

# SURE-based Optimization for Adaptive Sampling and Reconstruction

## Supplementary Materials



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# PART I

## Equal-Time Comparison

### Compared Methods:

- Monte Carlo
- Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011]
- Random Parameter Filtering [Sen and Darabi, ACM TOG 2012]
- SURE-based Optimization (our approach, using cross bilateral filters)

# SPONZA

## Global Illumination (Path Tracing) Motion Blur



1600 x 1200

**SPONZA**

**Equal-time Monte Carlo**, 68 spp, 890.5 sec.



# SPONZA

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 63.84 spp, 906.2 sec.



**SPONZA**

**Random Parameter Filtering [Sen and Darabi, ACMTOG 2012], 16 spp, 1676.1 sec.**



**SPONZA**

**SURE-based Optimization (Our Approach)**, 63.24 spp, 896.0 sec.



**SPONZA**

**Reference**, 8192 spp



# TOWN

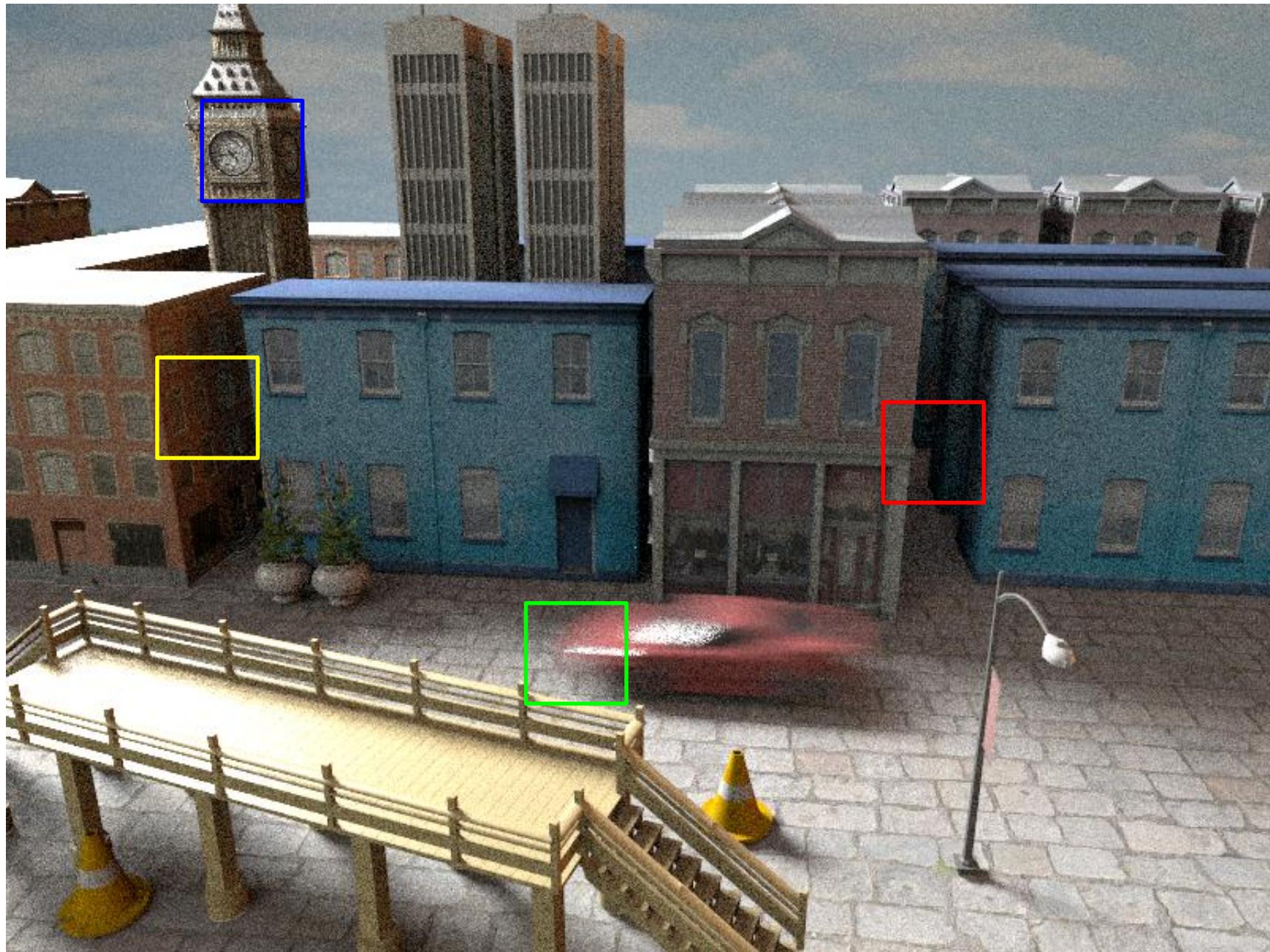
Environment Lighting  
Area Lighting  
Motion Blur



800 x 600

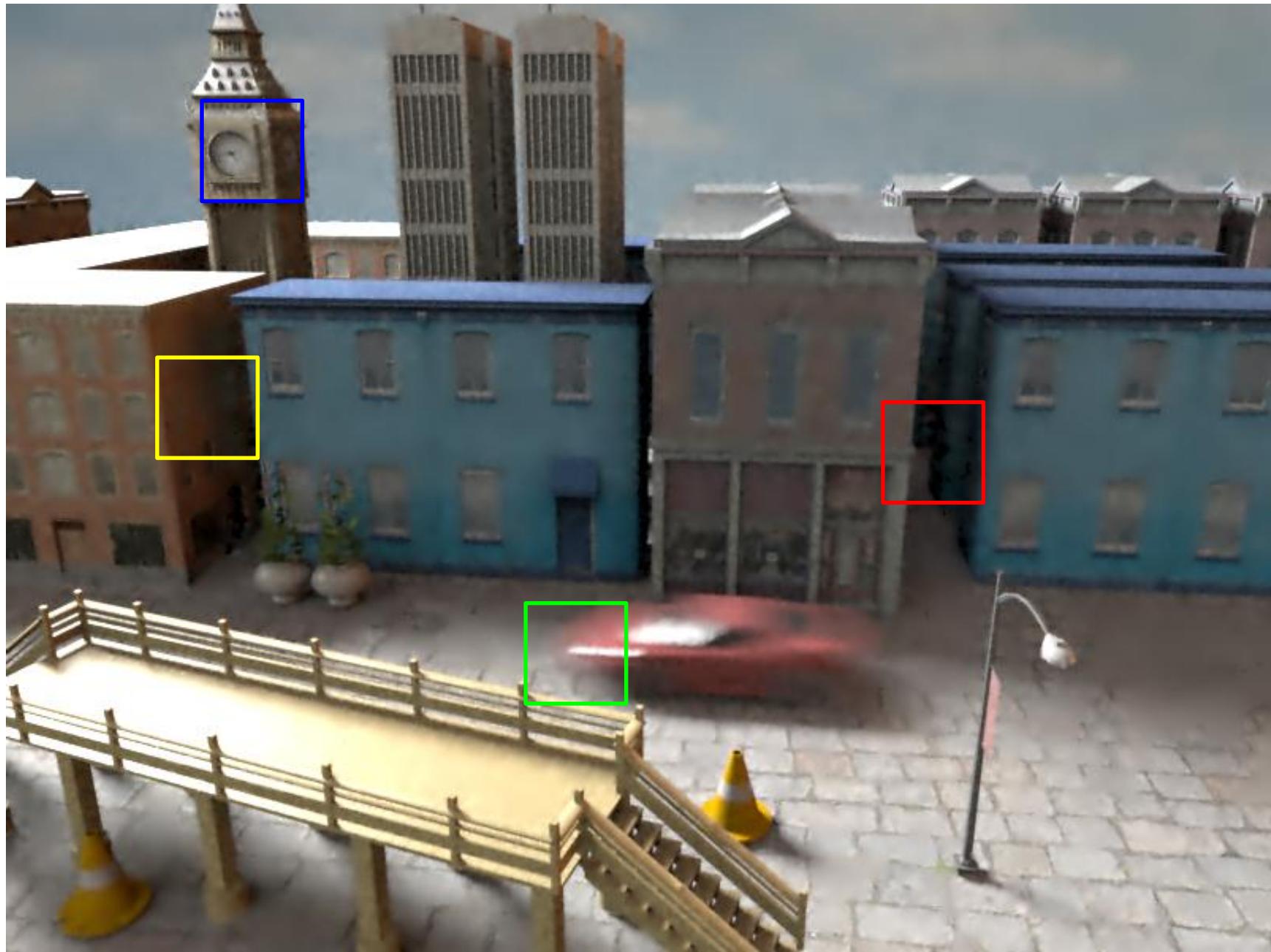
**TOWN**

**Equal-time Monte Carlo, 82 spp, 59.9 sec.**



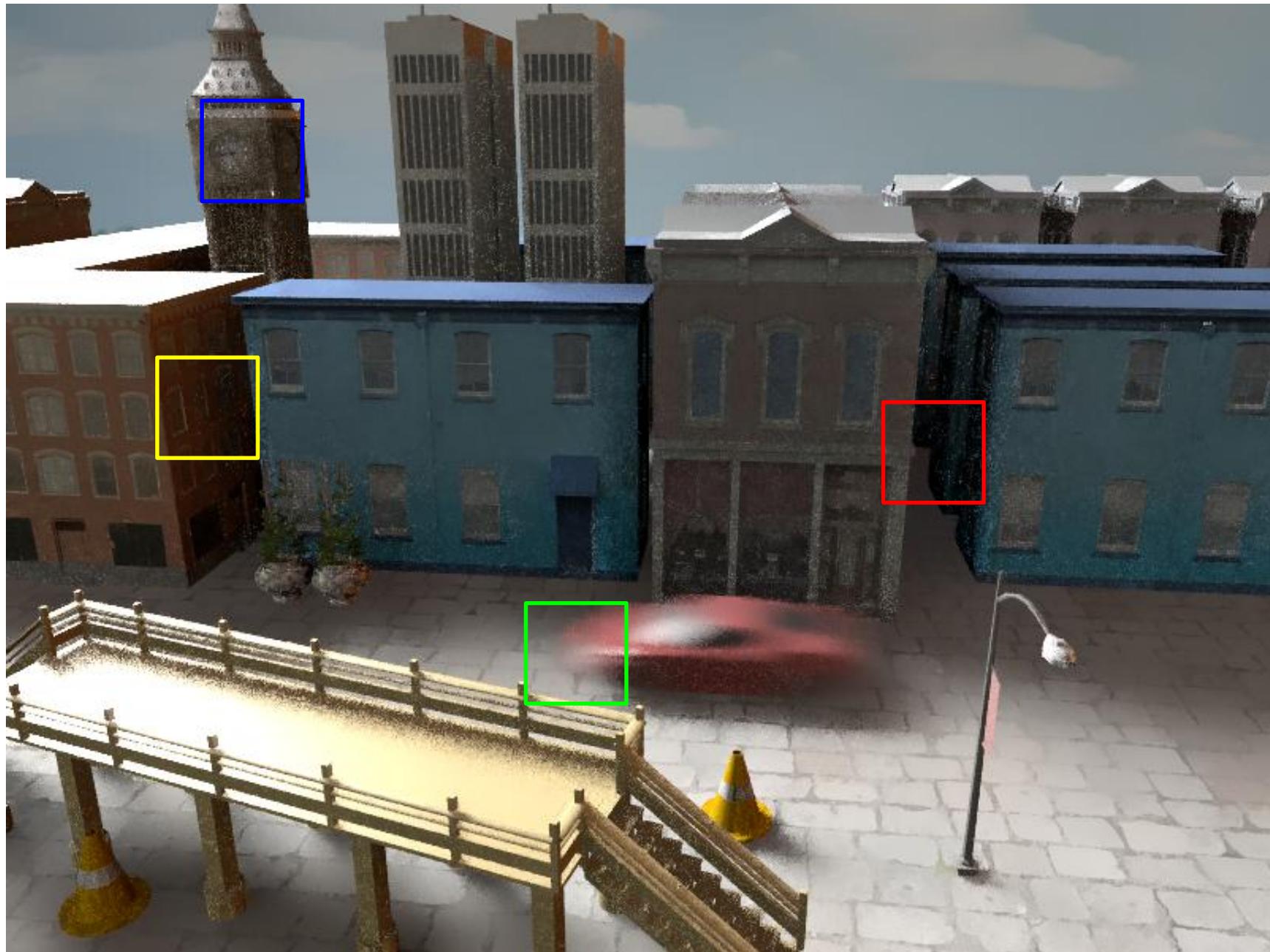
**TOWN**

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 51.82 spp, 61.8 sec.



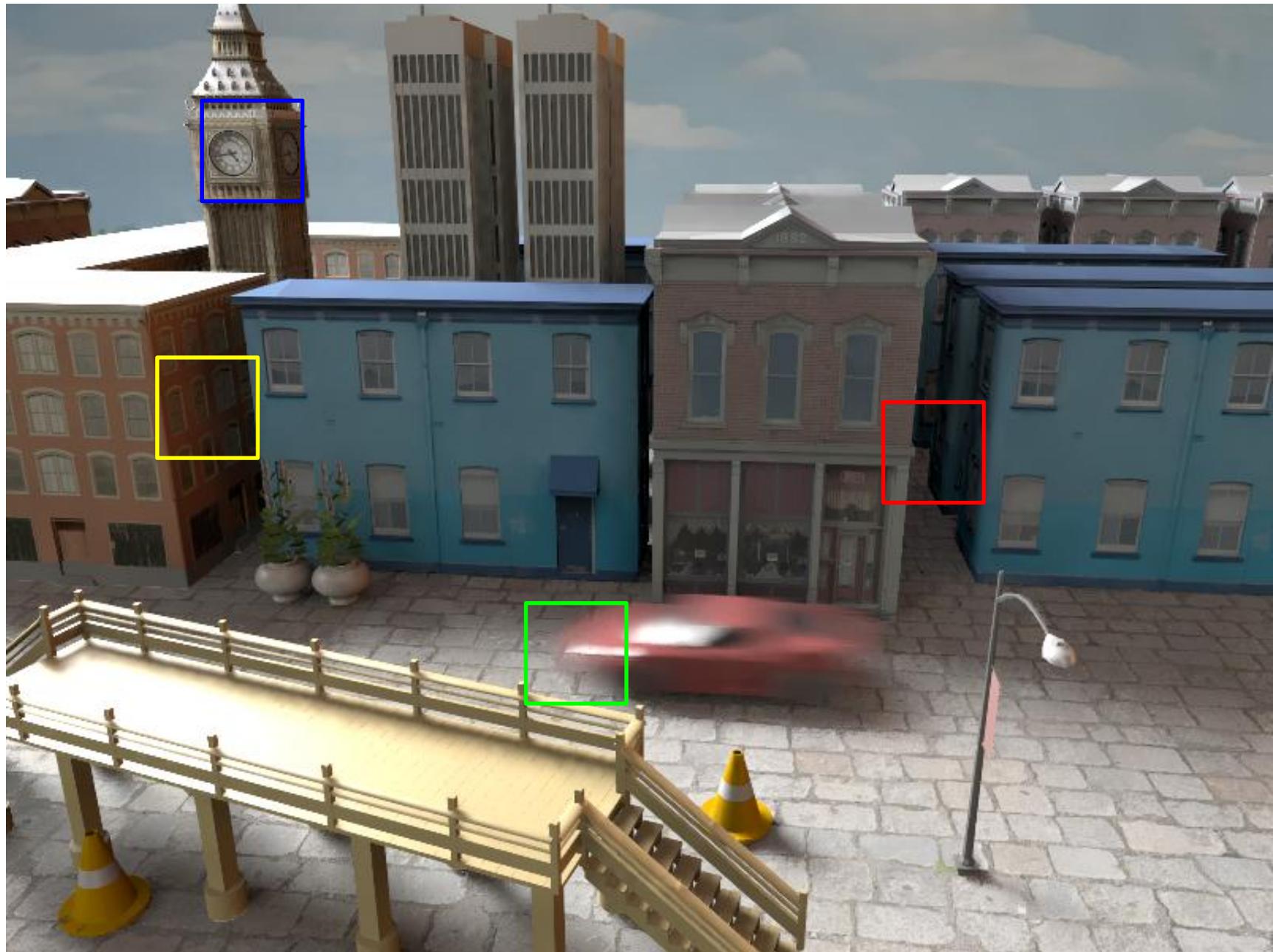
TOWN

Random Parameter Filtering [Sen and Darabi, ACMTOG 2012], 8 spp, 272.4 sec.



