 256 pixels: (a1-d1) clear, pinhole, particles and scratch images for inspection; (a2-d2) detection results as merged binary images for (a1-d1), respectively.
Figure 12. Detection results from line images of length 320 pixels: (a1-d1) clear, pinhole, particles and scratch images for inspection; (a2-d2) detection results as merged binary images for (a1-d1), respectively.
Figure 13. Detection results from line images of length 400 pixels: (a1-d1) clear, pinhole, particles and scratch images for inspection; (a2-d2) detection results as merged binary images for (a1-d1), respectively.