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Technicolor at a Glance

Who We Are

Technicolor, a worldwide technology leader in the media and entertainment sector, is at the forefront of digital innovation.

Our world class research and innovation laboratories and our creative talent pool enable us to lead the market in delivering advanced services to content creators and distributors.

We also benefit from an extensive intellectual property portfolio focused on imaging and sound technologies, supporting our thriving licensing business.

Our Mission

Developing, creating and delivering immersive augmented digital life experiences that ignite our imagination.



Agenda

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Piracy of Entertainment Content

Piracy Modality Characterization

Luminance Flicker Analysis

Pirate Devices Identification

Questions and Answers





Piracy of Entertainment Content



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The Challenging Transition to Digital

Key specificities of digital content

- Clones rather than copies i.e. no more generational degradation
- Assets can be tangible or intangible
- Ease of dissemination i.e. the world is at your doorstep

Apparition of a bestiary of pirates (Courtesy: Irdeto)



On the cost of piracy... CNBC's Crime Inc #10: Hollywood Robbery (August 2012)



Threat Analysis





Anti-Piracy Arsenal

Regulate

- WIPO 1996 (DMCA, EUCD, Hadopi, etc.)
- SOPA/PIPA

Inform / Educate

- FA©T anti-piracy information campaigns
- Hard-to-counterfeit security features
 - Intaglio, color-shifting inks, holograms, CDIs

Prevent

- Content encryption aka. CAS and DRM
- Anti rip
- Playback/record control

Interfere / Jam

- Anti-recording e.g. Macrovision
- Anti-camcording

Monitor / Scout

- Data loss prevention systems
- Content fingerprinting

Trace

- Digital watermarking
- Passive forensics





The Forensics Landscape





Piracy Modality Characterization







Motivation

Different delivery channels \Rightarrow different piracy behaviors

- Camcorder theft in (digital) cinemas
- Ripping of optical disks
- HDMI stripping/screencasting of streamed content

Different delivery channels \Rightarrow different forensic watermarks

Cinema vs. disk vs. broadcast vs. OTT

Objective: isolate statistical discrepancies from pirated audio visual content to infer its piracy modality (and thus its delivery channel) to optimize watermark detection costs



Prior Works

Camcording

- Combing artifacts due to video interlacing [Lee:JIVP2012]
- Global motion jitter [Visentini:MMSP2013]
- Ghosting artifacts [Bestagini:ICIP2013]

Optical disk rip

- Double compression detection [...]
 - Holes in histograms of DCT modes
 - Occasional Unexpected large prediction error

Piracy classification [Technicolor]

- Legacy cinema vs. digital cinema [Rolland-Nevière:SPIE2012]
- DVD rip vs. digital cinema [Moreira-Pérez:WIFS2013, Chupeau:WIFS2014]



Luminance Flicker Feature

Interplay between the projector and the camcorder



Periodical high-frequency frame refresh (at 96 fps or higher) captured at 25 or 29.97 fps translates into low frequency aliased components

Feature vector

Reduce the video into a 1-D temporal signal (luminance row average)

Record the P highest peaks of the magnitude of the temporal FFT



Color Gamut Properties

Color feature vectors

- Color areas (saturation): number of colors at all luminance (Y) level, normalized by the maximum area of feasible colors at the same level
- Color centroids (bias): coordinates of the color centroids in C_bC_r plane at all luminance levels



Color feature behavior (DB mean)



Sequential Feature Selection



Number of time each component has been included into the optimal subset estimated by the SFS algorithm (repeated 100 times)

17 components in total (out of 768)

- Color saturation: 5 (out of 256)
- Color bias: 12 (out of 512)



Distribution of Edge Orientations

Observation: camcorder recapture induces planar perspective distortion

Distribution of straight edges impacted









Histogram of edge orientations (DB mean)

- 180-D histogram \rightarrow 4-D descriptor
 - Standard deviation and shape parameter of the Generalized Gaussian fit around the horizontal and vertical directions

$$f(o) = \frac{\beta}{2\alpha\Gamma(1/\beta)} \exp\left(\frac{|o-\mu|}{\alpha}\right)^{\beta}$$



Feature Sensitivity Synthetic Validation





Experimental Setup

Dataset of pirate samples

- 113 real-life camcorded DCI movies from TCH Forensics Operations
- 103 DVD rips: 48 from the web, 55 generated in the lab
- Random 15k frames long extract with no overlap \Rightarrow ~1.8k samples

SVM-based classifier

- Record AUC for 100 random train (90%) / test (10%) splits
- Score averaging to combine individual classifiers





Experimental Results



Take-away lessons

- 1. Good performance of individual classifiers
- 2. Boost of performances by combining features
- 3. Increased stability when combining features (reduced boxplot width)



Luminance Flicker Analysis



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Motivation

Early piracy has a drastic impact on revenues

- Poor quality camcord of new released movies in cinema
- Good quality camcord of premium VoD movies at home



Tell-tale visual artifact = small and periodic spatio-temporal variations of the luminance in the pirate sample e.g. dark/bright stripes rolling down/up the frames

<u>Objective #1:</u> model the flicker signal, estimate the parameters of the model and compensate for this effect to improve watermark detection

Objective #2: model the flicker signal to be able to synthesize distortion introduced by camcorder recapture for efficient benchmarking