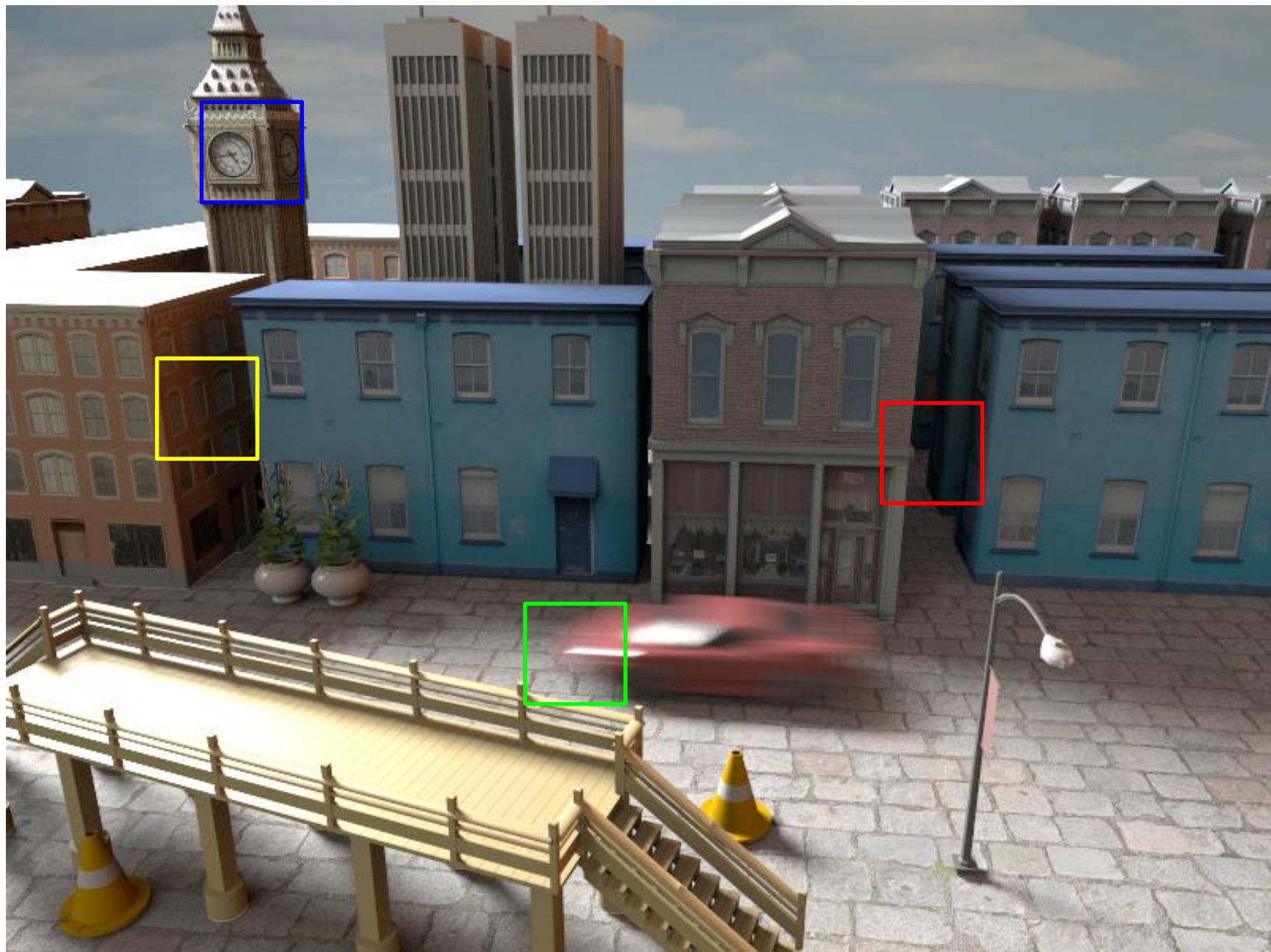
**TOWN**

**SURE-based Optimization (Our Approach), 39.79 spp, 59.6 sec.**



**TOWN**

**Reference**, 4096 spp



# SIBENIK

Global Illumination (One-Bounce Path Tracing)  
Depth of Field



1024 x 1024

# SIBENIK

Equal-time Monte Carlo, 44 spp, 140.0 sec.



# SIBENIK

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 39.86 spp, 135.0 sec.



# SIBENIK

Random Parameter Filtering [Sen and Darabi, ACMTOG 2012], 8 spp, 363.0 sec.



# SIBENIK

**SURE-based Optimization (Our Approach), 26.69 spp, 140 sec.**



**SIBENIK**

**Reference**, 4096 spp



# TEAPOT

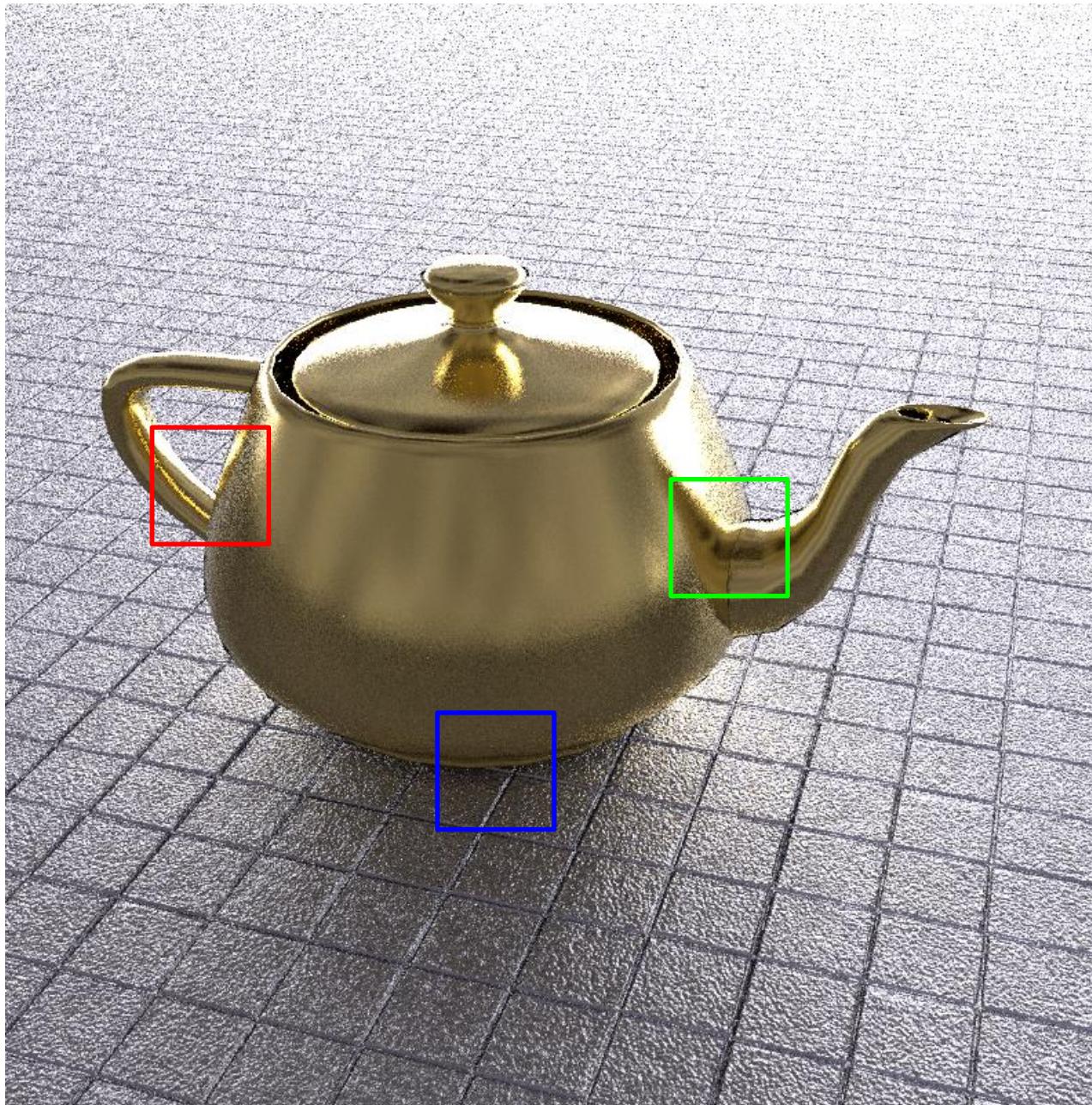
Environment Lighting  
Glossy Reflection



800 x 800

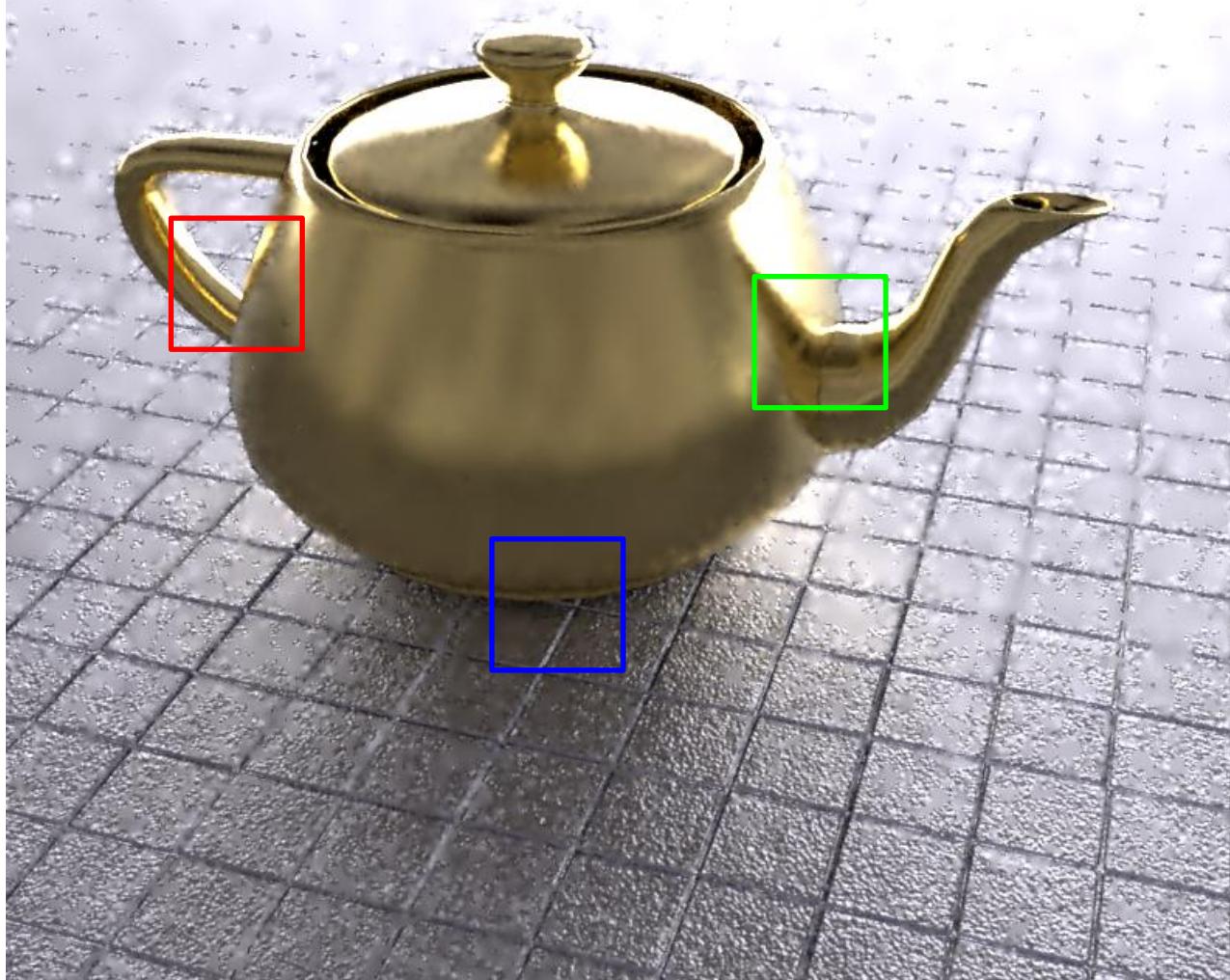
# TEAPOT

Equal-time Monte Carlo, 35 spp, 42.0 sec.



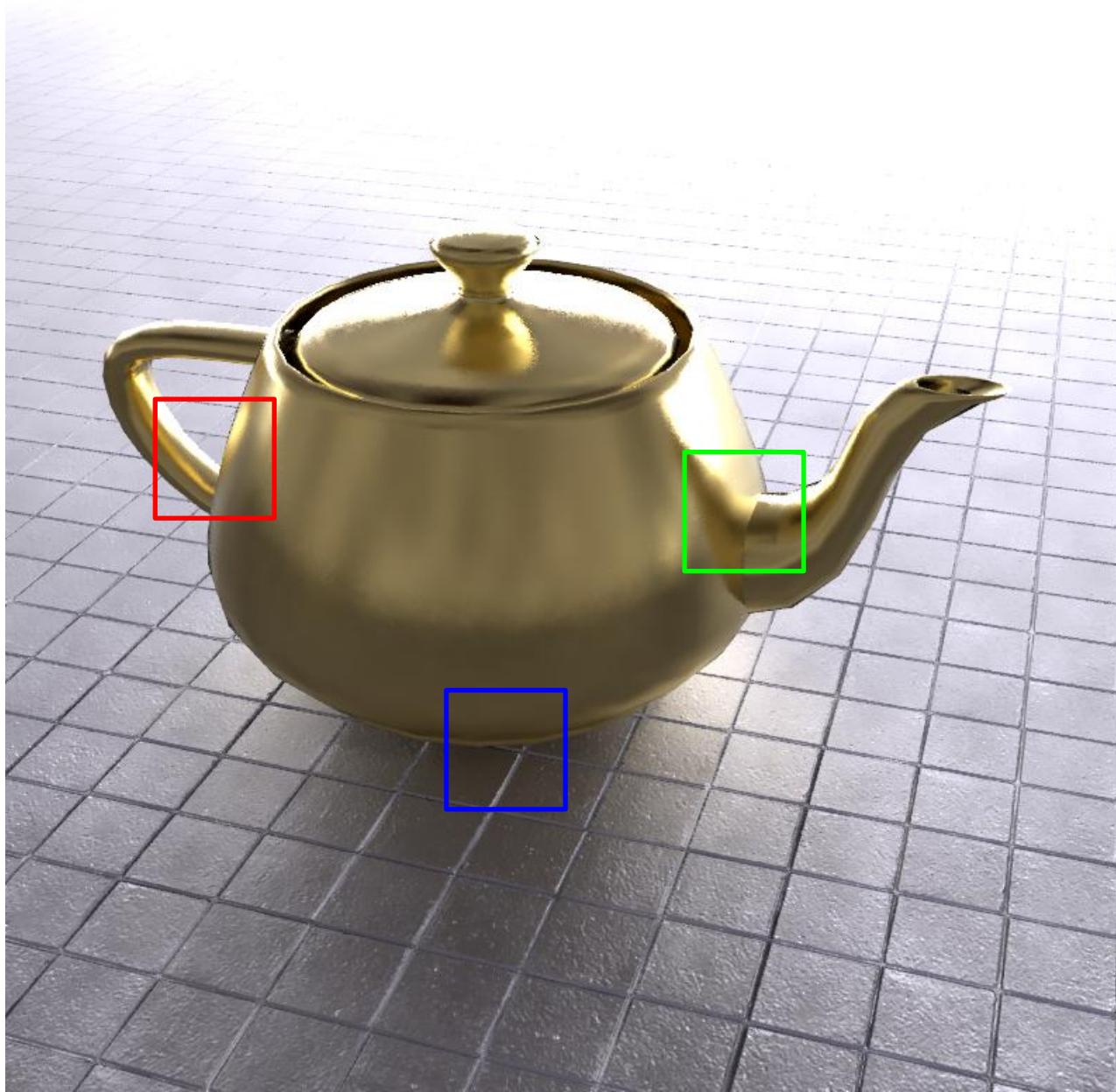
# TEAPOT

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 23.96 spp, 44.3 sec.



