

Final Presentation

# Turbulence Simulator

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Context: Project A

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In Collaboration with: RAFAEL



# Presentation Outline

- Project goal
- Background
- Optional solution
- Chosen solution
- Results
- Summary
- Conclusions



# Project Goal

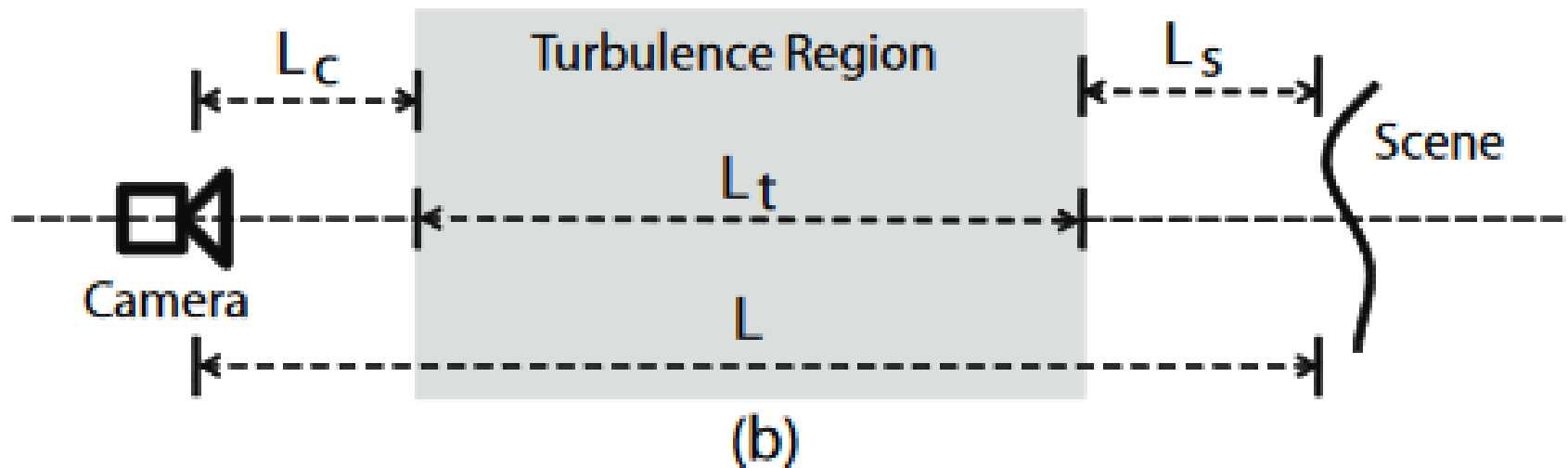
- Simulate distortion of an image caused by atmospheric turbulence using given parameters of user input.
- Create a video from simulated images.
- Create a data base of pictures affected by turbulence for machine learning algorithm.



# Background

- A long-distance imaging system can be strongly affected by atmospheric turbulence.
- random changes of the refractive index along the optical transmission path.
- causes geometric distortion (motion), space and time-varying blur.

# Physical Background





# Physical Background

- Turbulence strength is determined by  $C_n$
- $C_n$  is function of temperature, pressure and more.

- $$\langle \alpha^2 \rangle = 2.914 D^{-\frac{1}{3}} \int_0^L C_n^2(z) \cdot \left( \frac{z}{L} \right)^{\frac{5}{3}} dz$$

- $\alpha$  is the arrival angle.
- $D$  is the camera's aperture.

