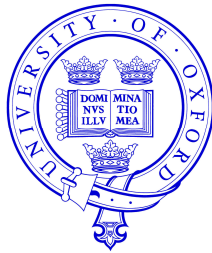


# The Square Root Law of Steganographic Capacity



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# Background

*The more information you hide, the greater your risk of discovery.*

The steganographic capacity question is:

*Given a particular limit on risk, how much can you hide?*

A key part of this question, though rarely asked, is:

*How does the secure capacity depend on the size of the cover?*

Ross Anderson in the 1<sup>st</sup> Information Hiding Workshop (1996):

*“...the more covertext we give the warden, the better he will be able to estimate its statistics, and so the smaller the rate at which Alice will be able to tweak bits safely. The rate might even tend to zero...”*

# The Square Root Law of Steganographic Capacity

## Outline

- *Background*
- *Steganographic capacity theorems*
- *Experimental design*
- *Results*
- *Conclusions*

# Capacity theorems

## Randomly modulated codes (Wang & Moulin, 2008)

Hides information which is perfectly undetectable, at a **linear** rate, but requires the embedder to have complete knowledge of their cover source.

## Batch steganographic capacity theorem (Ker, 2007)

Applies to multiple independent covers. Under certain assumptions, total capacity is asymptotically proportional to **square root** of number of covers.

## Batch steganographic strategies (Ker, 2008)

Under certain conditions, KL divergence is locally quadratic; implies that total capacity is asymptotically proportional to **square root** of number of covers.

## Square root law for Markov chains (Filler, Ker, & Fridrich, 2009)

When cover can be modelled as a Markov chain, and embedding as independent state change, if embedding is not perfect then the max change rate is asymptotically inversely proportional to square root of cover size.

# Aim

*Investigate empirically the relationship between cover size, payload size, and detectability of payload.*

Related work can be found in:

- Ker, IHW 2004
- Böhme, IHW 2005
- Böhme & Ker, SPIE EI 2006

The idea is to fix the:

- parent cover set,
- embedding method,
- detector,
- detectability metric,

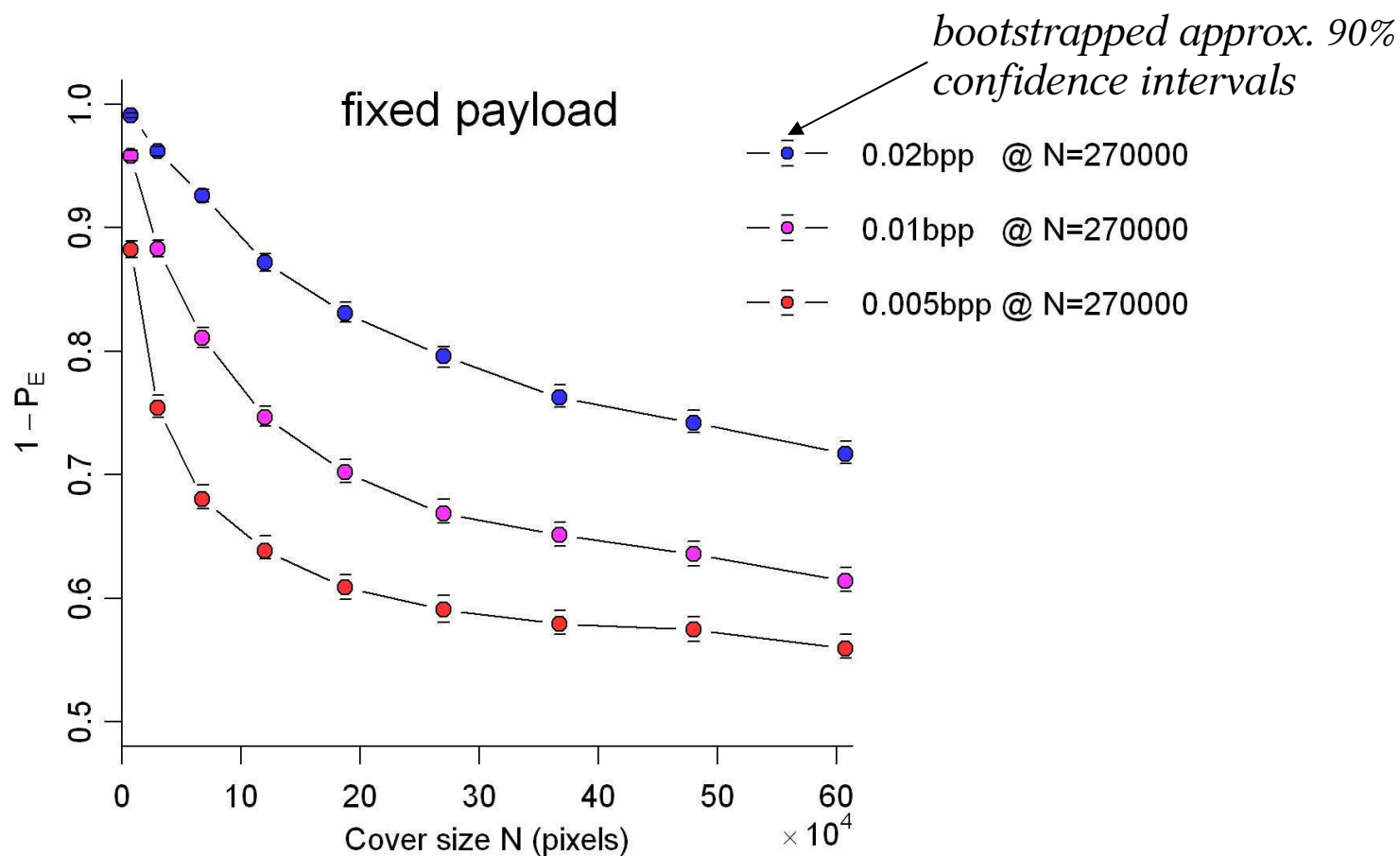
then produce cover images of different sizes from the parent set, and investigate detectability.