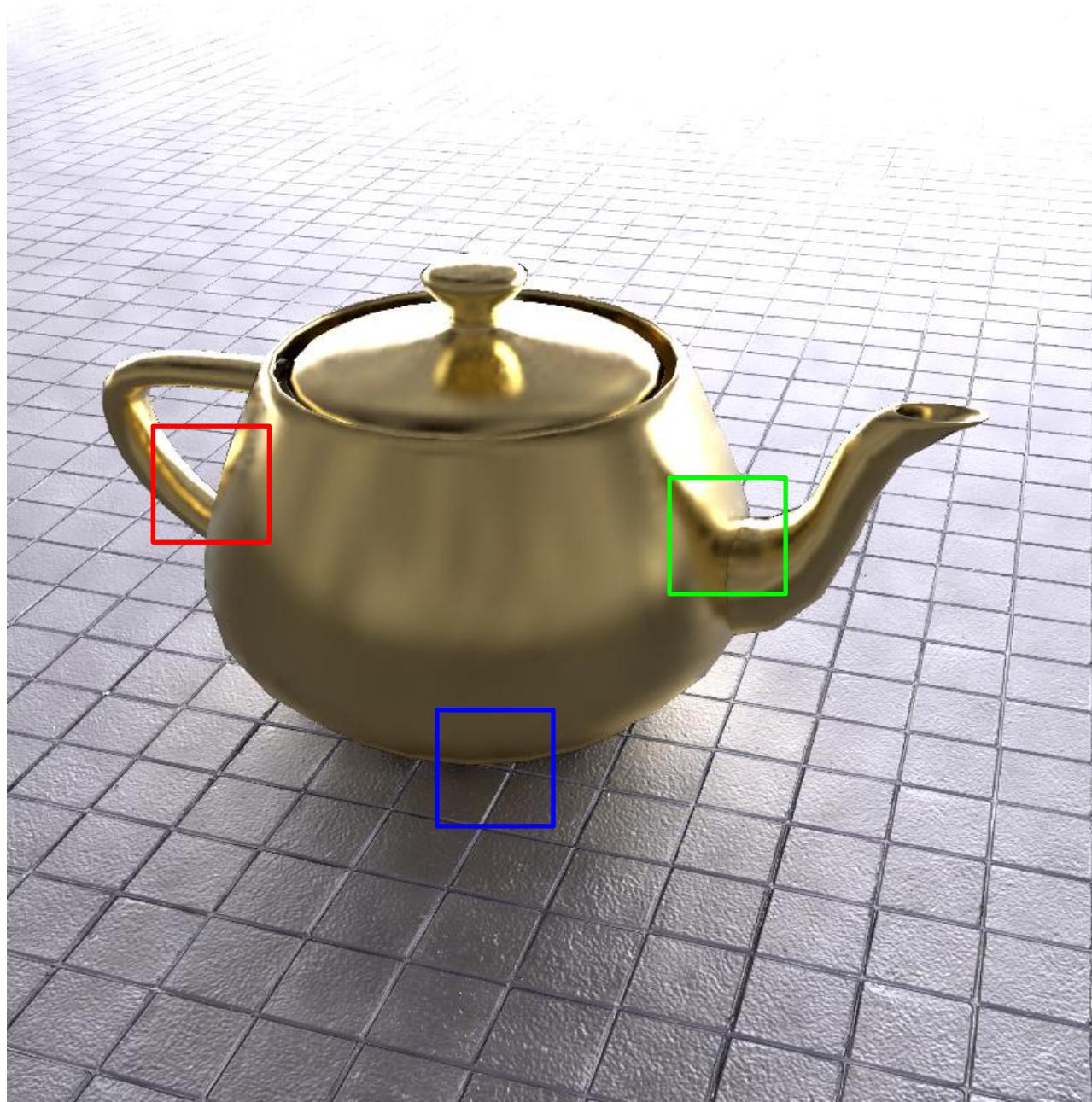
# TEAPOT

Random Parameter Filtering [Sen and Darabi, ACMTOG 2012], 8 spp, 374.4 sec.



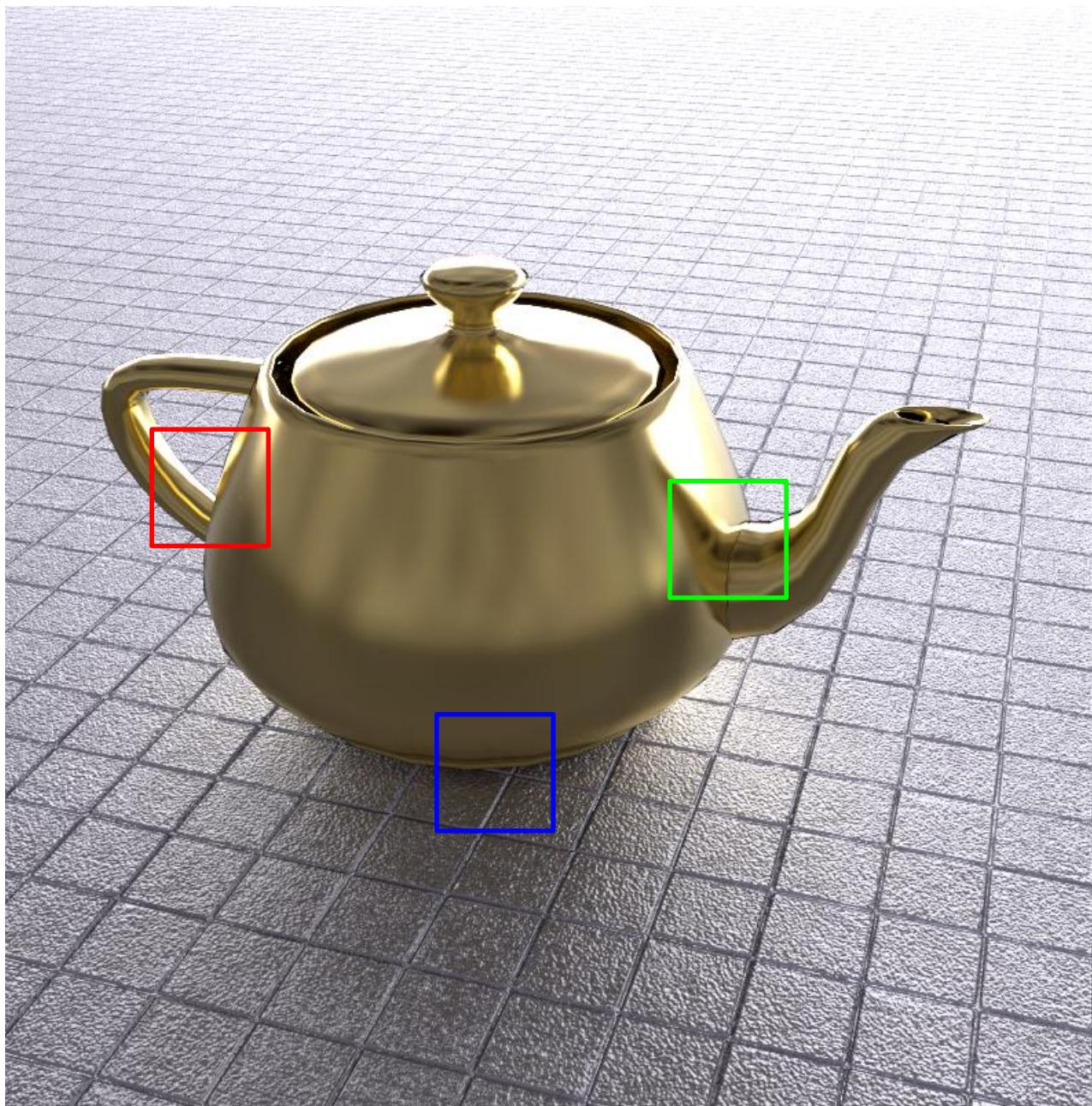
# TEAPOT

SURE-based Optimization (Our Approach), 8 spp, 40.4 sec.



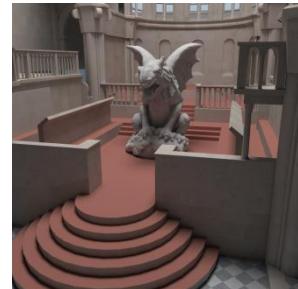
# TEAPOT

Reference, 4096 spp



# GARGOYLE

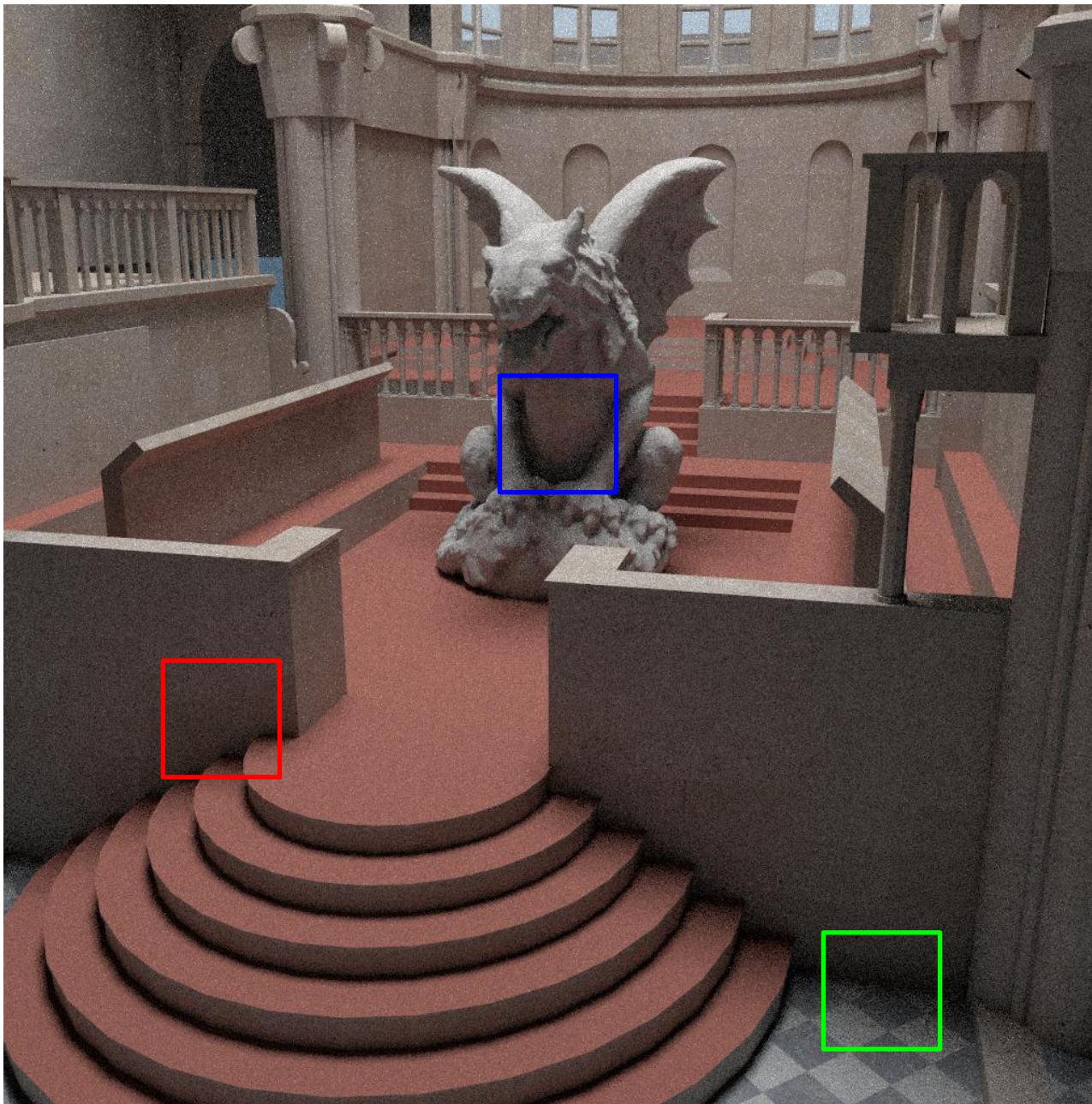
Global Illumination (One-Bounce Path Tracing)



1024 x 1024

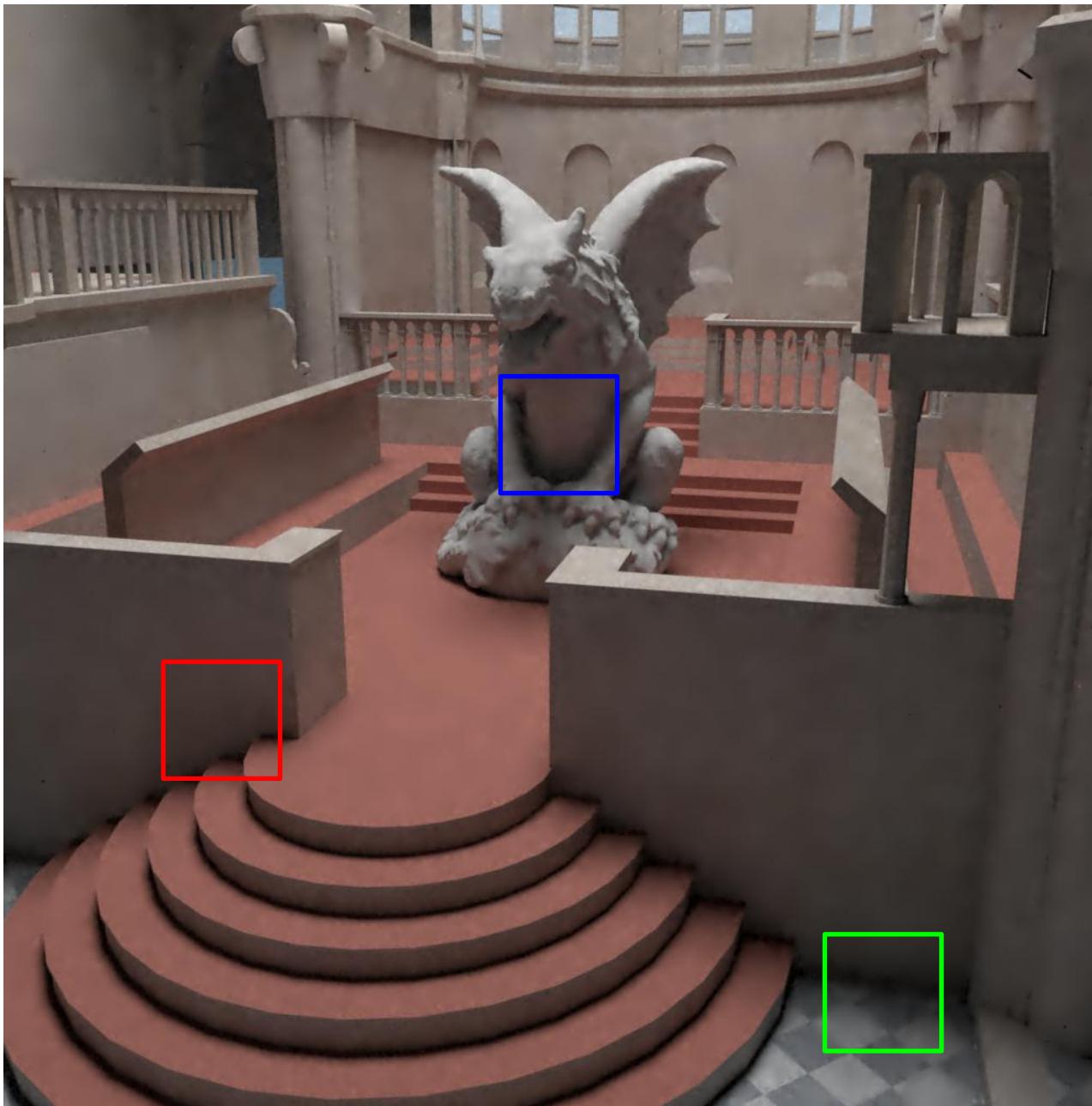
# GARGOYLE

Equal-time Monte Carlo, 56 spp, 161.7 sec.



# GARGOYLE

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 43.92 spp, 167.4 sec.



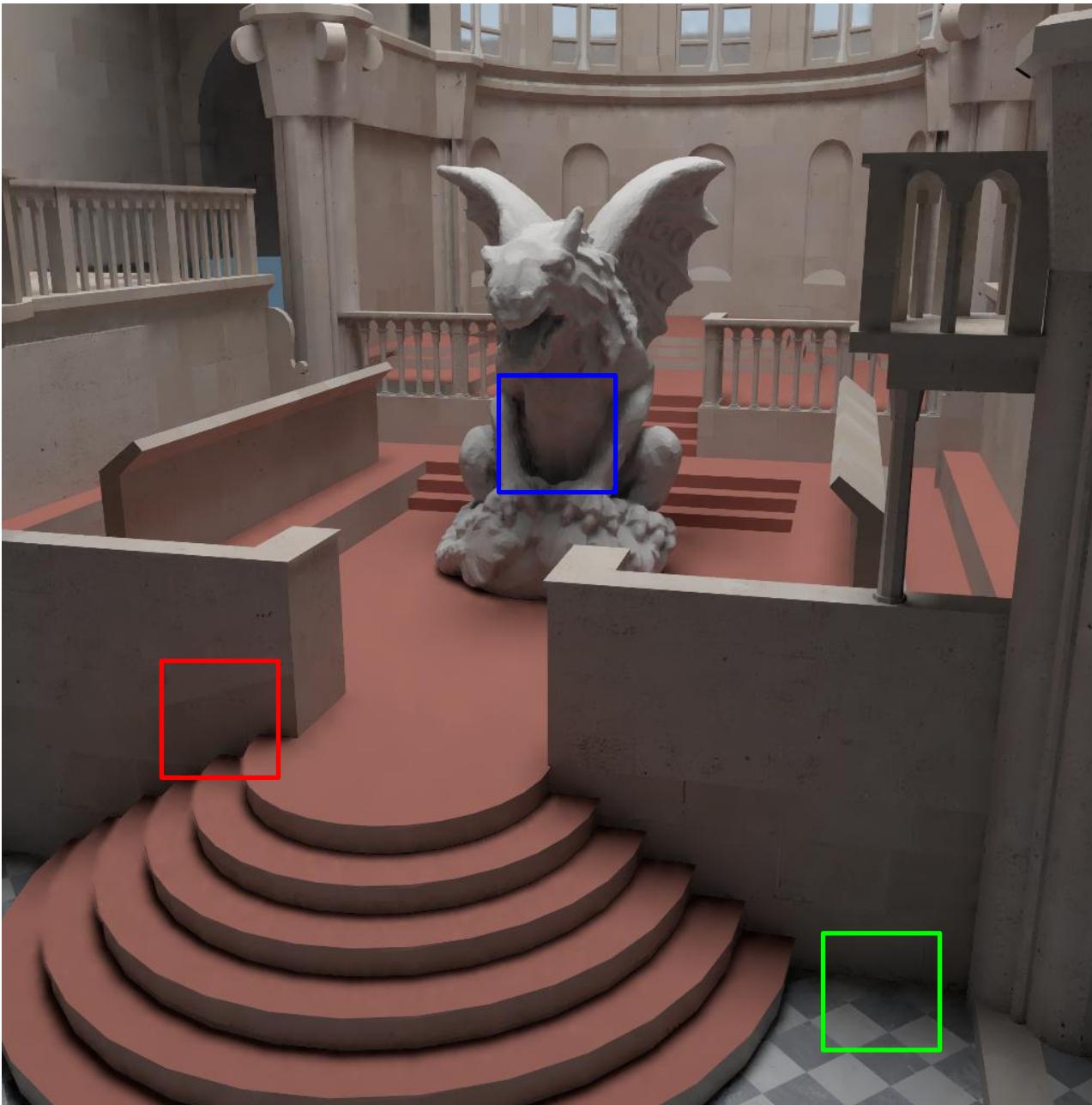
# GARGOYLE

Random Parameter Filtering [Sen and Darabi, ACMTOG 2012], 8 spp, 608.3 sec.



