

Malware Detection in JPEG

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Abstract: *Cyberattacks on people, companies, and organizations have grown in frequency. Cybercriminals are constantly searching for efficient ways to infect targets with malware in order to initiate an attack. Millions of people use images every day all throughout the world, and the majority of users think pictures to be secure for usage, however some kinds of pictures have the potential to carry a malware payload and execute detrimental acts. The main reason JPEG is the most widely used image format is because of its lossy compression. It's applied almost everyone, from small businesses to major corporations, and is present on nearly all devices (on digital cameras, cellphones, social networking, websites, etc.). Due of their reputation for being innocuous, enormous JPEG images have a lot of potential for misuse.*

Keywords: JPEG, Automatic Interpretation, Image Processing, Artificial Intelligence, Malware Detection, CNN, Deep learning

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