*But it is difficult to make image sets of different sizes which do not also have different characteristics: noise, density, etc.*

The best we can do is **crop down** large images to smaller ones, choosing the crop region to preserve local variance as much as possible.

# Results

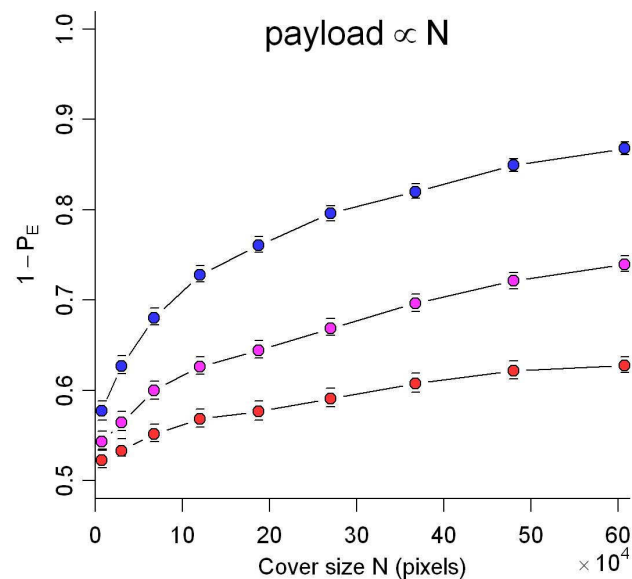
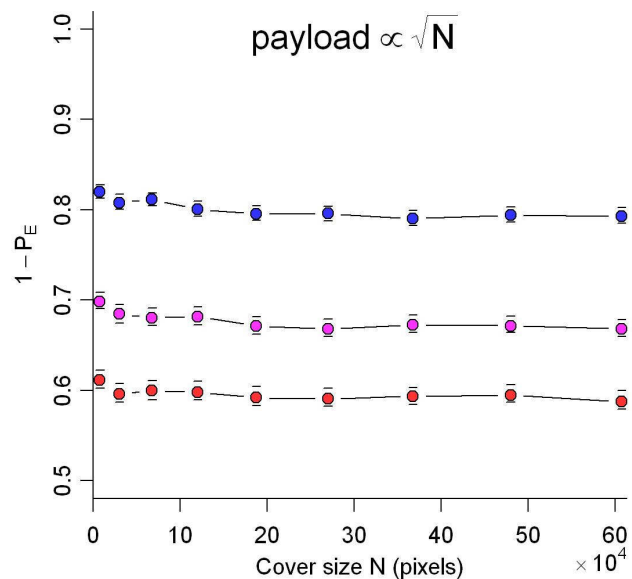
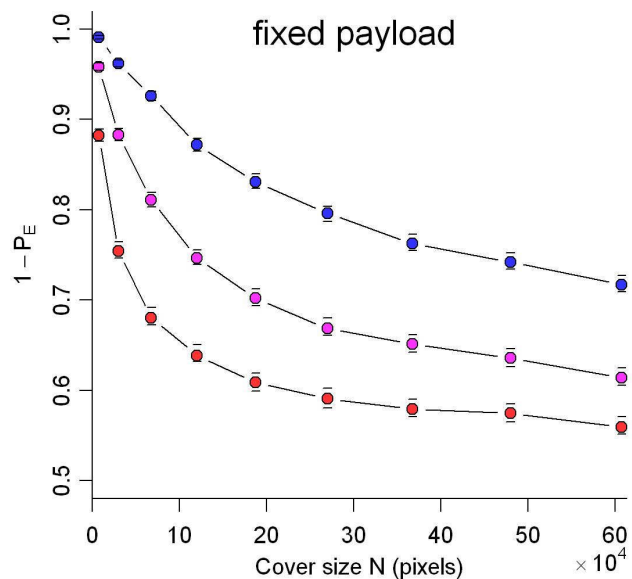
- Covers: *never-compressed grayscale scanned images (1024×732)*
- Embedding: *LSB replacement*
- Detector: *improved WS*
- Metric: *maximum probability of correct classification*