# GARGOYLE

SURE-based Optimization (Our Approach), 30.90 spp, 160.0 sec.



# GARGOYLE

Reference, 4096 spp



# SANMIGUEL

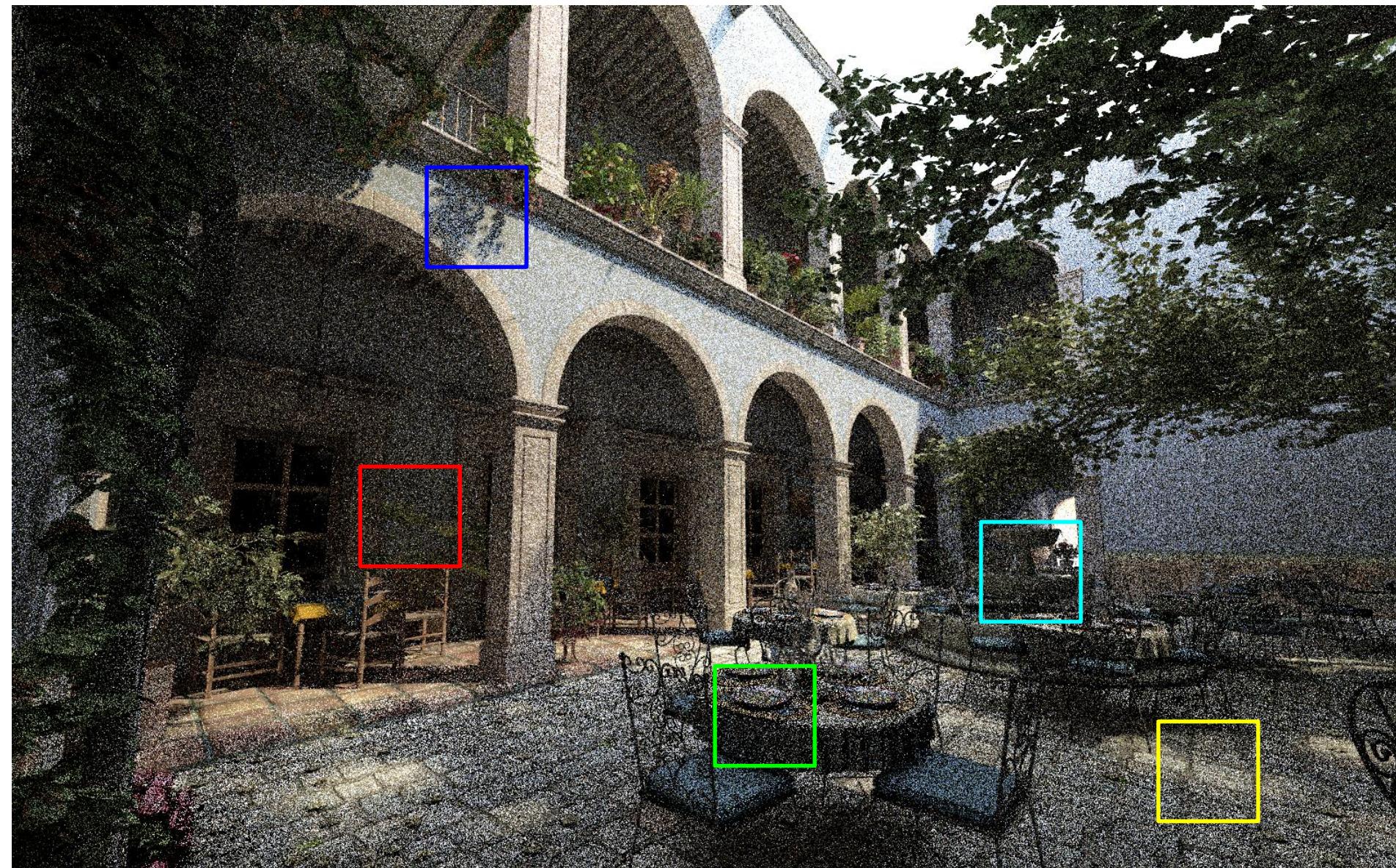
Global Illumination (Path Tracing)



1580 x 986

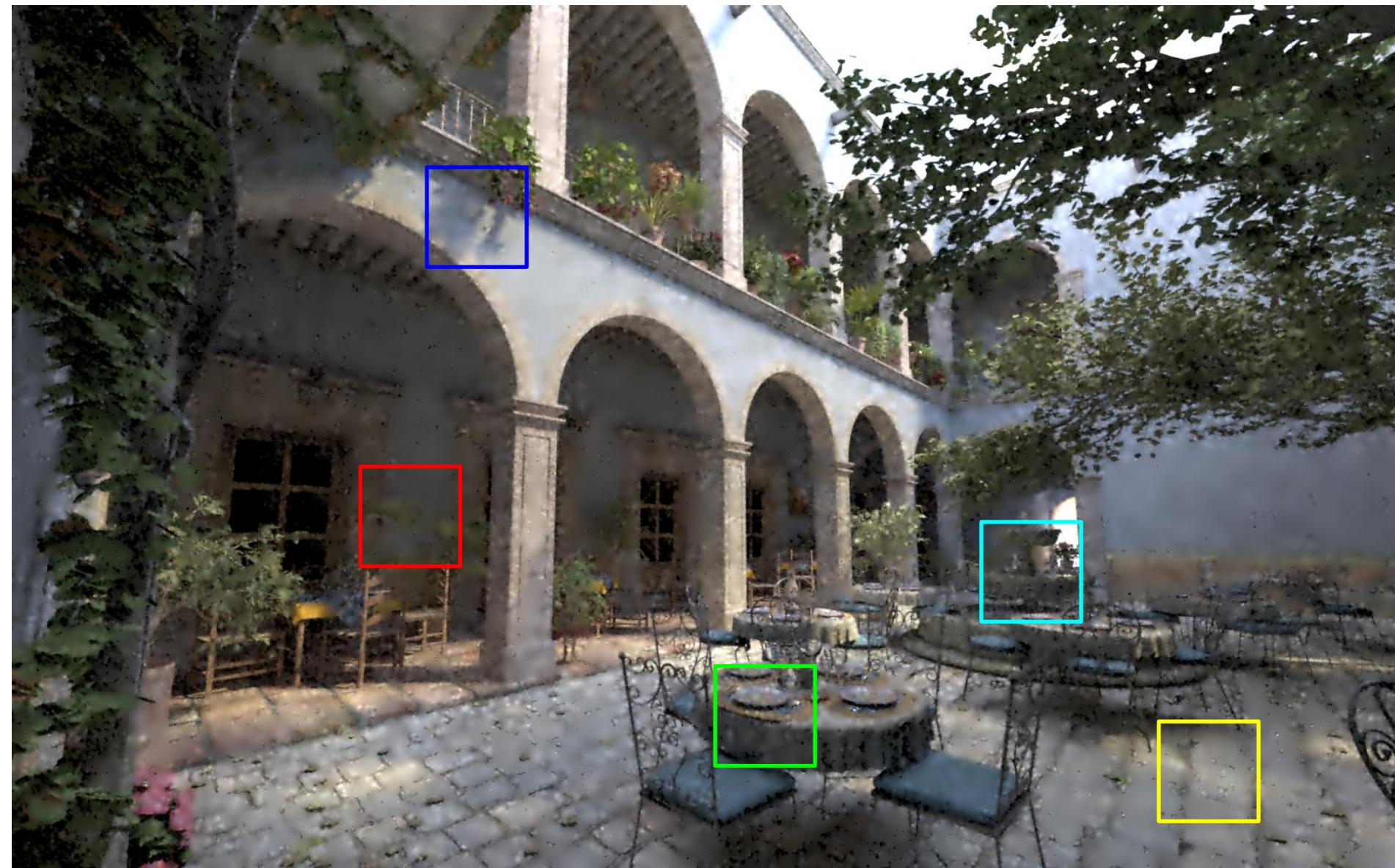
# SANMIGUEL

Equal-time Monte Carlo, 70 spp, 1209.4 sec.



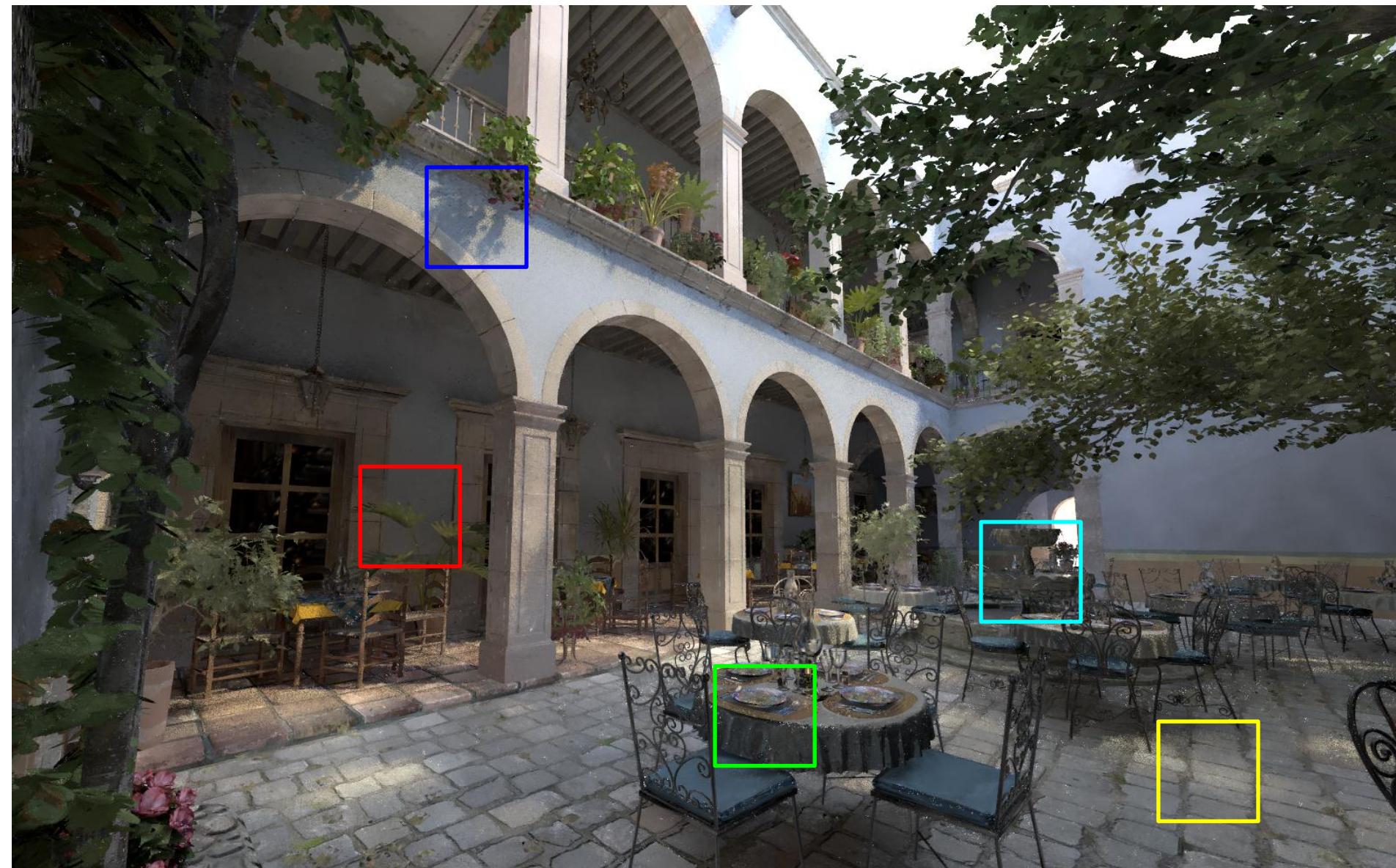
**SANMIGUEL**

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 63.59 spp, 1239.9 sec.