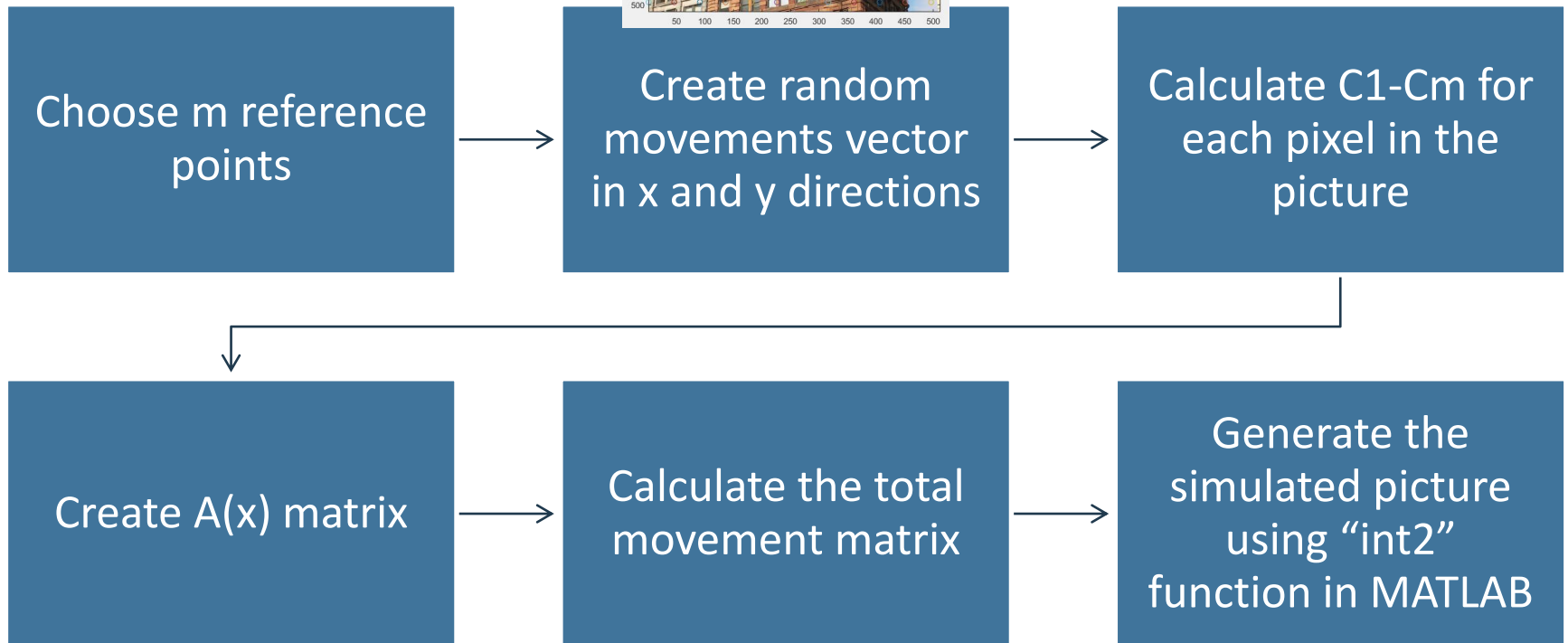


# Optional Solutions

- Random pixel movements
- Milanfar algorithm – statictical and physical

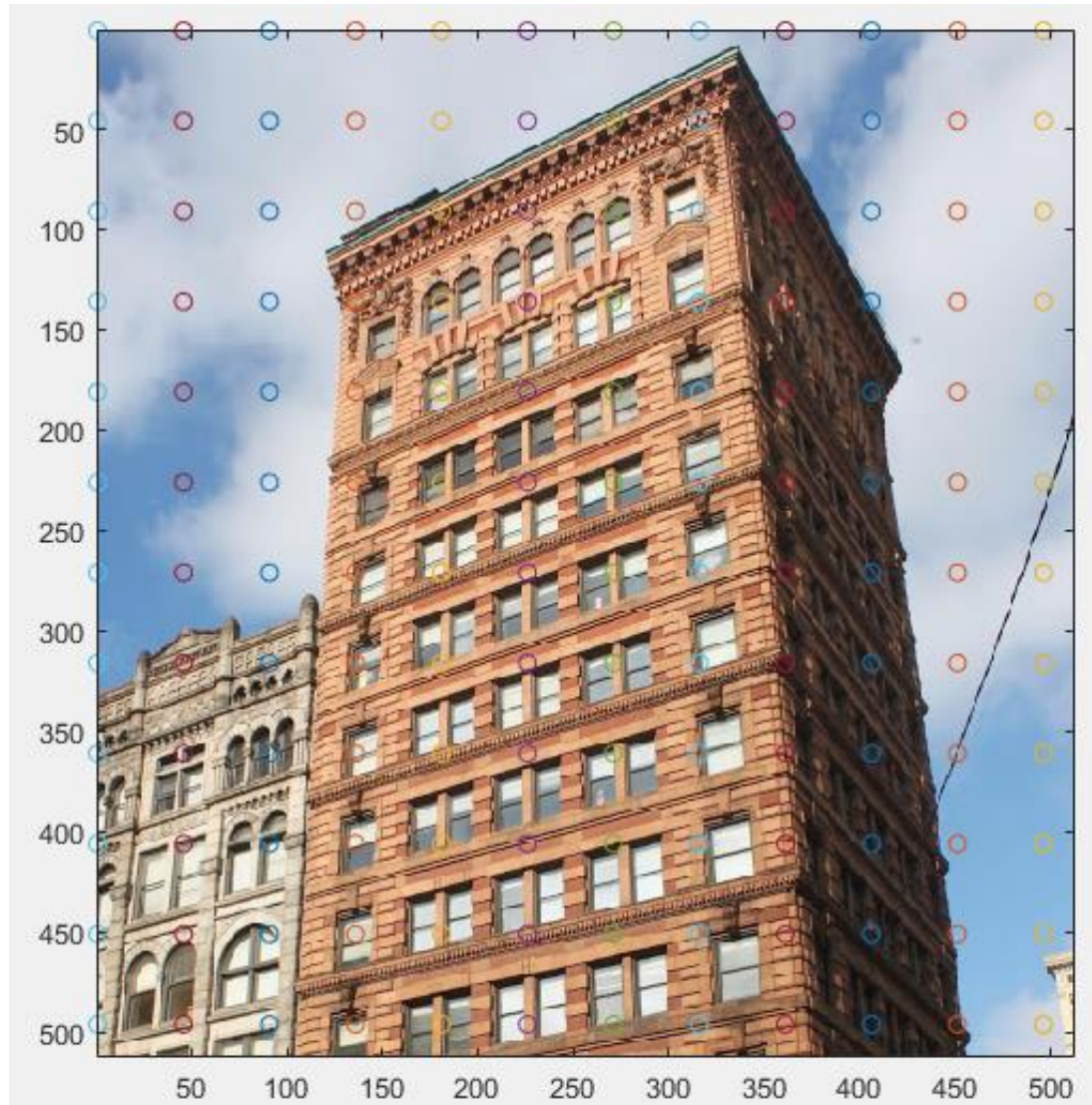


# Milanfar solution: Block Diagram





# Choose $m$ points in a picture



# Milanfar solution: details

- Choose  $m$  points
- Draw movements vector  $\mathbf{p}$
- Calculate  $C_i$  for every pixel and calculate  $A(\mathbf{x})$



$$\mathbf{A}(\mathbf{x}) = \begin{bmatrix} c_1 & \dots & c_m & 0 & \dots & 0 \\ 0 & \dots & 0 & c_1 & \dots & c_m \end{bmatrix}$$

$$c_i = \beta\left(\frac{x - \hat{x}_{0i}}{\epsilon_x}\right) \beta\left(\frac{y - \hat{y}_{0i}}{\epsilon_y}\right),$$