# Results

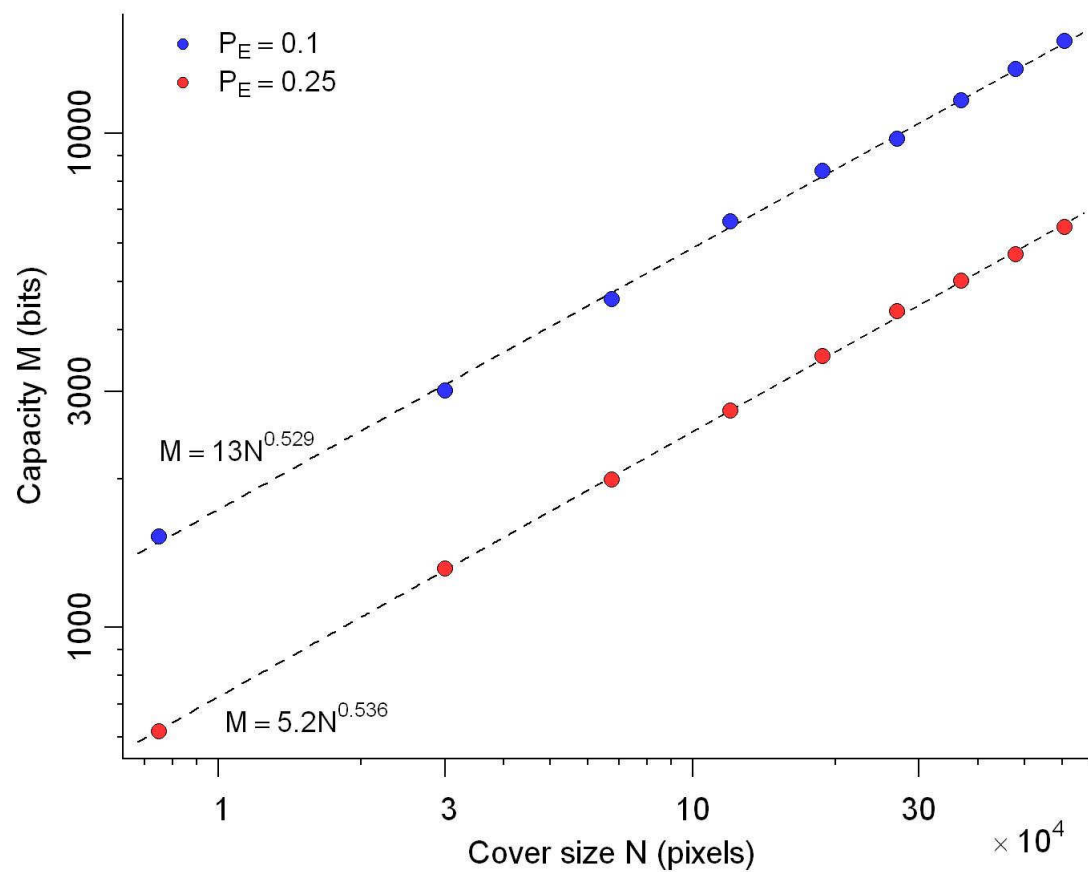
- Covers: *never-compressed grayscale scanned images (1024×732)*
- Embedding: *LSB replacement*
- Detector: *improved WS*
- Metric: *maximum probability of correct classification*