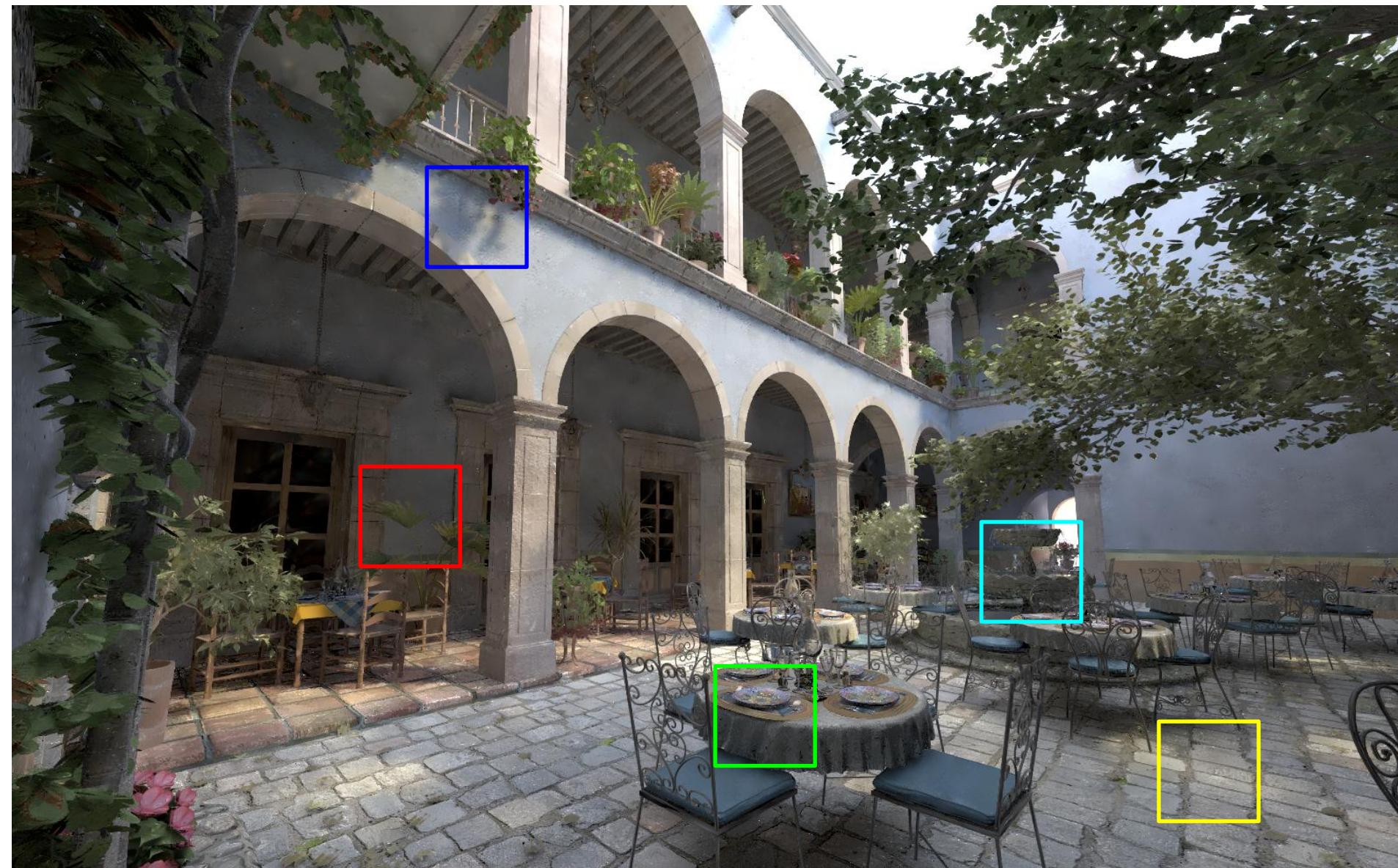
**SANMIGUEL**

**Random Parameter Filtering [Sen and Darabi, ACMTOG 2012]**, 16 spp, 2617.9 sec.



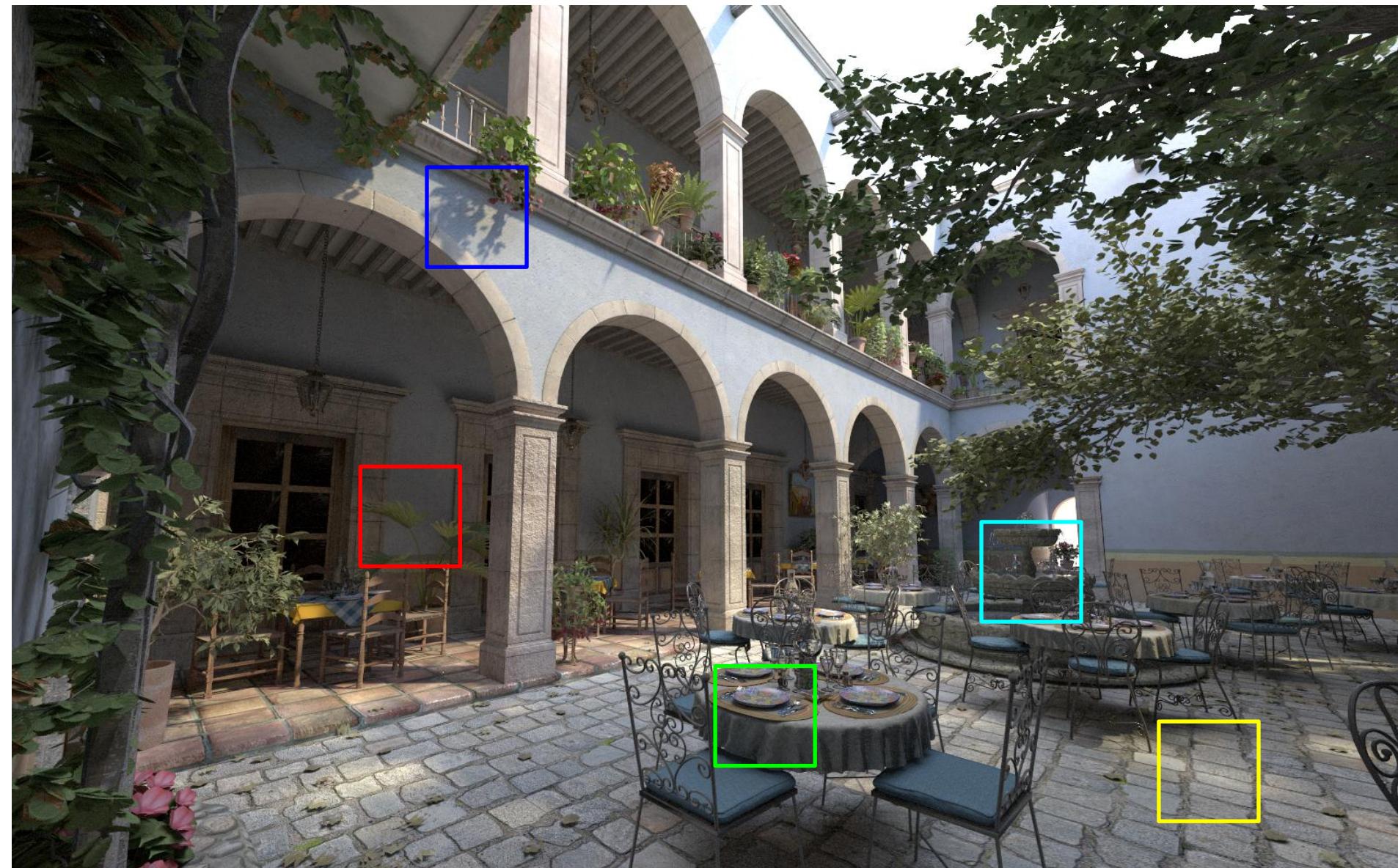
**SANMIGUEL**

**SURE-based Optimization (Our Approach)**, 61.69 spp, 1228.9 sec.



# SANMIGUEL

Reference, 8192 spp



# PART II

## Equal-Sample Comparison

### Compared Methods:

- Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011]
- Random Parameter Filtering [Sen and Darabi, ACM TOG 2012]
- SURE-based Optimization (our approach, using cross bilateral filters)

# SPONZA

## Global Illumination (Path Tracing) Motion Blur



1600 x 1200

**SPONZA**

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 16 spp, 210.0 sec.



**SPONZA**

**Random Parameter Filtering [Sen and Darabi, ACMTOG 2012], 16 spp, 1676.1 sec.**



**SPONZA**

**SURE-based Optimization (Our Approach)**, 16 spp, 273.3 sec.



**SPONZA**

**Reference**, 8192 spp



# TOWN

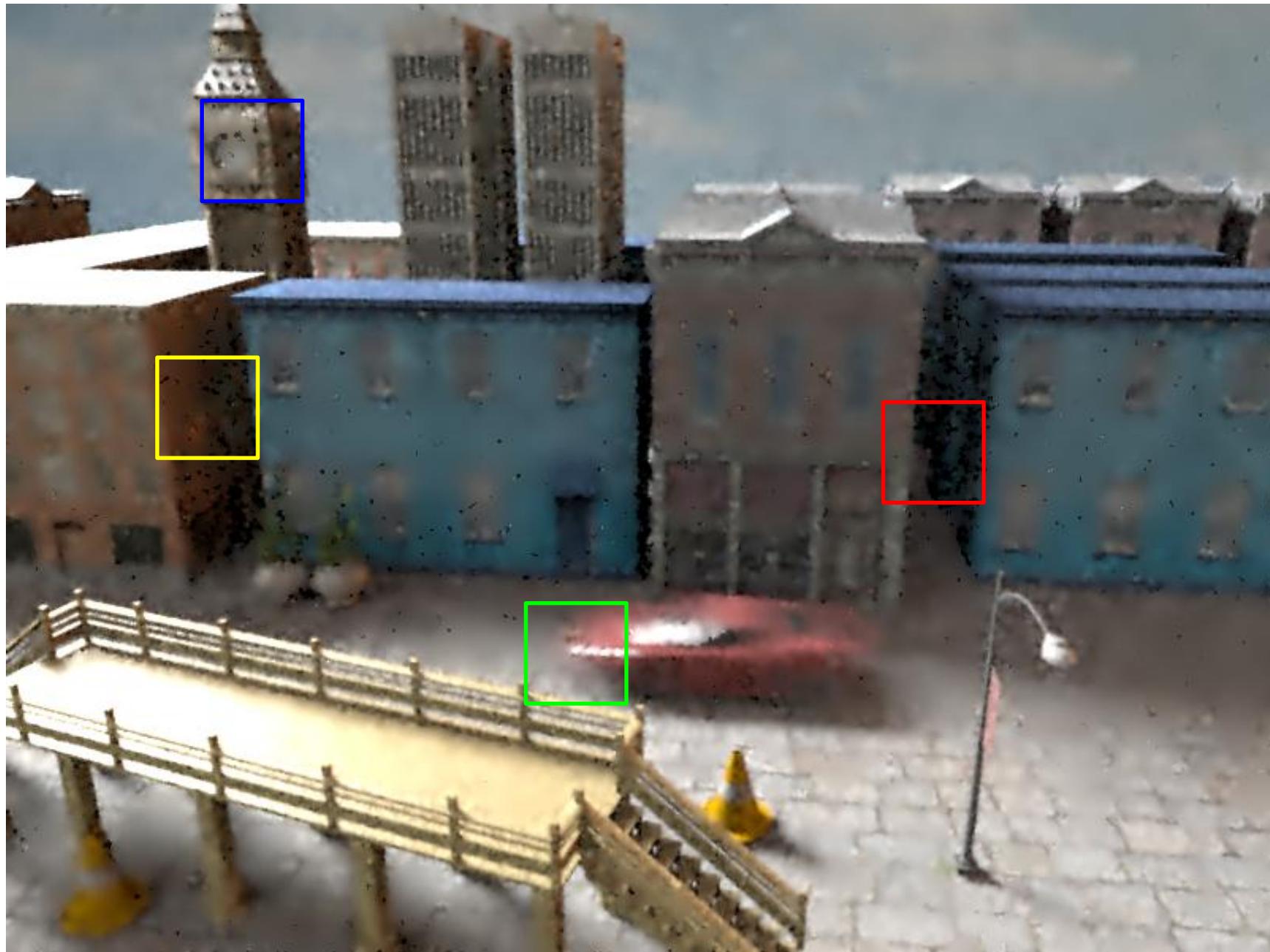
Environment Lighting  
Area Lighting  
Motion Blur



800 x 600

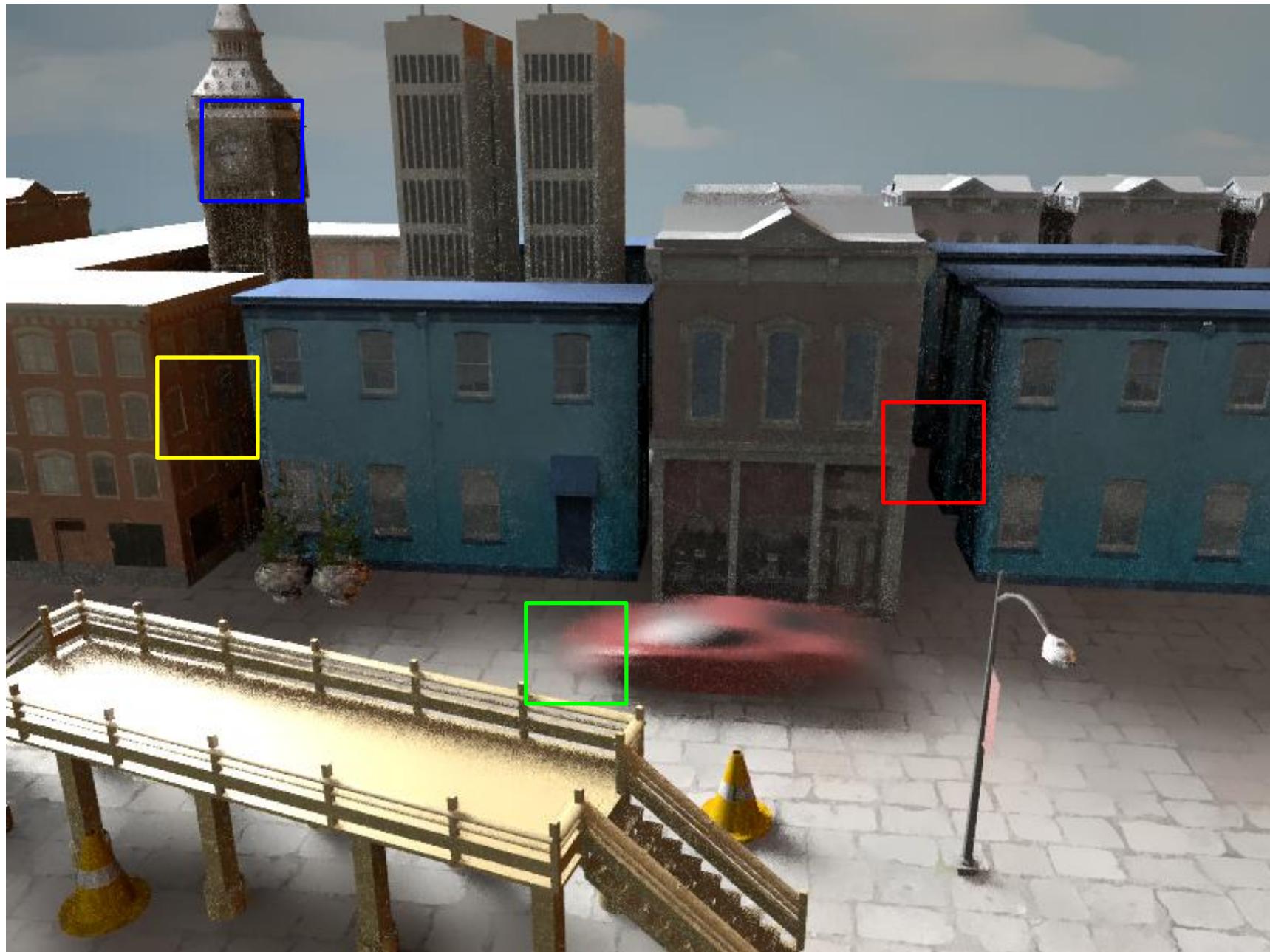
**TOWN**

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 8 spp, 9.4 sec.



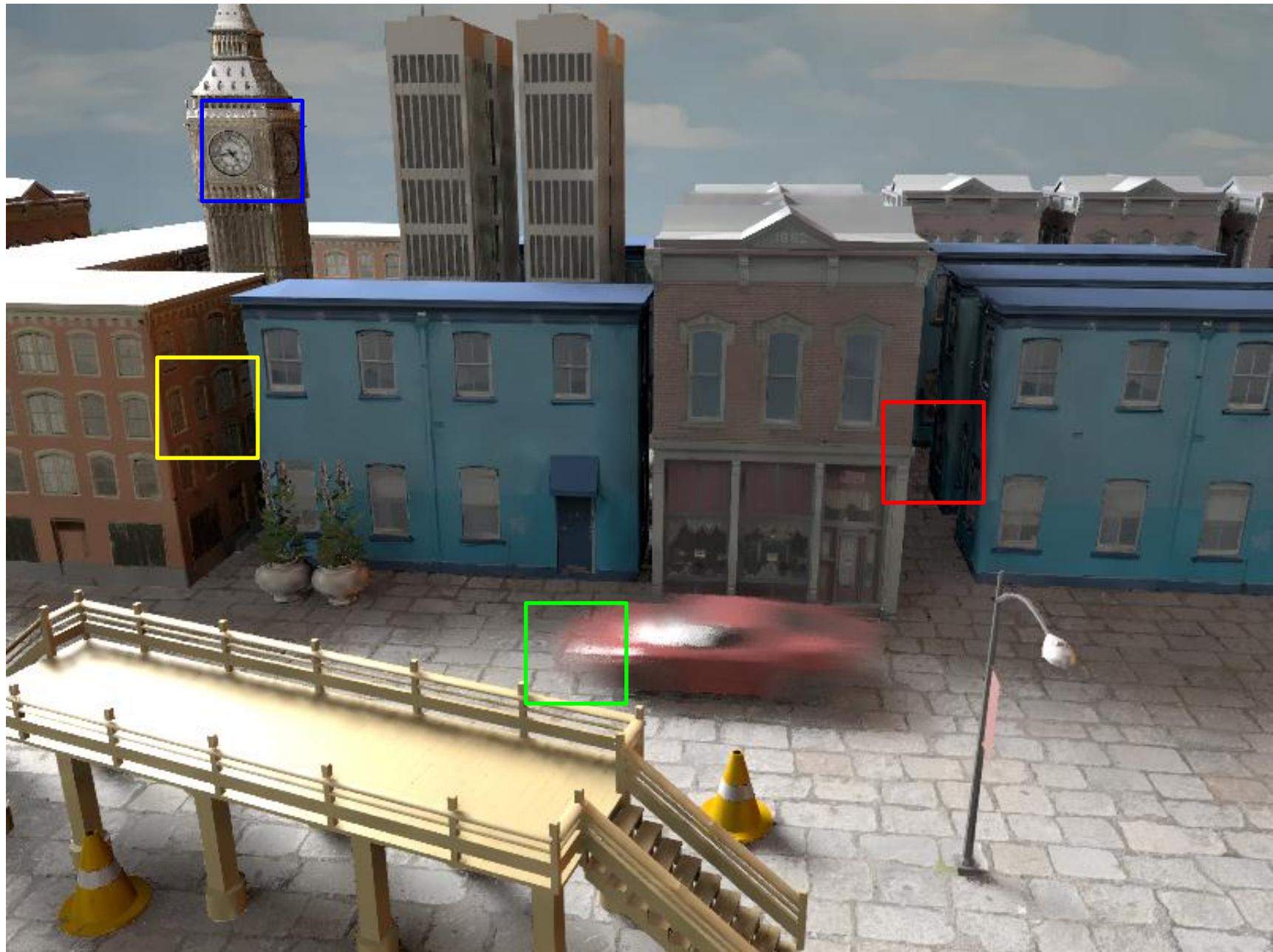
TOWN

Random Parameter Filtering [Sen and Darabi, ACMTOG 2012], 8 spp, 272.4 sec.



