

Faculty of Computer Science Institute of Systems Architecture, Privacy and Data Security Research Group

On Resampling Detection in Re-Compressed Images

WIFS 2009

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Image Forensics

- use image statistics for identification of source device or detection of manipulations
- variety of different forensic methods can be found in the literature
- existing schemes work well under laboratory conditions

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On Resampling Detection in Re-Compressed Images

Resampling Detection

- image manipulations often rely on geometric transformations (scaling, rotation, ...) of images or parts thereof
- resampling to a new image grid; involves an interpolation step
- ▶ interpolation introduces periodic linear correlations between neighboring pixels



 resampling artifacts can be detected by the analysis of linear predictor residue [Popescu & Farid, 2005], [Kirchner, 2008]







large absolute prediction errors indicate minor degree of linear dependence



- ► predictor residue: $e(\omega t') = s(\omega t') \sum_{k=-K}^{K} \alpha_k s(\omega t' + \omega k)$ $(\alpha_0 := 0)$
- ► large absolute prediction errors indicate minor degree of linear dependence
- optimal weights in the example: $\alpha = (0.5, 0, 0.5)$ • every second sample defaults to 0



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- ► predictor residue: $e(\omega t') = s(\omega t') \sum_{k=-K}^{N} \alpha_k s(\omega t' + \omega k)$ $(\alpha_0 := 0)$
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- optimal weights α can be determined with an EM algorithm [Popescu & Farid, 2005] or set to a fixed linear filter mask [Kirchner, 2008]
- **• p-map:** $p(\omega t') \propto \exp(-\sigma |e(\omega t')|^{\tau})$
- measure for the strength of linear dependence

Typical Detection Results



each spectrum graph has been individually normalized and processed with a maximum filter

- resampling causes periodic pattern in the p-map and distinct peaks in the p-map's DFT
- peak position is characteristic for resampling parameters

 $\mathbf{f}_{\text{re}} = \left| \left(\mathbf{A}' \right)^{-1} \begin{pmatrix} k \\ l \end{pmatrix} - \left[\left(\mathbf{A}' \right)^{-1} \begin{pmatrix} k \\ l \end{pmatrix} \right] \right|$

 typical resampling detectors employ a frequency domain peak detector



resampling detection

On Resampling Detection in Re-Compressed Images









current literature only considered post-compression

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- ► JPEG introduces 8 × 8 pixel block structure
 - increased prediction error at block boundaries leads to new periodic artifacts
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 current literature concludes that resampling detection under moderate JPEG post-compression (QF_{post} < 95) is practically **impossible** [Popescu & Farid, 2005], [Mahdian & Saic, 2008]

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► shifted JPEG peaks:
$$\mathbf{f}_{\text{pre}} = \left| (\mathbf{A}')^{-1} \begin{pmatrix} k/8 \\ l/8 \end{pmatrix} - \left[(\mathbf{A}')^{-1} \begin{pmatrix} k/8 \\ l/8 \end{pmatrix} \right] \right|$$



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- depending on the pre-compression quality, shifted JPEG peaks can be more robust to post-compression than resampling peaks
 - ▷ resampling detection in re-compressed images **benefits** from additional peaks!

periodic resampling artifacts can be best measured in the frequency domain

General Procedure

 pre-processing to reduce disturbing lowfrequency components

decision based on the existence of distinct peaks

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decision based on the existence of distinct peaks
 correlation Popescu & Farid, 2005) and cumulative periodogram [Kirchner, 2008] based approaches not optimal
 exhaustive search best for one distinct peak

• input: DFT of the p-map, P = DFT(p-map)





On Resampling Detection in Re-Compressed Images

- ▶ input: DFT of the p-map, P = DFT(p-map)
- normalization

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peak finding

$$P_m(\mathbf{f}) = \begin{cases} P_n(\mathbf{f}) & \text{if } P_n(\mathbf{f}) = \max P_n(\mathbf{f} + \mathbf{w}) \\ & \mathbf{w} \in \{-W, \dots, W\}^2 \\ 0 & \text{else.} \end{cases}$$