—●— 0.02bpp @ N=270000  
—●— 0.01bpp @ N=270000  
—●— 0.005bpp @ N=270000



# Results

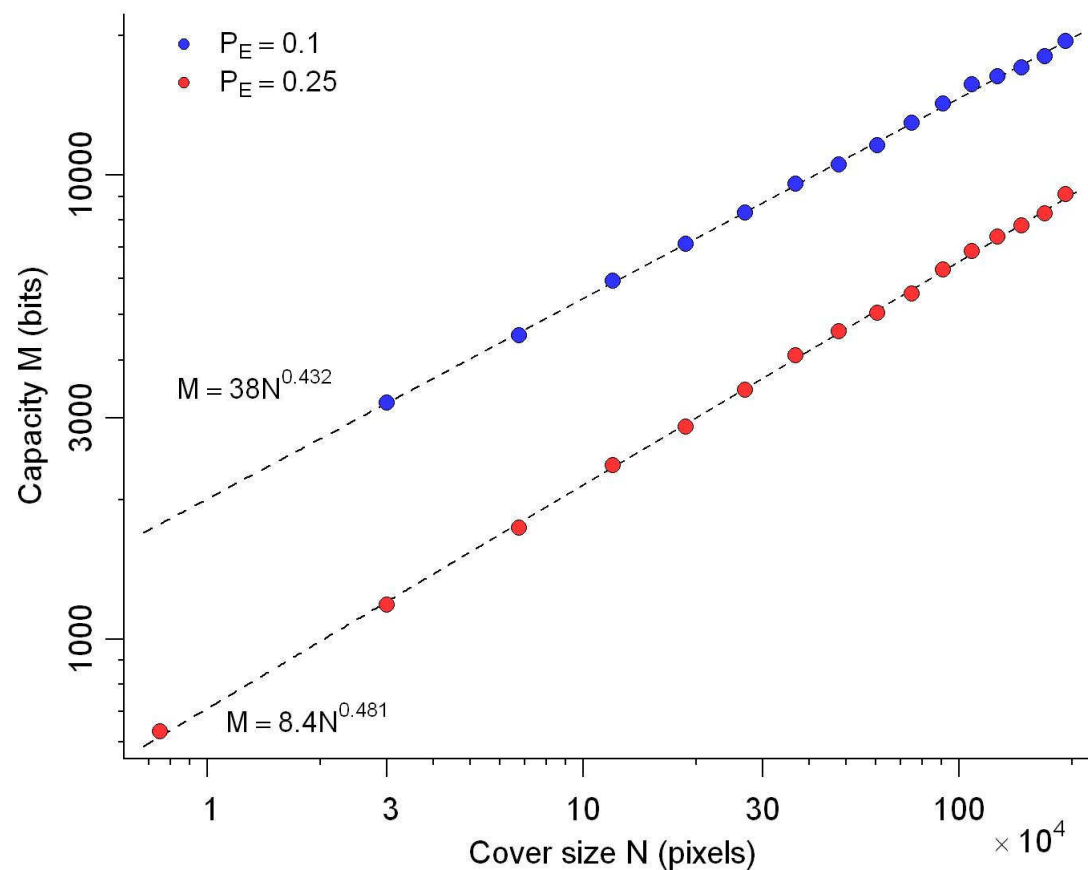
- Covers: *never-compressed grayscale scanned images (1024×732)*
- Embedding: *LSB replacement*
- Detector: *improved WS*
- Metric: *maximum probability of correct classification*





