

CSCI 497P/597P: Computer Vision

Lecture 25

Epipolar Geometry
Structure From Motion
Multiview Stereo

Announcements

- P3 is out!
- Deadline for grads to opt in for P/NP is Friday
 - Undergrads have until June 5th

Goals

- Understand some of the properties of the fundamental matrix:
 - rank deficiency
 - epipolar lines; epipoles
- Understand the general idea of how Structure From Motion is solved.

(a tiny bit more whiteboard)

8-point algorithm

$$\mathbf{x}'^T \mathbf{F} \mathbf{x} = 0$$

$$\mathbf{F} = \begin{bmatrix} f_{11} & f_{12} & f_{13} \\ f_{21} & f_{22} & f_{23} \\ f_{31} & f_{32} & f_{33} \end{bmatrix}$$

Let $\mathbf{x} = (u, v, 1)^T$ and $\mathbf{x}' = (u', v', 1)^T$,

Each match yields **one** equation:

$$uu' f_{11} + vu' f_{12} + u' f_{13} + uv' f_{21} + vv' f_{22} + v' f_{23} + uf_{31} + vf_{32} + f_{33} = 0$$

8-point algorithm

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Let $\mathbf{x}=(u, v, 1)^T$ and $\mathbf{x}'=(u', v', 1)^T$,

Each $uu' f_{11} + vv' f_{12} + u' f_{13} + uv' f_{21} + vv' f_{22} + v' f_{23} + uf_{31} + vf_{32} + f_{33} = 0$ match yields equation:

$$uu' f_{11} + vv' f_{12} + u' f_{13} + uv' f_{21} + vv' f_{22} + v' f_{23} + uf_{31} + vf_{32} + f_{33} = 0$$

8-point algorithm

$$\begin{array}{l} \downarrow \\ \left[\begin{array}{cccccccc|c} u_1 u_1' & v_1 u_1' & u_1' & u_1 v_1' & v_1 v_1' & v_1' & u_1 & v_1 & 1 \\ u_2 u_2' & v_2 u_2' & u_2' & u_2 v_2' & v_2 v_2' & v_2' & u_2 & v_2 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ u_n u_n' & v_n u_n' & u_n' & u_n v_n' & v_n v_n' & v_n' & u_n & v_n & 1 \end{array} \right] \begin{array}{c} f_{11} \\ f_{12} \\ f_{13} \\ f_{21} \\ f_{22} \\ f_{23} \\ f_{31} \\ f_{32} \\ f_{33} \end{array} = 0 \end{array}$$

As with homographies, this is has the form $Ax = 0$
 Solve homogeneous system using the SVD.

8-point algorithm: Problem

- **Solution is (generally) not rank 2.**

- Fix: More SVD!


The image shows two handwritten equations for Singular Value Decomposition (SVD). The top equation is $\begin{bmatrix} T \\ S \end{bmatrix} = \begin{bmatrix} U \\ \Sigma \\ V^T \end{bmatrix}$, where Σ contains σ_1 , σ_2 , and σ_3 (circled). An arrow labeled "unitary" points to the U matrix. The bottom equation is $\begin{bmatrix} F \end{bmatrix} = \begin{bmatrix} U \\ \Sigma \\ 0 \\ V^T \end{bmatrix}$, where Σ contains σ_1 and σ_2 , and there is a zero below σ_2 . An arrow labeled "unitary" points to the U matrix. A curved arrow labeled "rank 2" points from the top equation to the bottom equation.

8-point algorithm

- **Solution is (generally) not rank 2.**
- Fix: More SVD!

8-point algorithm: Problem 2

$$\begin{bmatrix}
 u_1 u_1' & v_1 u_1' & u_1' & u_1 v_1' & v_1 v_1' & v_1' & u_1 & v_1 & 1 \\
 u_2 u_2' & v_2 u_2' & u_2' & u_2 v_2' & v_2 v_2' & v_2' & u_2 & v_2 & 1 \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\
 u_n u_n' & v_n u_n' & u_n' & u_n v_n' & v_n v_n' & v_n' & u_n & v_n & 1
 \end{bmatrix}
 \begin{bmatrix}
 f_{11} \\
 f_{12} \\
 f_{13} \\
 f_{21} \\
 f_{22} \\
 f_{23} \\
 f_{31} \\
 f_{32} \\
 f_{33}
 \end{bmatrix}
 = 0$$