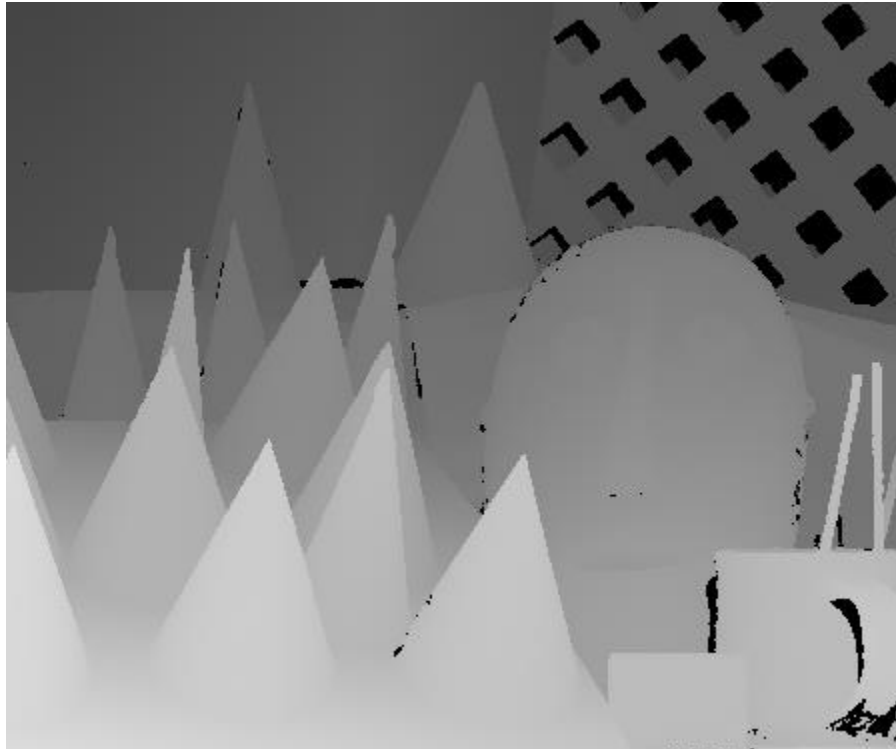
$$\beta(\kappa) = \begin{cases} 2/3 - (1 - |\kappa|/2)\kappa^2, & \text{if } 0 \leq |\kappa| \leq 1 \\ (2 - |\kappa|)^3/6, & \text{if } 1 < |\kappa| < 2 \\ 0, & \text{otherwise,} \end{cases}$$

$$\mathbf{A}(\mathbf{x}) \overrightarrow{\mathbf{p}}$$



# Take distance into consideration

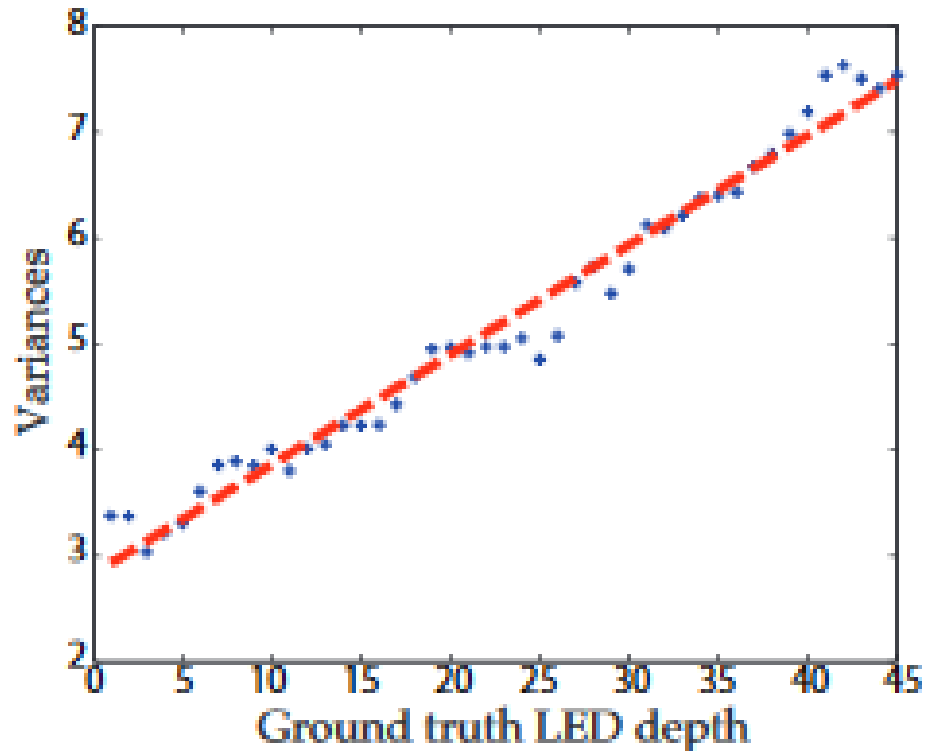
- Using a distance map
- Calculate variance by distance of pixel