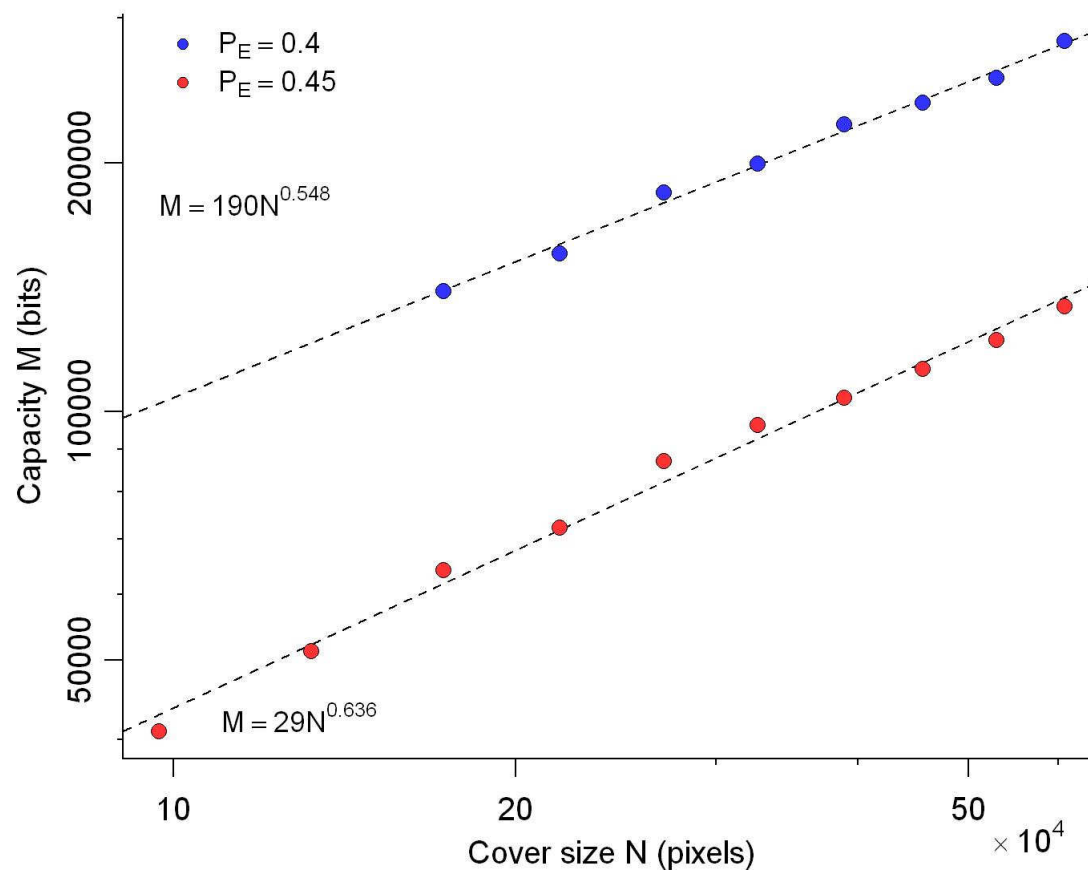
# Results

- Covers: *digital camera images, once JPEG compressed (2000×1500)*
- Embedding: *LSB replacement*
- Detector: *Triples*
- Metric: *maximum probability of correct classification*



# Results

- Covers: *never-compressed grayscale scanned images (1024×732)*
- Embedding: *LSB matching ( $\pm 1$ )*
- Detector: *adjacency HCF COM*
- Metric: *maximum probability of correct classification*



# JPEG embedding

Two additional challenges:

## *1. How to measure size?*

We count nonzero quantized DCT coefficients.