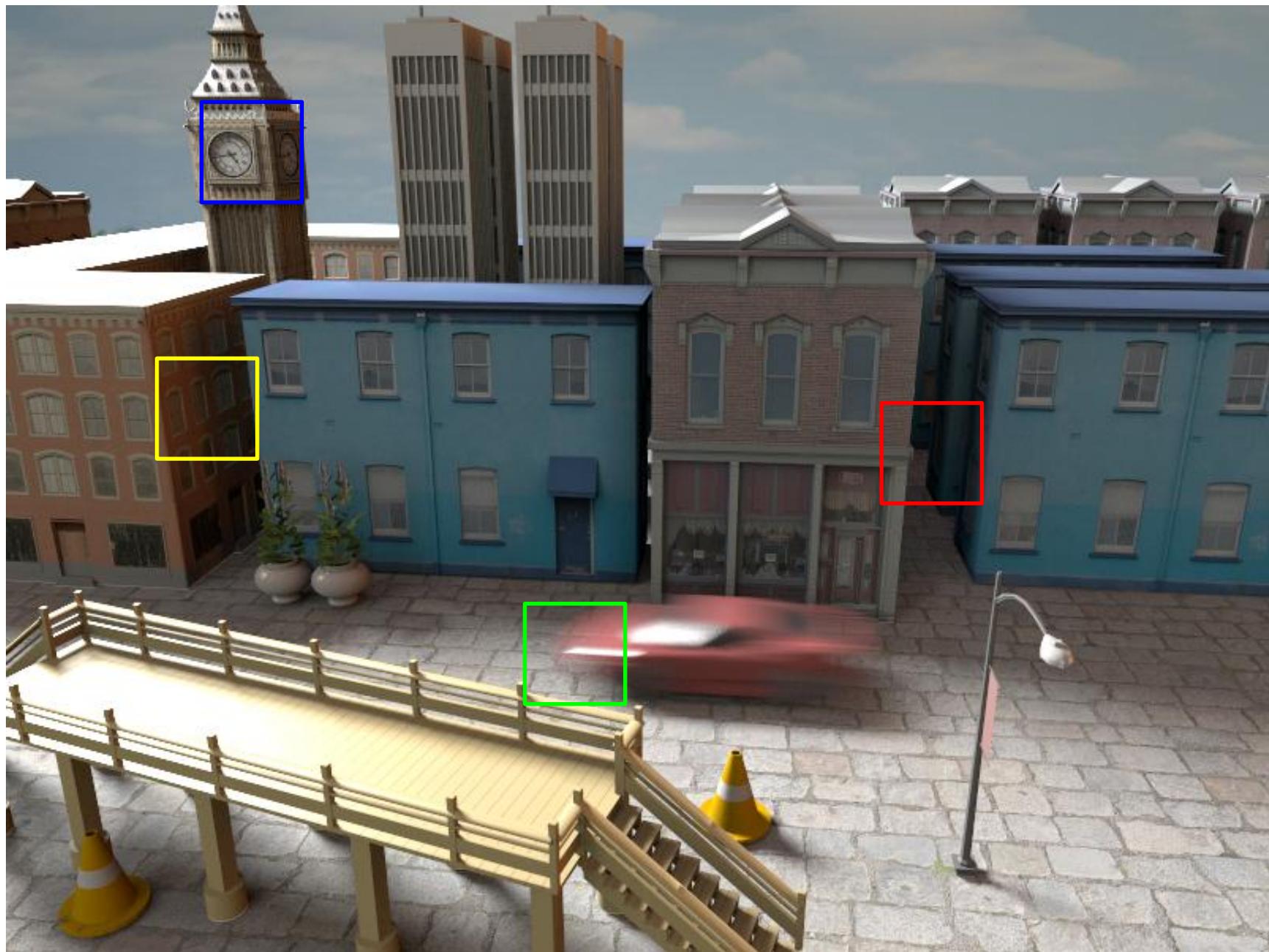
**TOWN**

**SURE-based Optimization (Our Approach), 8 spp, 20.0 sec.**



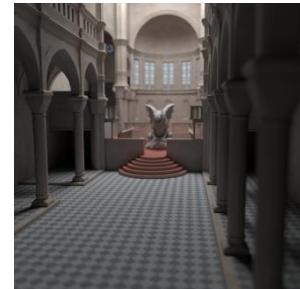
**TOWN**

**Reference**, 4096 spp



# SIBENIK

Global Illumination (One-Bounce Path Tracing)  
Depth of Field



1024 x 1024

# SIBENIK

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 8 spp, 27.6 sec.



# SIBENIK

Random Parameter Filtering [Sen and Darabi, ACMTOG 2012], 8 spp, 363.0 sec.



# SIBENIK

**SURE-based Optimization (Our Approach), 8 spp, 64.2 sec.**



**SIBENIK**

**Reference**, 4096 spp



# TEAPOT

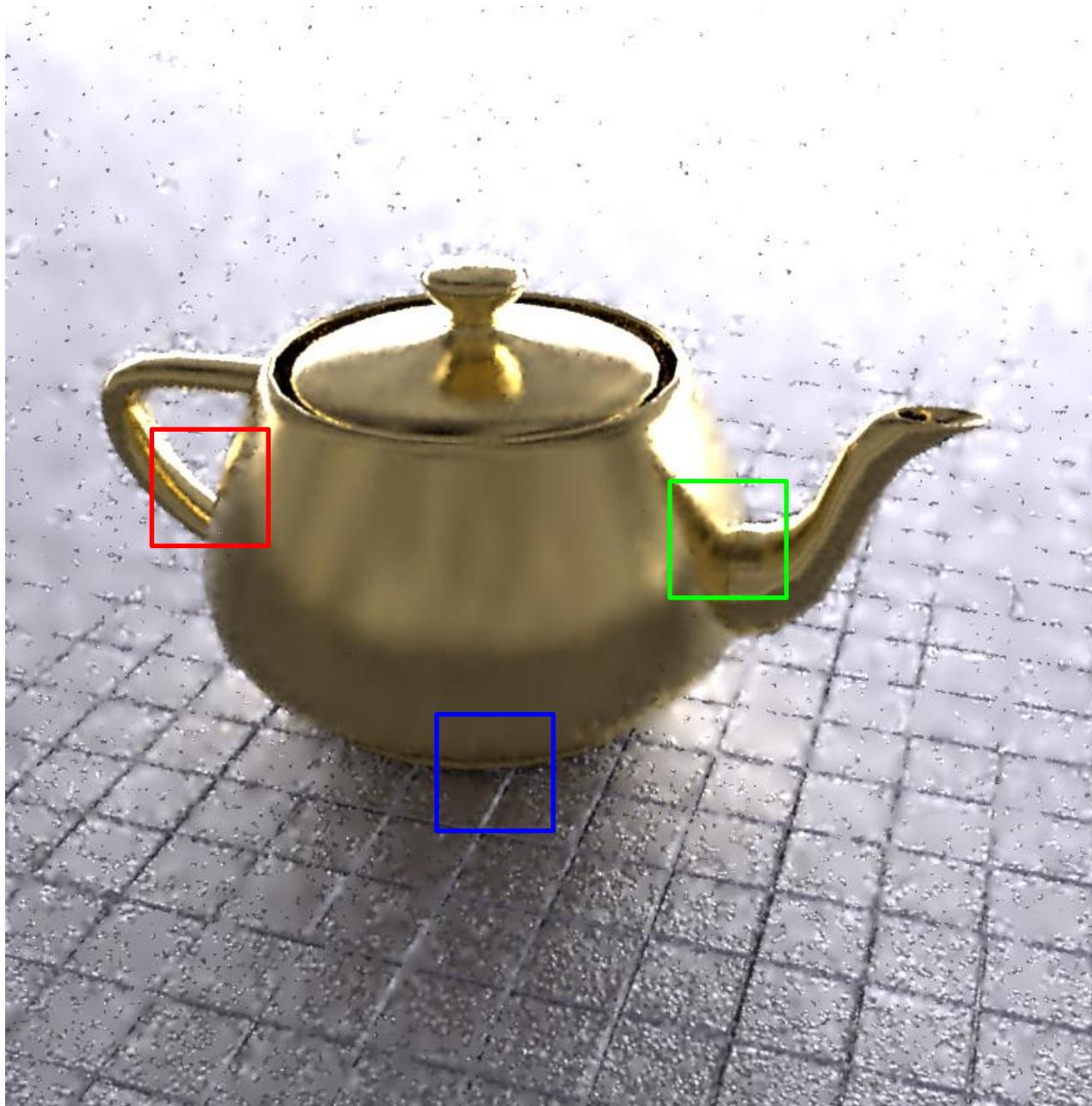
Environment Lighting  
Glossy Reflection



800 x 800

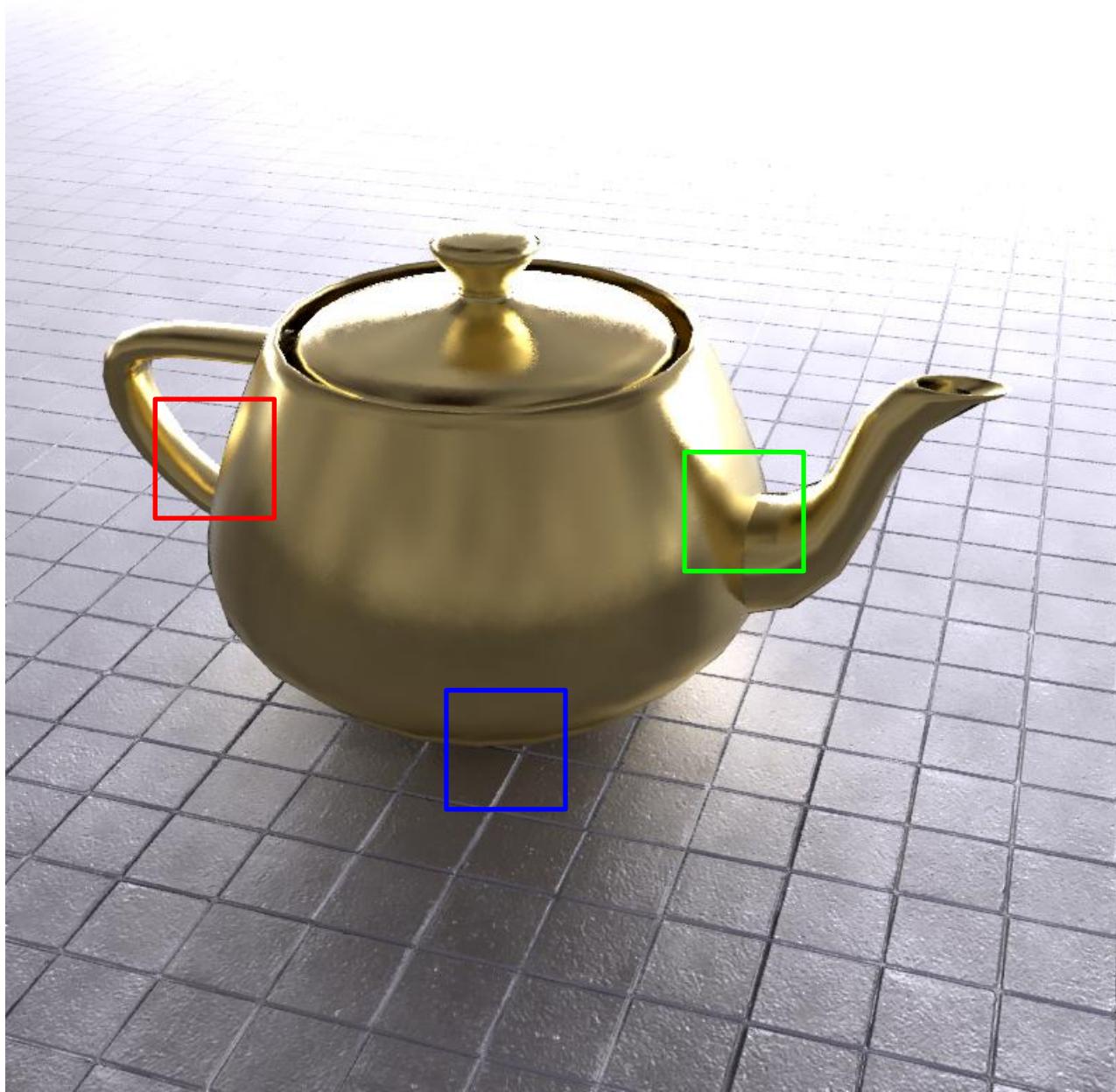
# TEAPOT

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 8 spp, 14.1 sec.



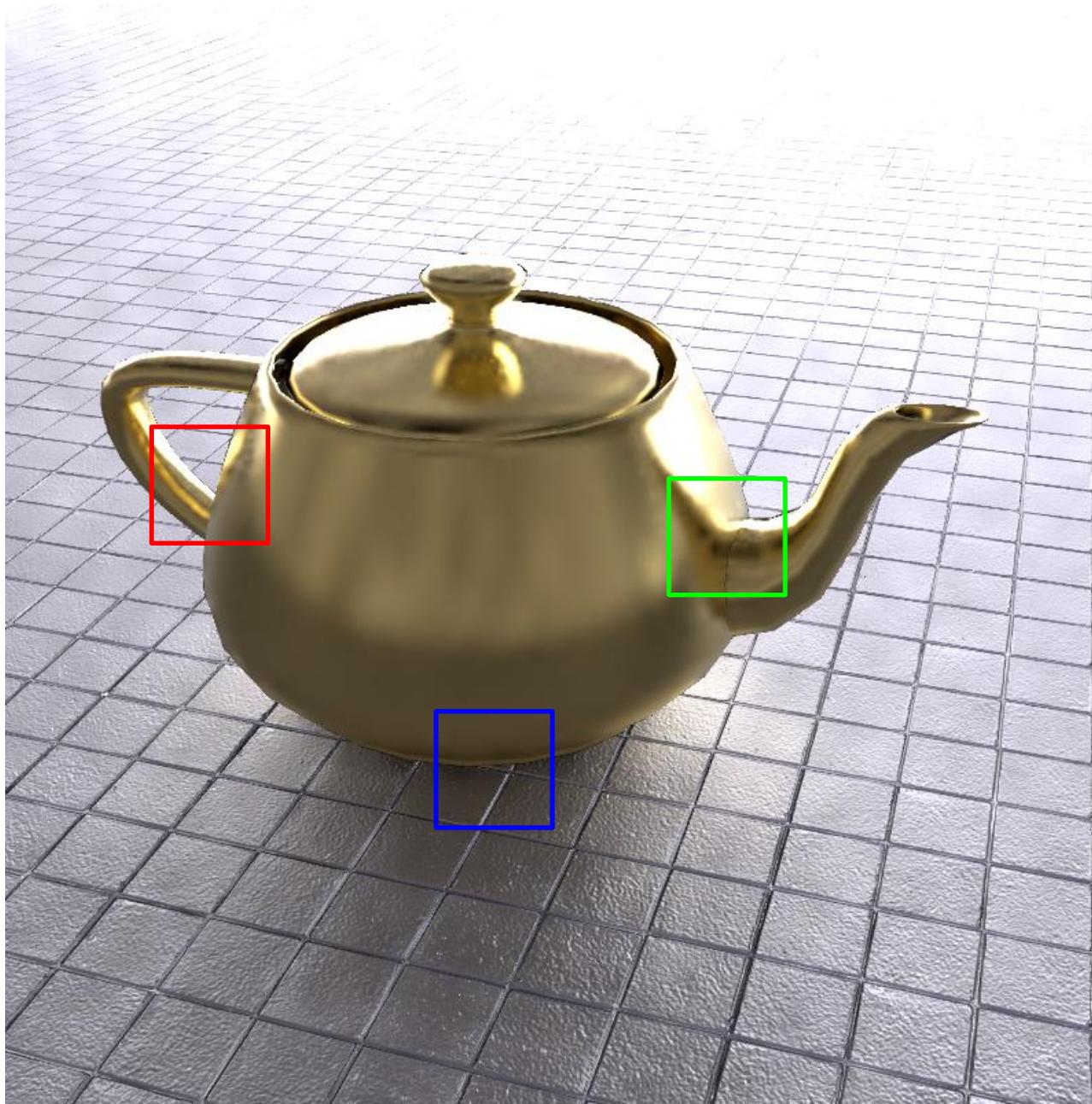
# TEAPOT

Random Parameter Filtering [Sen and Darabi, ACMTOG 2012], 8 spp, 374.4 sec.



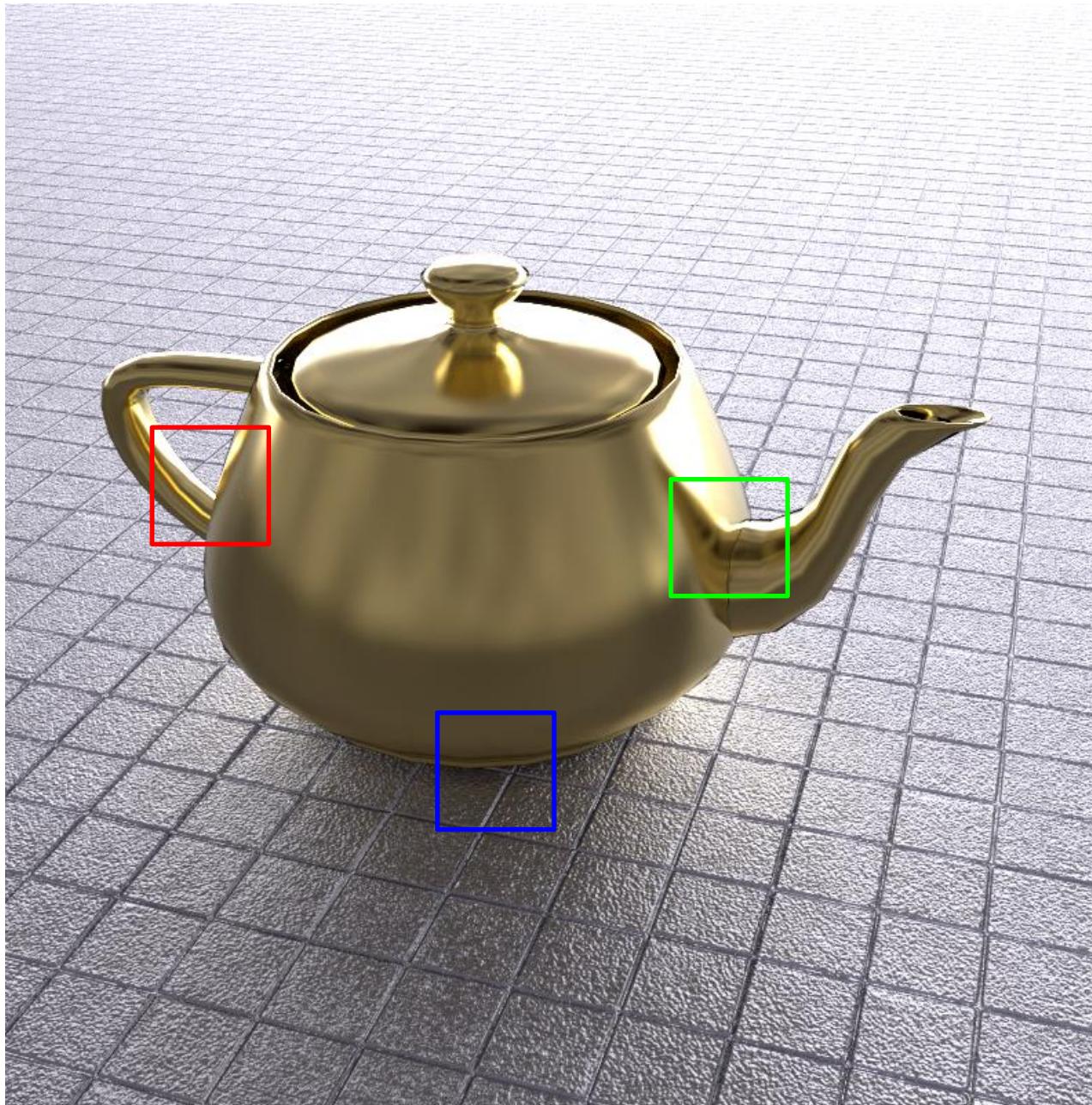
# TEAPOT

SURE-based Optimization (Our Approach), 8 spp, 40.4 sec.