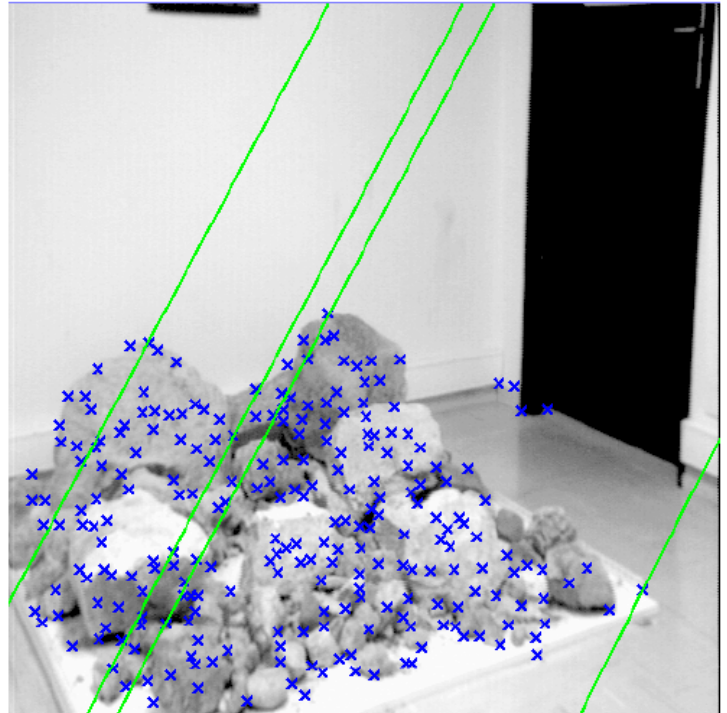
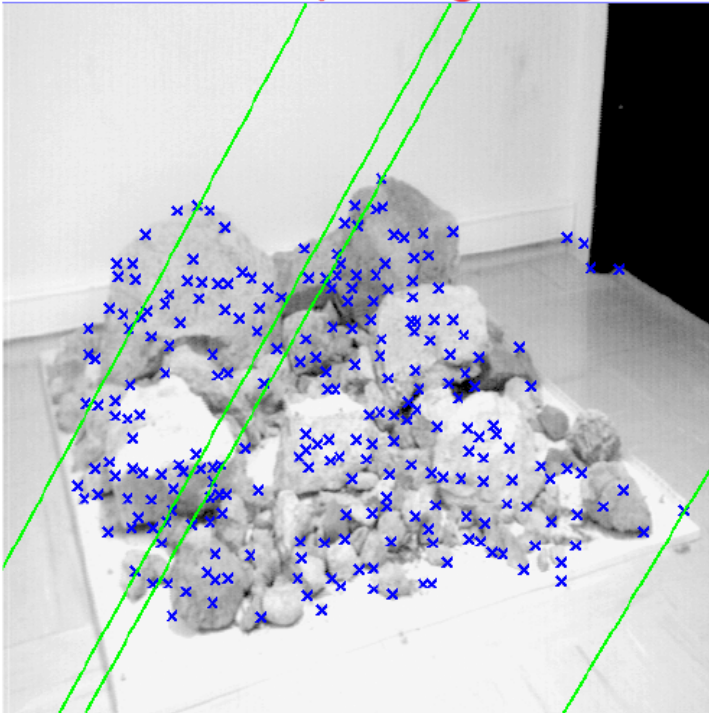

Orders of magnitude difference between column of data matrix → least-squares yields poor results

~10000 ~10000 ~100 ~10000 ~10000 ~100 ~100 ~100 1

Fix: scale image positions to the range [0,1], solve, then scale back.

8-point algorithm: Results

■ Normalized 8-point algorithm



What about more than 2 views?