## *2. How to crop images?*

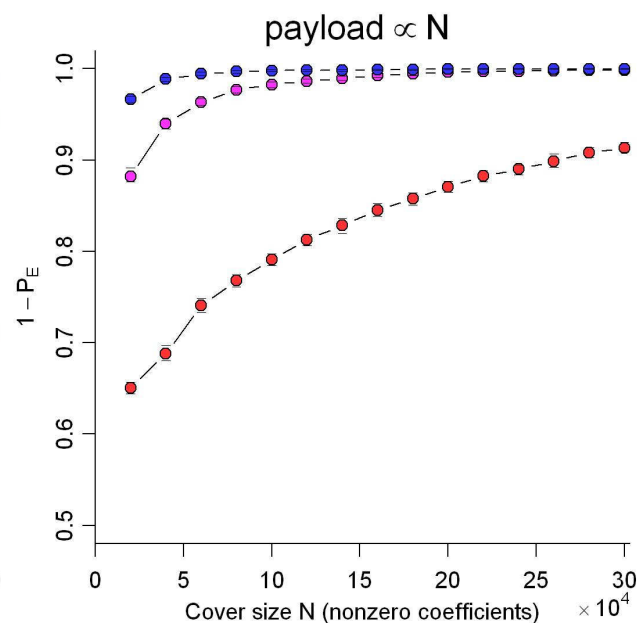
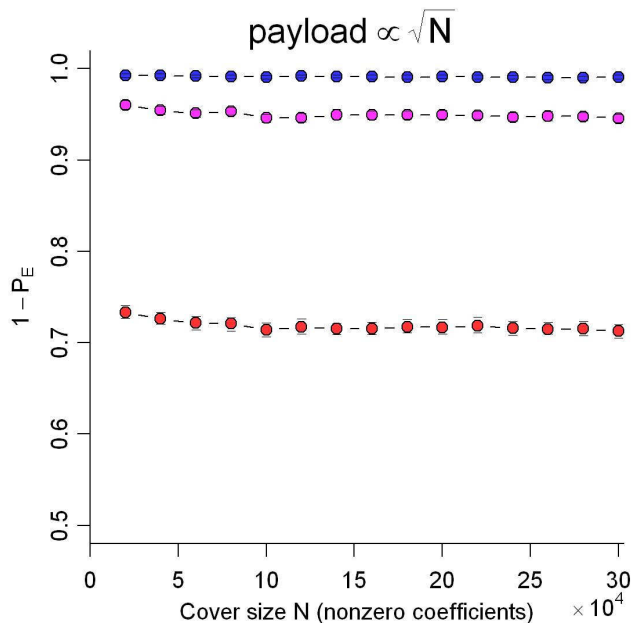
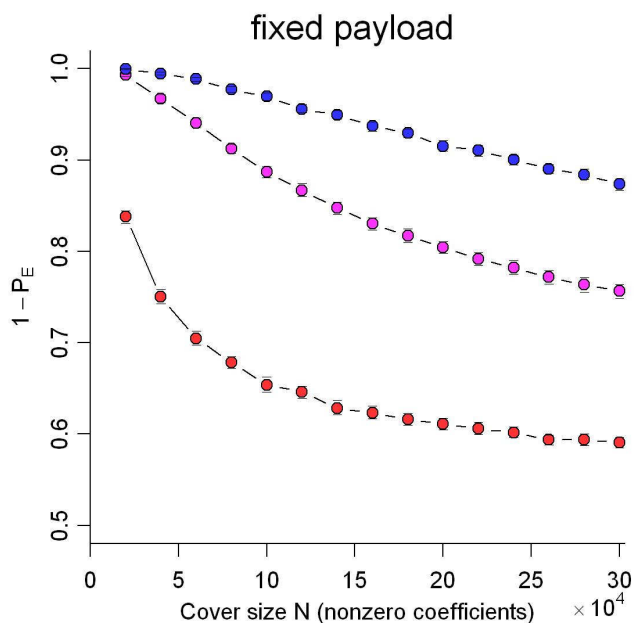
We crop to an  $8 \times 8$  grid, choosing region:

- to select desired number of nonzero coefficients,
- while preserving the ratio of nonzero coefficients to all coefficients as much as possible.