# Take distance into consideration

- Based on the article's we read
- Linear connection between depth and var





# Blur caused by turbulence

- The blur can be modeled by a overall optical transfer function (OTF)
- The OTF include the atmospheric OTF and the diffraction OTF.
- The overall OTF is giving by:

$$H(\rho) = H_{\text{atm}}(\rho)H_{\text{dif}}(\rho)$$

$$H_{\text{dif}}(\rho) = \begin{cases} \frac{2}{\pi} \left[ \cos^{-1} \left( \frac{\rho}{2\rho_c} \right) - \frac{\rho}{2\rho_c} \sqrt{1 - \left( \frac{\rho}{2\rho_c} \right)^2} \right], & \rho \leq \rho_c \\ 0, & \text{otherwise} \end{cases},$$

$$H_{\text{atm}}(\rho) = \exp \left\{ -3.44 \left( \frac{\lambda \rho}{r_0} \right)^{5/3} \left[ 1 - \alpha \left( \frac{\lambda \rho}{D} \right)^{1/3} \right] \right\},$$



# Blur caused by turbulence

- The overall OTF is giving by:

$$H(\rho) = H_{\text{atm}}(\rho)H_{\text{dif}}(\rho)$$

- The atmospheric OTF is giving by:

$$H_{\text{atm}}(\rho) = \exp\left\{-3.44\left(\frac{\lambda l \rho}{r_0}\right)^{5/3}\left[1 - \alpha\left(\frac{\lambda l \rho}{D}\right)^{1/3}\right]\right\},$$

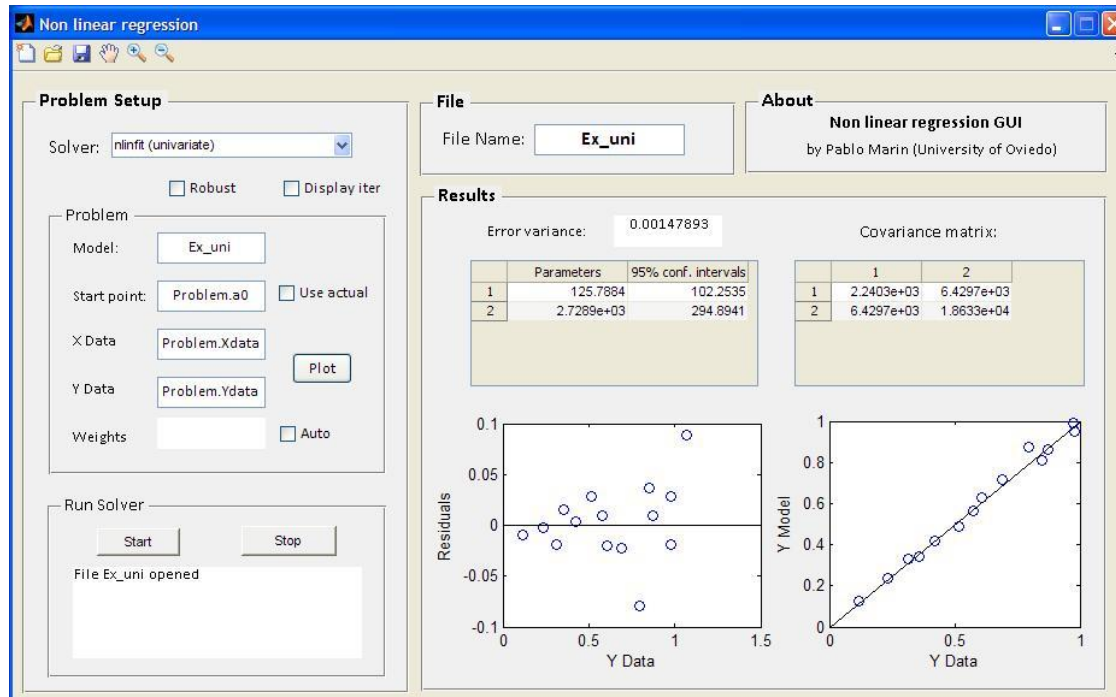
- The diffraction OTF is giving by:

$$H_{\text{dif}}(\rho) = \begin{cases} \frac{2}{\pi} \left[ \cos^{-1}\left(\frac{\rho}{2\rho_c}\right) - \frac{\rho}{2\rho_c} \sqrt{1 - \left(\frac{\rho}{2\rho_c}\right)^2} \right], & \rho \leq \rho_c \\ 0, & \text{otherwise} \end{cases},$$



# GUI

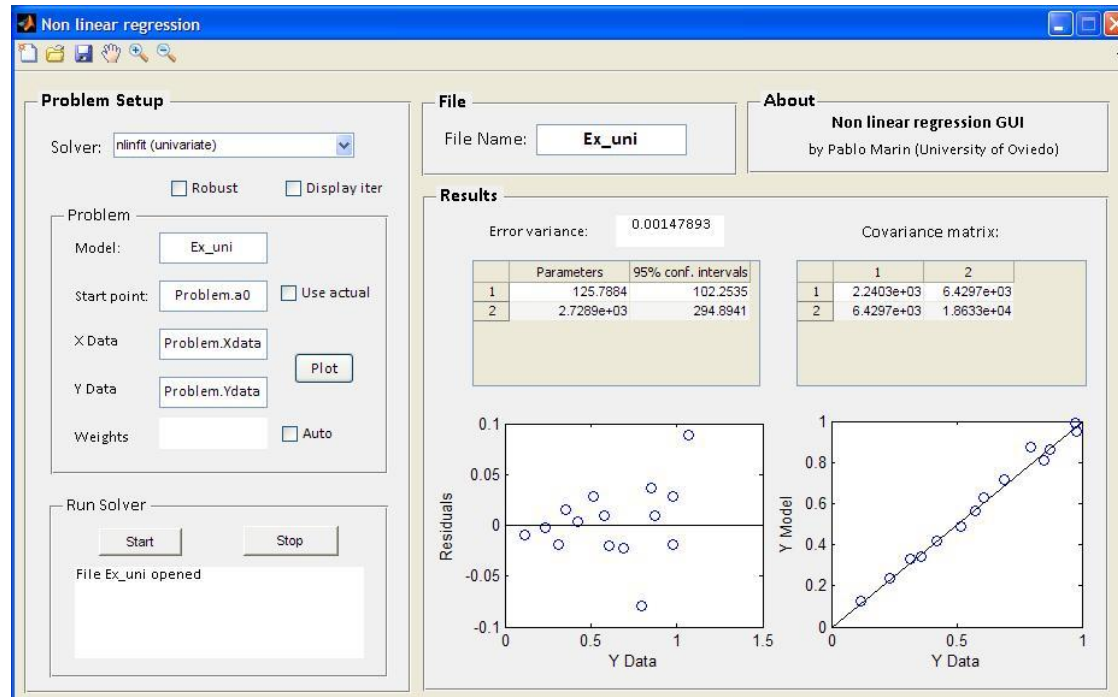
- We created simulator using GUI.
- Simulator tool for the user.
- Fast and convenient. (GUI VIDEO)





# Problem with Time complexity

- The simulation took about 6 minutes.
- After some improvements – 4 seconds.
- 0.01% of the initial time.