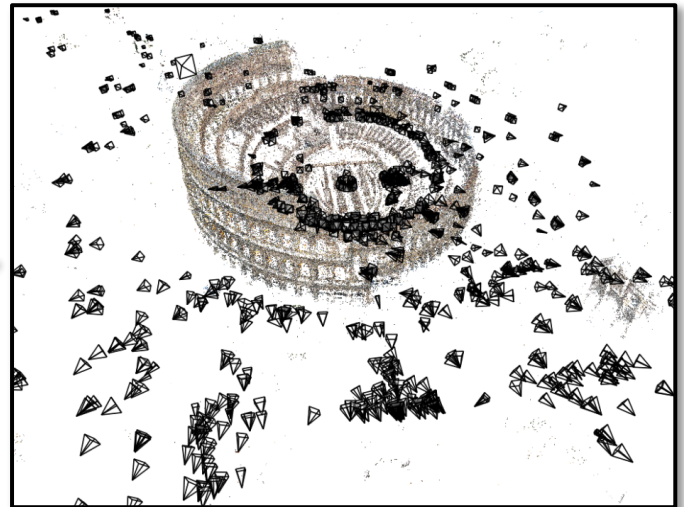
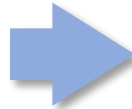
- 2 views: fundamental matrix
- 3 views: trifocal tensor
- 4 views: *quadrifocal* tensor
- more views: $\backslash_{(\Psi)}_{/}$ (it gets complicated...)

Large-scale structure from motion

- <https://www.youtube.com/watch?v=sQegEro5Bfo>

Structure from Motion

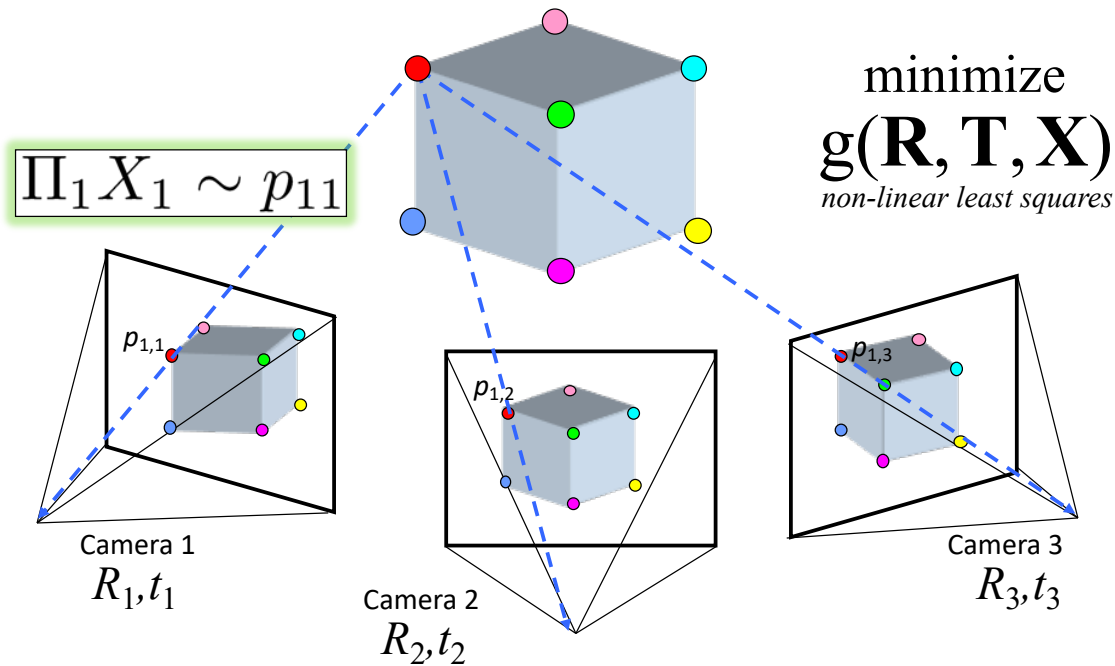
- Given many photos, reconstruct:
 - positions of the cameras
 - positions of 3D points



Chicken/Egg

- Step 1: solve for relative pose of pairs (or triples) of cameras using correspondences from feature matching.
- Step 2: alternate between solving:
 - given camera positions, solve for point locations
 - given point locations, solve for camera positions

Structure From Motion



$$g(\mathbf{X}, \mathbf{R}, \mathbf{T}) = \sum_{i=1}^m \sum_{j=1}^n w_{ij} \cdot \left\| \underbrace{\mathbf{P}(\mathbf{x}_i, \mathbf{R}_j, \mathbf{t}_j)}_{\text{predicted image location}} - \underbrace{\begin{bmatrix} u_{i,j} \\ v_{i,j} \end{bmatrix}}_{\text{observed image location}} \right\|^2$$