# TEAPOT

Reference, 4096 spp



# GARGOYLE

Global Illumination (One-Bounce Path Tracing)



1024 x 1024

# GARGOYLE

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 8 spp, 28.6 sec.



# GARGOYLE

Random Parameter Filtering [Sen and Darabi, ACMTOG 2012], 8 spp, 608.3 sec.



# GARGOYLE

SURE-based Optimization (Our Approach), 8 spp, 68.3 sec.



# GARGOYLE

Reference, 4096 spp



# SANMIGUEL

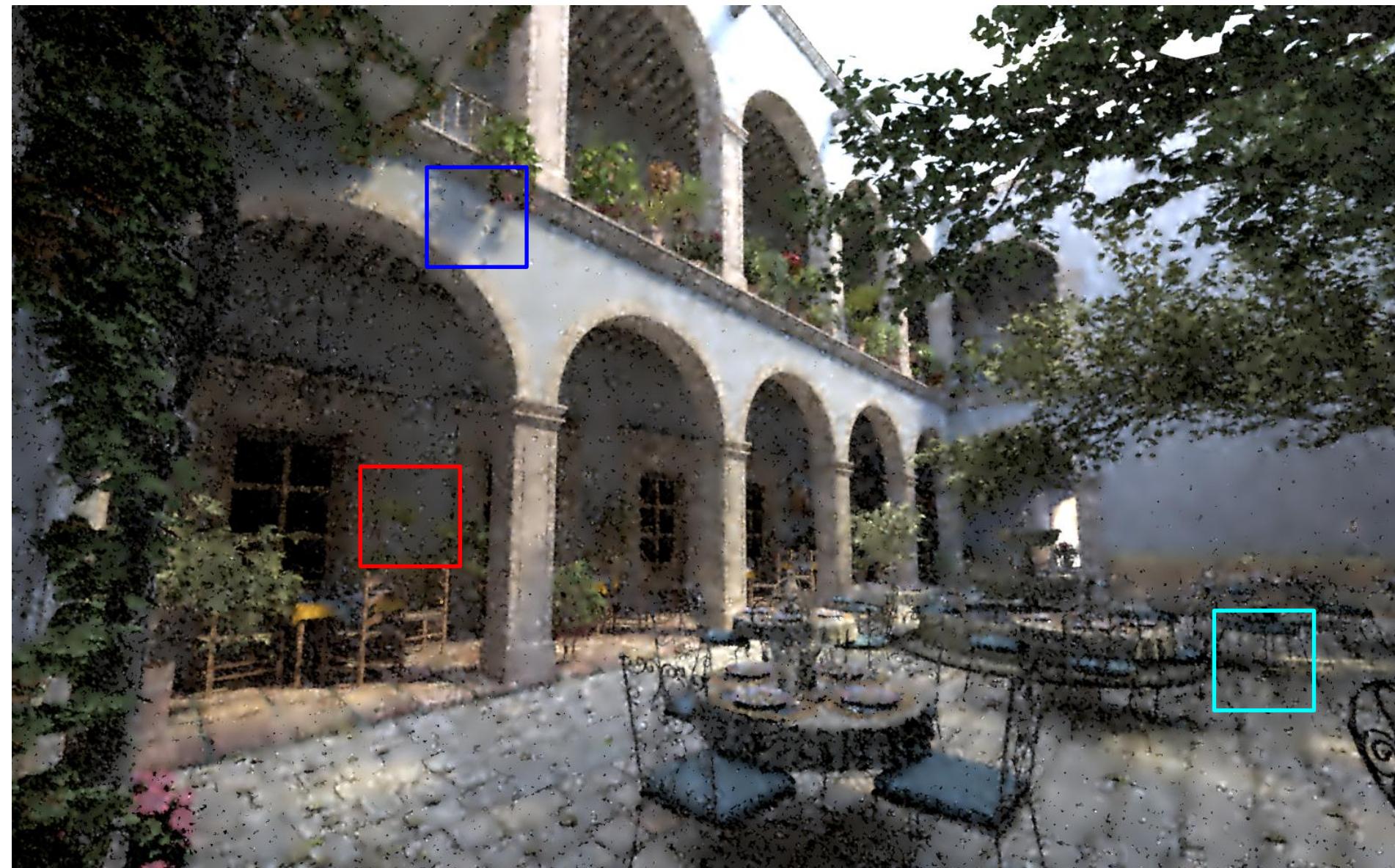
Global Illumination (Path Tracing)



1580 x 986

# SANMIGUEL

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011], 16 spp, 304.4 sec.



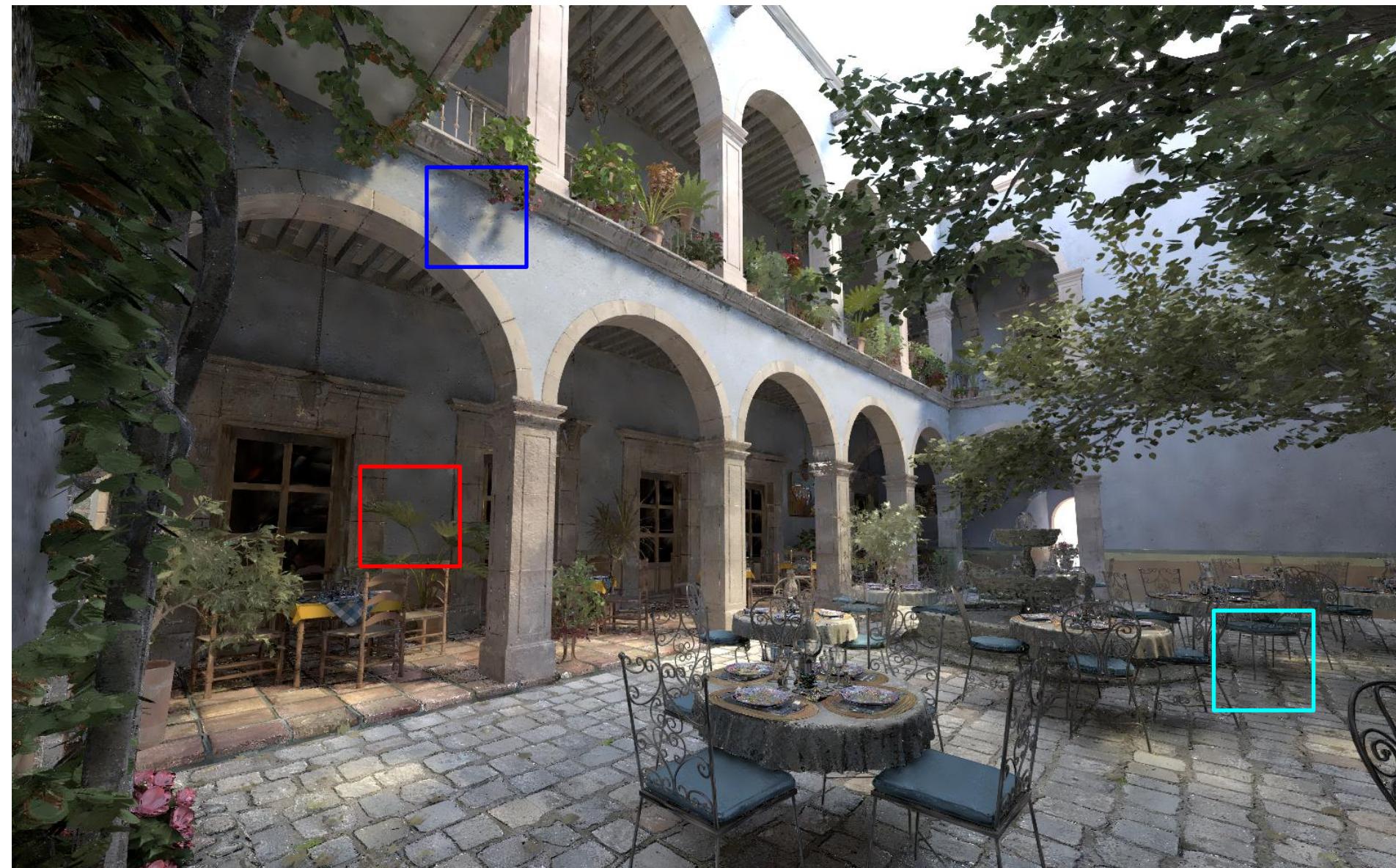
**SANMIGUEL**

**Random Parameter Filtering [Sen and Darabi, ACMTOG 2012]**, 16 spp, 2617.9 sec.



**SANMIGUEL**

**SURE-based Optimization (Our Approach)**, 16 spp, 336.3 sec.



**SANMIGUEL**

**Reference**, 8192 spp



# PART III

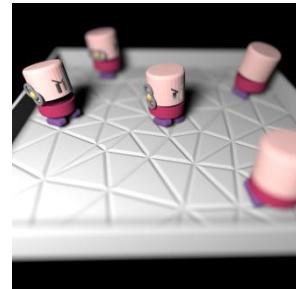
## Equal-Time Comparison for Isotropic Gaussian Filters

### Compared Methods:

- Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011]
- SURE-based Optimization (our approach, using isotropic Gaussian filters)

# TOASTERS

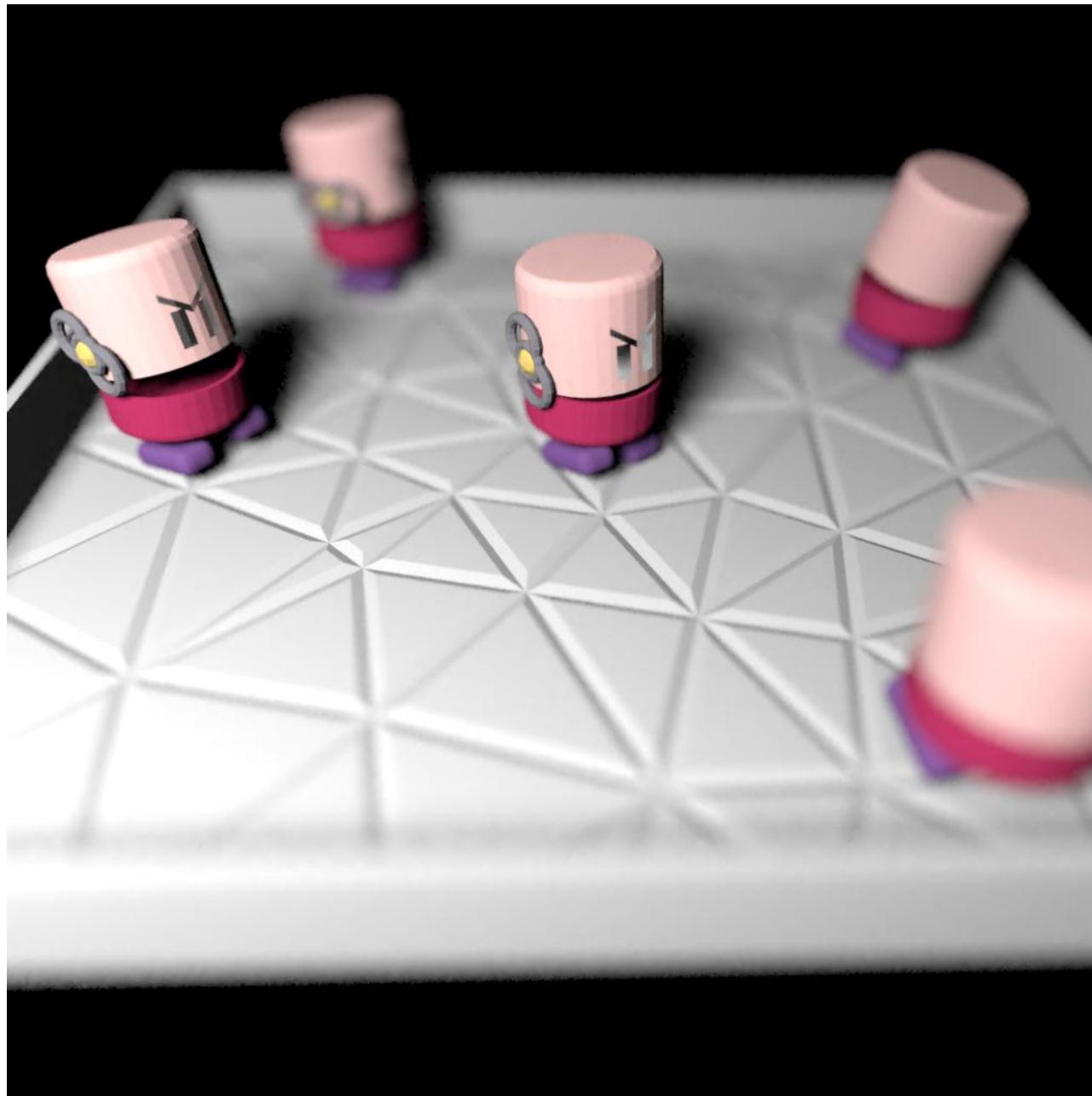
Area Lighting  
Depth of Field



1024 x 1024

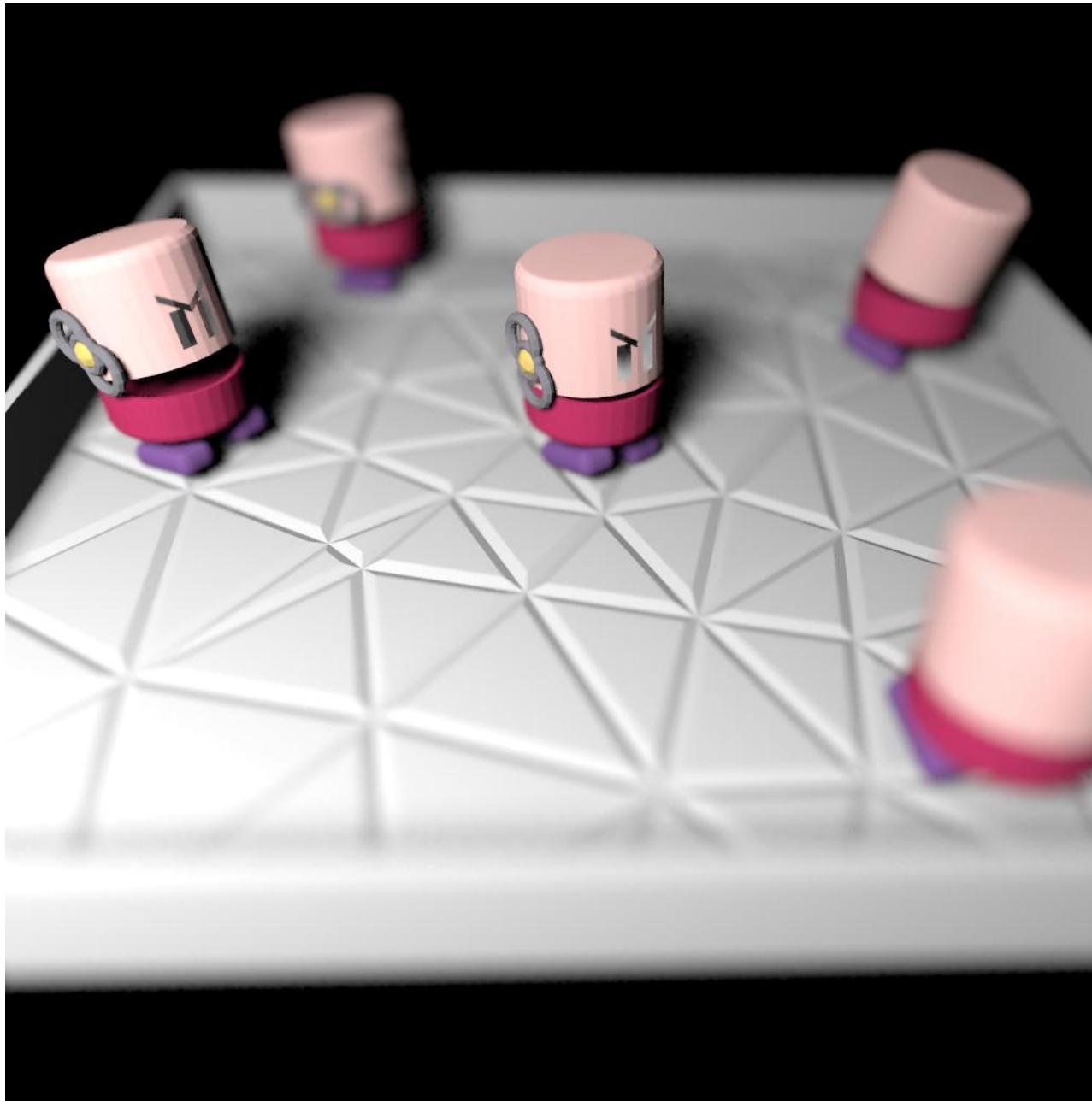
# TOASTERS

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011]