# Results

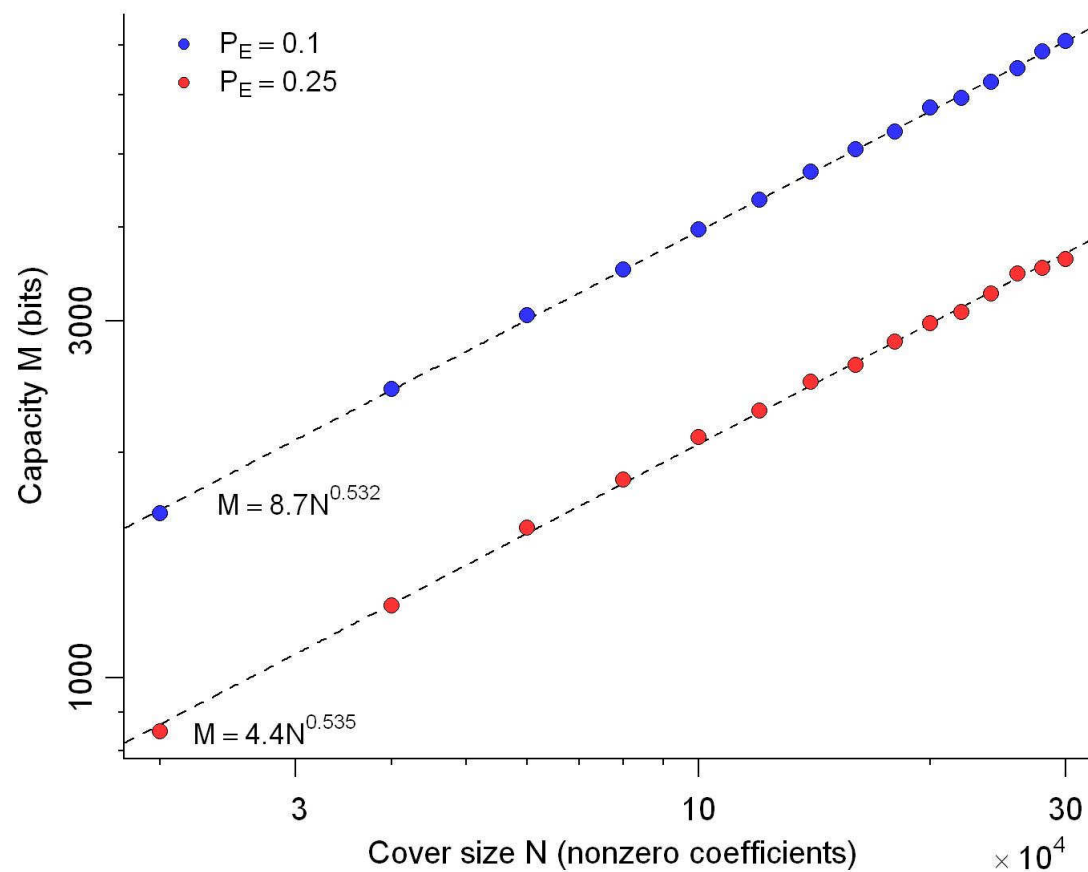
- Covers: *mixed camera JPEGs*
- Embedding: *nsF5, no matrix embedding*
- Detector: *274-feature SVM*
- Metric: *maximum probability of correct classification*

- 0.125bpnc @ N=50000
- 0.075bpnc @ N=50000
- 0.025bpnc @ N=50000



# Results

- Covers: *mixed camera JPEGs*
- Embedding: *nsF5, no matrix embedding*
- Detector: *274-feature SVM*
- Metric: *maximum probability of correct classification*



# Conclusions

- Exploring a range of cover image types, embedding methods, detectors, and detectability metrics, we observed close accordance with a square root law.

*Of course, this is no proof of a square root law in general. The full law will be a suite of theorems proving that it holds under a variety of conditions.*

- Two important limitations of the law:
  - It does not apply when **perfect steganography** is available.
  - It properly applies to the **number of embedding changes**, not payload.

*It turns out that payload capacity is of order  $\sqrt{N} \log N$ , given best adaptive source coding.*

- Has many consequences for the practice of steganography and steganalysis:
  - Must take care when designing steganographic file systems.
  - “Bits per pixel”, “bits per nonzero coefficient” are the wrong measures...