# Creating Video

- Considering all of the previous we created a video.
- Comparing this video to real turbulence video provided by Rafael.



# Time smoothing

- In order to create more "Realistic" videos we need to activate time smoothing on the video
- We used mean filter on the frames of the video

# Results

## distance effect – original Image





# Results – distance effect (80m)

After -  $C_n^2 = 5e-16$ ,  $f=1.2[m]$ ,  $D=0.2[m]$  Dist=80[m]



# Results – distance effect (330m)

After -  $C_n^2 = 5e-16$ ,  $f=1.2[m]$ ,  $D=0.2[m]$  Dist=330[m]







# Results – distance effect (580m)

After -  $C_n^2 = 5e-16$ ,  $f = 1.2[m]$ ,  $D = 0.2[m]$  Dist = 580[m]





# Results – distance effect (830m)

After -  $C_n^2 = 5e-16$ ,  $f=1.2[m]$ ,  $D=0.2[m]$  Dist=830[m]





# Results – distance effect (980m)

After -  $C_n^2 = 5e-16$ ,  $f=1.2[m]$ ,  $D=0.2[m]$  Dist=980[m]





# Results – distance effect (1590)

After -  $C_n^2 = 5e-16$ ,  $f = 1.2[m]$ ,  $D = 0.2[m]$  Dist = 1590[m]





# Results – Cn effect original image



# Results – Cn effect –(1e-16)

After -  $C_n^2 = 1e-16$ ,  $f = 1.2[m]$ ,  $D = 0.2[m]$



# Results – Cn effect –(2e-16)

After -  $C_n^2 = 2e-16$ ,  $f=1.2[m]$ ,  $D=0.2[m]$





# Results – Cn effect –(3e-16)

After -  $C_n^2 = 3e-16$ ,  $f=1.2[m]$ ,  $D=0.2[m]$



# Results – Cn effect –(4e-16)

After -  $C_n^2 = 4e-16$ ,  $f=1.2[m]$ ,  $D=0.2[m]$



# Results – Cn effect –(5e-16)

After -  $C_n^2 = 5e-16$ ,  $f = 1.2[m]$ ,  $D = 0.2[m]$





# Checking goodness

- Video from Raphael.
- Avg on video to get clear frame.
- Distort using our simulation.





# Rafael Video – dis 1500m



# Simulated video





# Quantitative Comparison

- Using sophisticated algorithm for tracking pixels.
- Calculating var of pixel movements.
- Var for real video :  $\text{var.x} = 0.301$   $\text{var.y} = 0.441$
- Var for simulate video :  
 $\text{var.x} = 0.175$   $\text{var.y} = 0.233$
- Results are fantastic!



# Summary

- We have managed to apply “statistic” turbulence.
- We will apply a physical model in the future.
- We will create a GUI.
- If time allows – we’ll make the system capable of handling with videos and apply blur.



# Conclusions

- We have learned that as the number of reference points grows up (Milanfar), the image distortion level grows.
- Milanfar solution is a good statistical solution.
- In the future – simulate more physical distortion for better results.



# References

- [1] Endre Repasi and Robert Weiss ,  
“Computer Simulation of Image Degradations  
by Atmospheric Turbulence for Horizontal  
Views” , 2011
- [2] Peyman Milanfar , “Removing Atmospheric  
Turbulence via Space-Invariant  
Deconvolution” , 2013