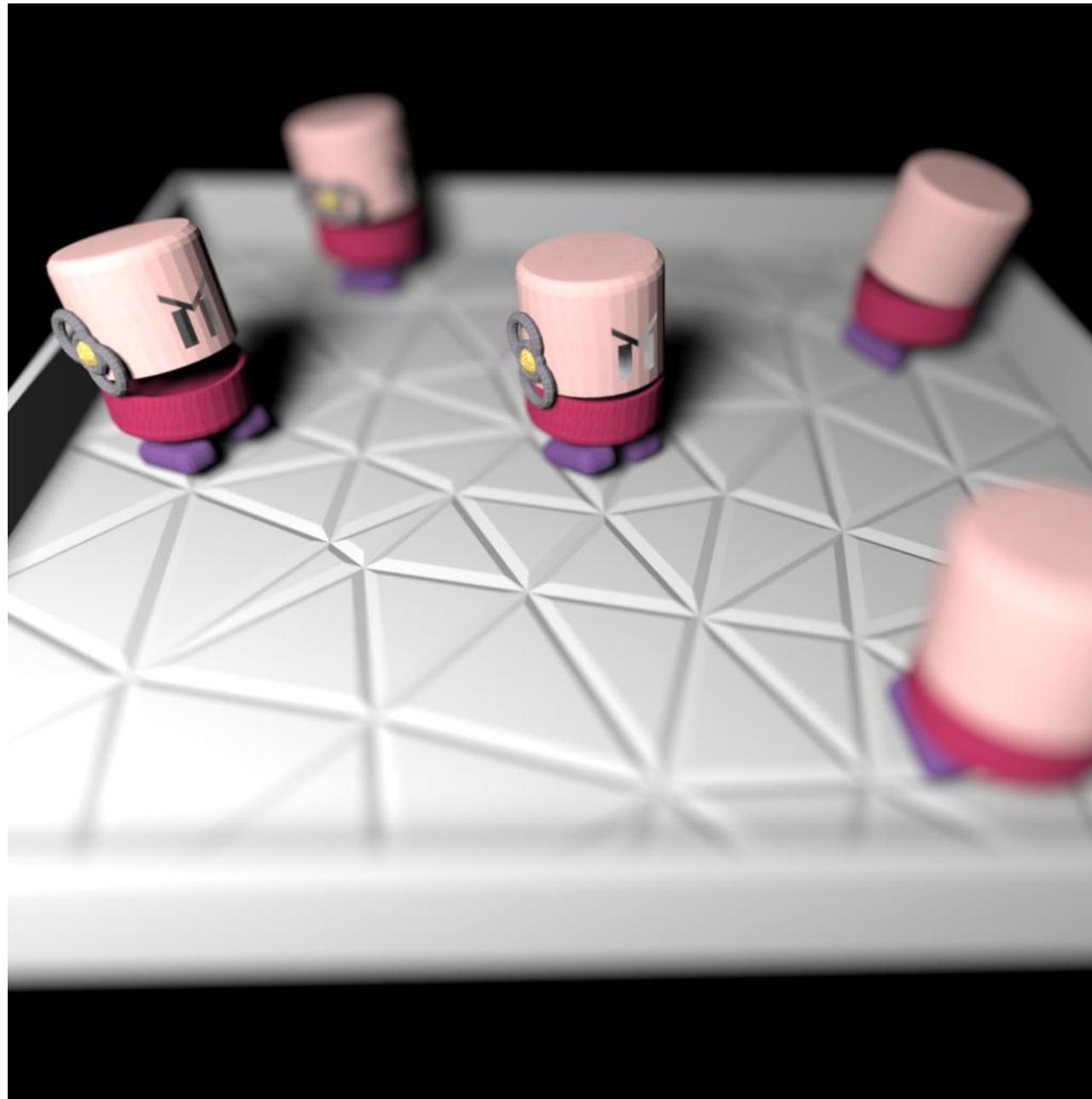
## TOASTERS

SURE-based Optimization (Our Approach), using Isotropic Gaussian Filters



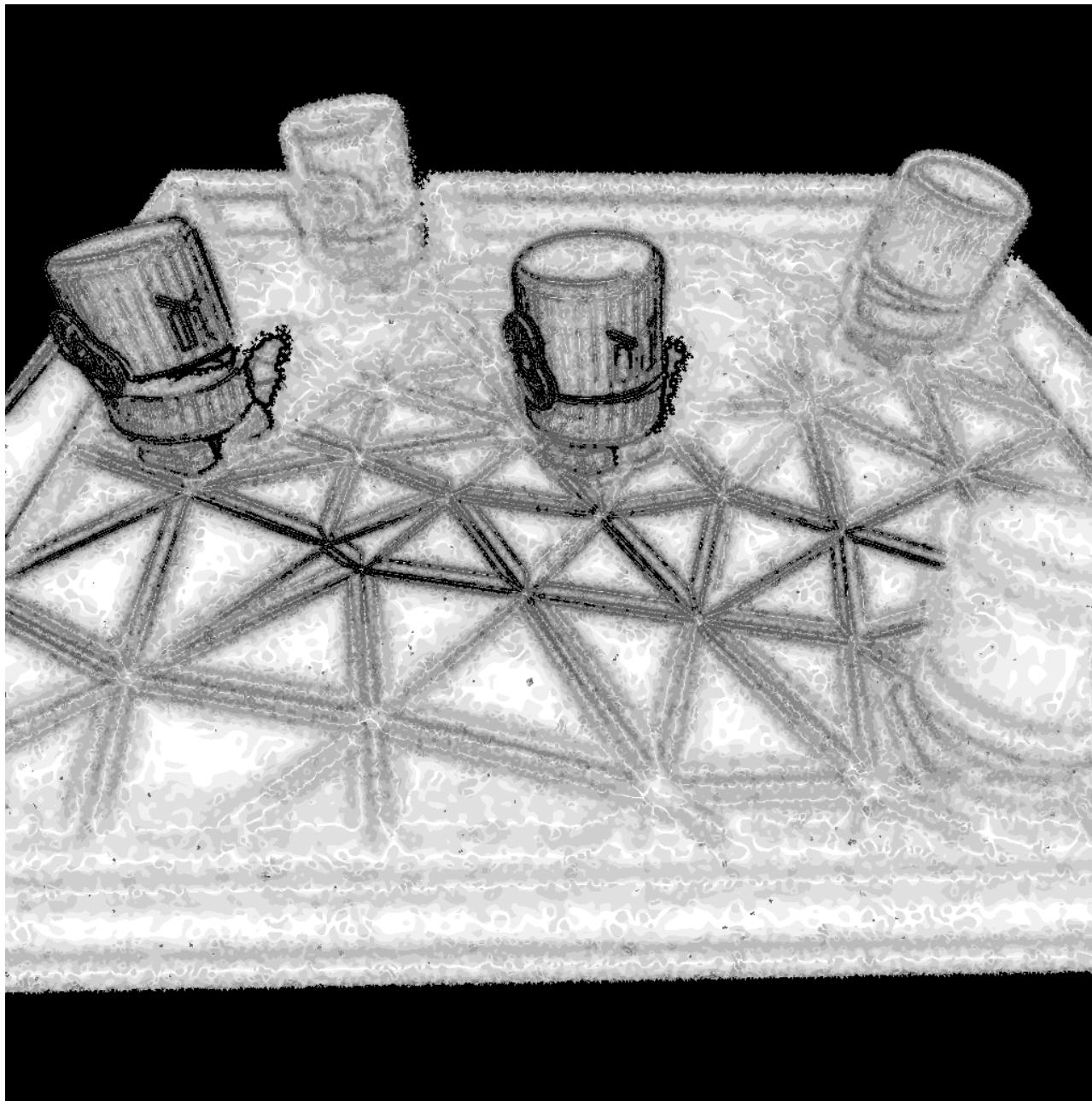
# TOASTERS

Reference, 4096 spp



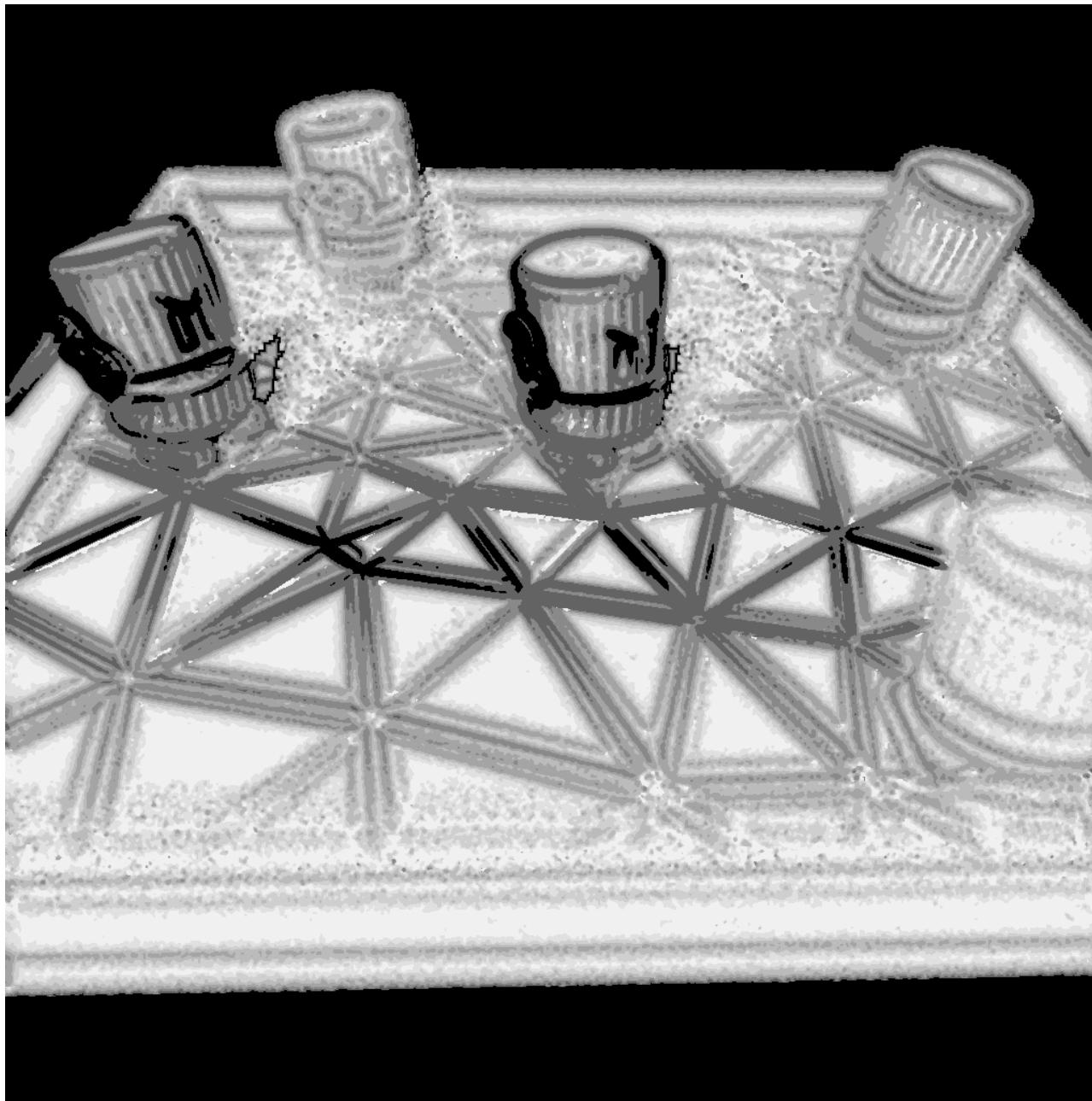
# TOASTERS – Scale Selection Map

Greedy Error Minimization [Rousselle et al., SIGGRAPH Asia 2011]



## TOASTERS - Scale Selection Map

SURE-based Optimization (Our Approach), using Isotropic Gaussian Filters



# PART IV

# Equal-Time Comparison for

# Cross Non-local Means Filters

## Compared Methods:

- Global cross non-local means filters
- SURE-based Optimization (our approach, using cross non-local means filters)

# TOWN

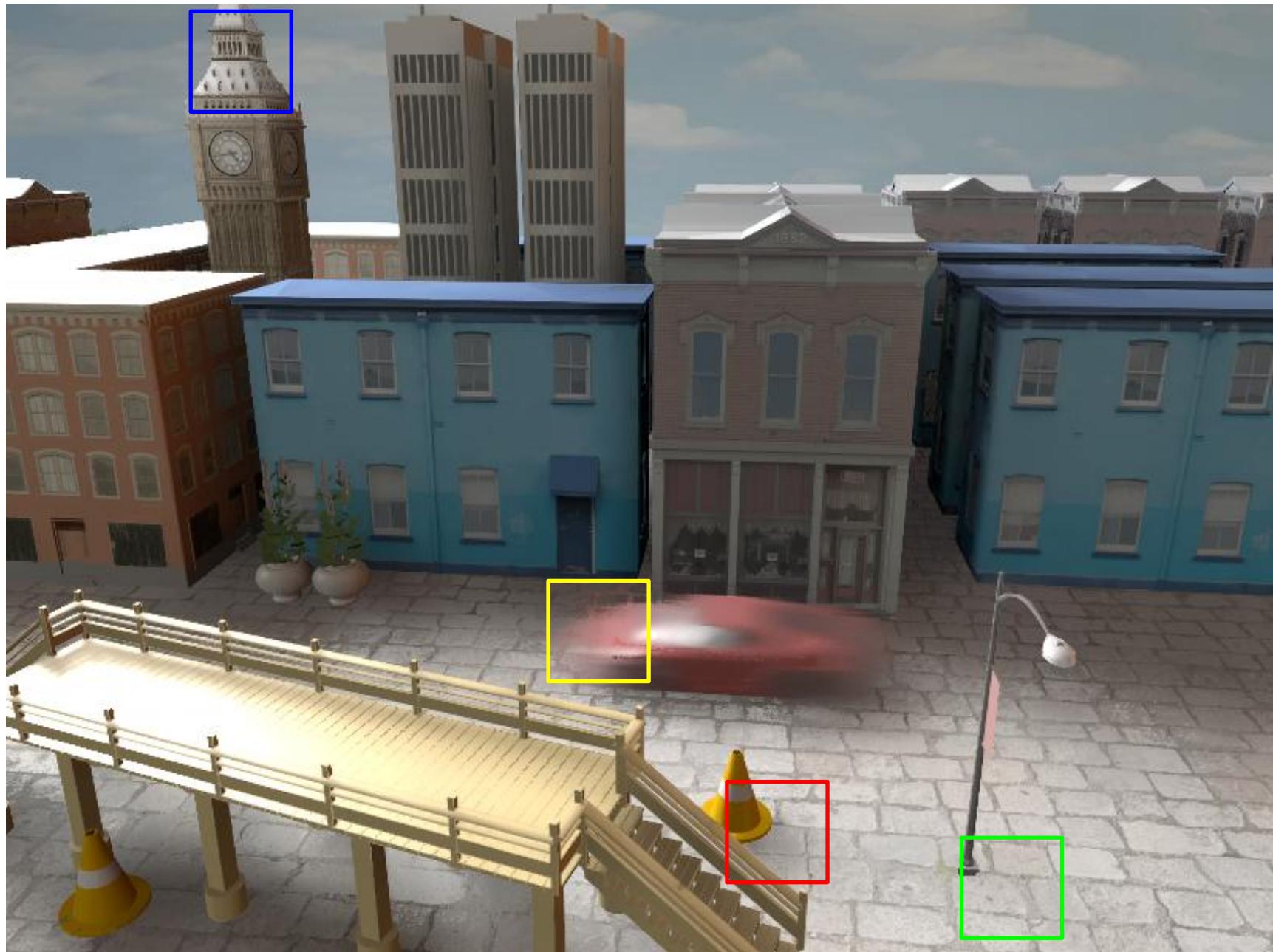
Environment Lighting  
Area Lighting  
Motion Blur



800 x 600

**TOWN**

**Global Non-local Means Filter, 41.2 spp**



**TOWN**

**SURE-based Optimization (Our Approach), using Cross Non-local Means Filters, 41.2 spp, 244.7 sec.**



# TOWN

Reference, 4096 spp