Camcording an LCD Screen

LCD screen backlight

- Light source LCD luminance level
- Typical frequency: 120-1000 Hz

Camcorder shutter

- Global shutter (CCD) vs. rolling shutter (CMOS)
- Typical frequency: 25-50 fps







Aliasing due to the backlight signal being sampled below Nyquist frequency



Recapture Flicker Modeling [Baudry:WIFS2014]







Parameters Estimation [Baudry:WIFS2014]

Periodicity parameters

- Video frame → row luminance average (1-D vector)
- Temporal Fourier transform analysis for a single row \rightarrow spectral peak at ω_t
- Phase of Fourier coefficients at ω_t for all rows \rightarrow linear slope equal to ω_y



Amplitude parameters

 \blacksquare Near replicate in the spectrum at 0 and ω_t

0.4

DFT magnitude 0.3 0.2 0.2 0.1

'unwrapped" phase

0

200

Peak at ω_t

6

400

Slope = ω_v

8

600

line index

Frequency (Hz)

10

800

1000

12

Least means square minimization

$$(\alpha,\beta) = \arg\min\sum_{\varepsilon>-\rho}^{\varepsilon<\rho} \left(\mathbf{Y}(\omega_t+\varepsilon) - \frac{e^{i\psi}}{2} (\alpha \mathbf{Y}(\varepsilon) + \beta \delta(\varepsilon))\right)^2$$



Estimation Accuracy with Synthetic Flicker





Watermark Detection after Flicker Removal







Pirate Devices Identification



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Motivation

Forensic scenario

- Traitor-tracing analysis points to a suspect individual
- Camcorders/screens are seized at the home the suspect
- Assess if the pirate's flicker can be produced by a pair of these devices

Cross-referencing scenario

- Information about the piracy workflow is extracted from pirate samples
- Cross-reference such metadata to isolate piracy trends / hot spots

Objective: infer information about the intrinsic parameters of the pirate devices used to produce the pirate samples from the analysis of the induced underlying statistical anomaly



Flicker Parameters vs. Devices Parameters

$$\mathbf{f}(x, y, t) = (A \cdot \mathbf{c}(x, y, t) + B) \cdot \cos(\omega_t t + \omega_y y + \varphi)$$

- x column index
- y row index
- *t* frame index





Flicker Parameters vs. Devices Parameters

$$\mathbf{f}(x, y, t) = (A \cdot \mathbf{c}(x, y, t) + B) \cdot \cos(\omega_t t + \omega_y y + \varphi)$$

- *x* column index
- y row index
- *t* frame index



Can be seen as back-light frequency divided by sampling frequency ${}^{H}/{}_{T_{ro}}$ (in lines/sec)



Flicker-Based Pirate Device Identification

Piracy identity



Forensic protocol

- 1. Estimate ω_y from the pirate video incl. variant for corner cases
- 2. Extract ground truth $f_{\rm BL}$ and $T_{\rm ro}$ from suspect's devices
- 3. Assess if a pair of devices matches the piracy identity



Ground Truth Values Measurements

Backlight frequency of suspect screens

- Record the backlight signal with a photodetector (for a still gray image)
- Peak of signal's FFT $\Rightarrow f_{\rm BL}$



Read-out time of suspect camcorders

- Record a short video of a still gray image on a screen with known f_{BL}
- Estimate ω_y (cf. before)

$$\bullet T_{\rm ro} = \frac{\omega_y H}{2\pi . f_{\rm BL}}$$





Blind Identification Results [HajjAhmad:IHMMSec2015]

Screens	Camcorders	JVC 50 fps	Panasonic 50 fps	Sony 25 fps	Toshiba 29.97 fps
	$T_{\rm ro}$ (ms) $f_{\rm BL}$ (Hz)	13.5	16	15	32.65
Screen 1	240.06	J—1 🖌	P-1 🗸	S−1 🖌	T−1 🖌
Screen 2	180.43	S—3 🗙	P-2 🖌	S−2 🖌	T—2 ✓
Screen 3	159.98	J−3 🗸	P-3 🗸	P—5 🗙	T−3 🖌
Screen 4	120.00	J-4 🗸	P-1 🗙	J—4 🗙	T-4 🖌
Screen 5	146.61	J—5 🖌	S—5 🗙	S—5 🖌	T—5 🖌
Screen 6	226.70	J−6 🗸	P-6 🗸	S-6 🗸	T—6 🖌
Screen 7	172.80	P-5 🗙	P-7 🗸	S-7 🗸	T-7 🖌

X Screen-camcorder pairs are reduced to the product $\Pi = T_{ro} \times f_{BL}$ resulting in more errors, e.g. pairs J-2, J-7, S-3, and P-5 all have $\Pi \approx 2480 \pm 50$





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Conclusions

Passive forensics beyond content authentication

- Infringement analysis, quality control, context adaptation
- Complement the anti-piracy arsenal
 - Piracy modality identification for watermark detection budget optimization
 - Piracy path modeling for compensation and/or simulation
 - Pirate device characterization for attribution and/or cross-referencing

Flicker forensics

- Interplay between the backlight of the screen and the shutter of the camcorder
- Low-power periodic spatio-temporal signal
- Efficient estimation techniques: detection, cancellation, intrinsic parameters inference

Future work

- Flicker shape estimation
- Screencaster piracy



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Questions





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