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remove JPEG post-compression peaks

remove frequency components $(k/8, l/8), (k, l) \in \{-4, \dots, 4\}^2$





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gamma correction

$$P_{\gamma} = \max(P_m) \cdot \left(\frac{P_m}{\max(P_m)}\right)^{\gamma}$$





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decision criterion:

 $\delta = \max_{r > r_t} \Sigma_r / \operatorname{median}_{r > r_t} \Sigma_r$

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sum of the 4 maximum peaks on each radius r

Experimental Setup

- test database consists of 200 never-compressed images (1024 × 1024) from the 'Dresden Image Database' [Gloe & Böhme, 2010]
- bilinear resizing, JPEG pre- and post-compression using ImageMagick's convert

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 $\mathsf{QF}\,=\,\{40,\,50,\,70,\,75,\,80,\,90,\,95,\,98,\,100\}^2$

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p-map is calculated from the luminance channel for two fixed linear filter masks [Kirchner, 2008]:

$$\boldsymbol{\alpha}\mathbf{8} = \begin{bmatrix} -\frac{1}{4} & \frac{1}{2} & -\frac{1}{4} \\ \frac{1}{2} & 0 & \frac{1}{2} \\ -\frac{1}{4} & \frac{1}{2} & -\frac{1}{4} \end{bmatrix}, \quad \boldsymbol{\alpha}\mathbf{4} = \begin{bmatrix} 0 & \frac{1}{4} & 0 \\ \frac{1}{4} & 0 & \frac{1}{4} \\ 0 & \frac{1}{4} & 0 \end{bmatrix}$$

altogether more than 400 000 detector runs (size of analyzed region fixed to 512 × 512)

filter masks: $\alpha_4 \quad \alpha_8$ (FAR = 1 %)



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- downsampling not detectable (as per upsampling for QF_{post} < 90)

Re-Compression Detection Results



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 pre-compression should be lower than (at least equal to) post-compression quality

Concluding Remarks

- resampling detection in re-compressed images is of high practical relevance
 images are often already JPEGs before being further processed
- reliable detection after JPEG compression was believed to be a lost cause
- **but:** (p)re-compression introduces additional artifacts in terms of **shifted JPEG peaks** in the p-map's spectrum
 - results from 400 000 detector runs show notable performance gain for upsampling compared to the post-compression scenario

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 - downsampling remains to be the most problematic geometric transformation
 - detector is blind to transformations that interfere with post-compression peaks
 - sufficient size of the analyzed image region is critical to distinguish resampling and post-compression peaks

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 - sufficient size of the analyzed image region is critical to distinguish resampling and post-compression peaks
 - concept of shifted peaks also applies to CFA interpolation peaks



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Thanks for your attention

Questions?

<u>Matthias Kirchner</u>, Thomas Gloe Technische Universität Dresden

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Pre- and Post-Compression Quality

• detection rates for 150 % upscaling (FAR = 1%)

	40	50	60	70	75	80	90	95	98	100
40	70	83.5	94.5	99.5	99.5	99	100	100	100	99.5
50	8.5	61	82.5	98	99	99.5	99.5	99	99.5	99.5
60	2	11.5	63	96	100	100	100	100	100	100
70	1.5	2	7	86	98	99.5	100	100	100	100
75	4	3	2.5	55	94	95.5	100	100	100	100
80	3	1.5	1.5	24	71	91.5	100	100	100	100
90	2.5	2.5	1.5	5.5	10	25.5	99.5	100	100	100
95	3	2.5	2	7	11	11.5	87.5	100	100	100
98	2.5	3	2.5	5.5	9	10	56	99.5	100	100
100	2	3	3	5.5	8.5	10	58.0	99	100	100
*	2.5	2	2.5	5	9	10.5	52.5	98.5	100	100

post-compression quality

pre-compression quality

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