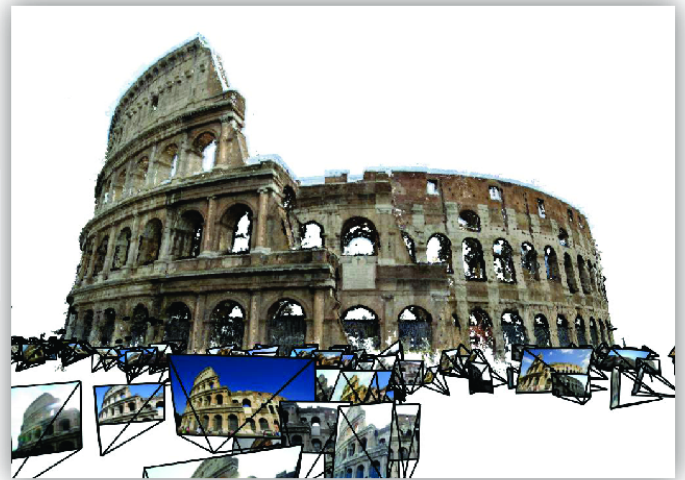
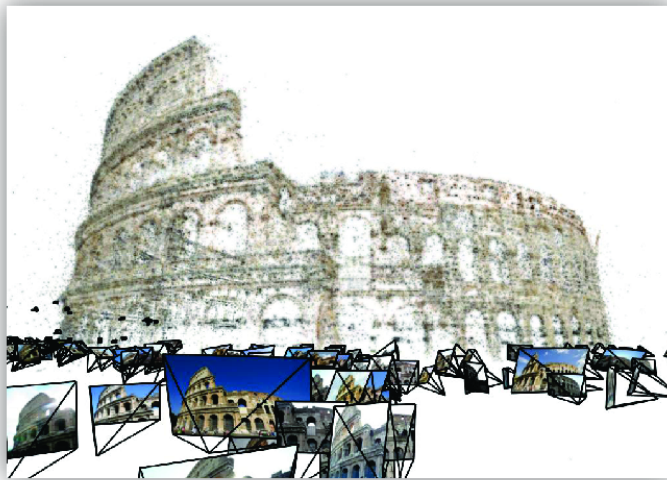
\downarrow
 indicator variable:
 is point i visible in image j ?

Applications

- Hyperlapse <https://www.youtube.com/watch?v=SOpwHaQnRSY>
- SLAM: <https://medium.com/scape-technologies/building-the-ar-cloud-part-three-3d-maps-the-digital-scaffolding-of-the-21st-century-465fa55782dd>
- Graphics, movies, games, self-driving cars, robots, ...

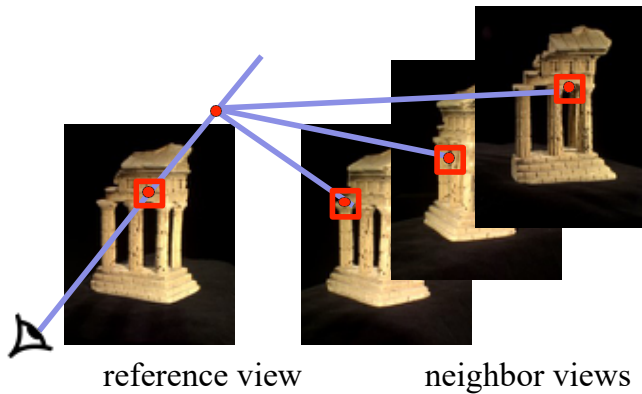
Multiview Stereo

- Once you've solved for all those camera positions, how good a 3D model can you create?

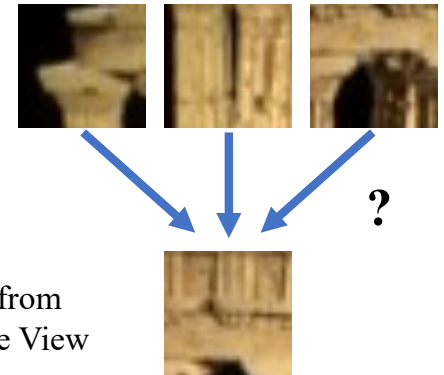


Multiview Stereo: Basic Idea

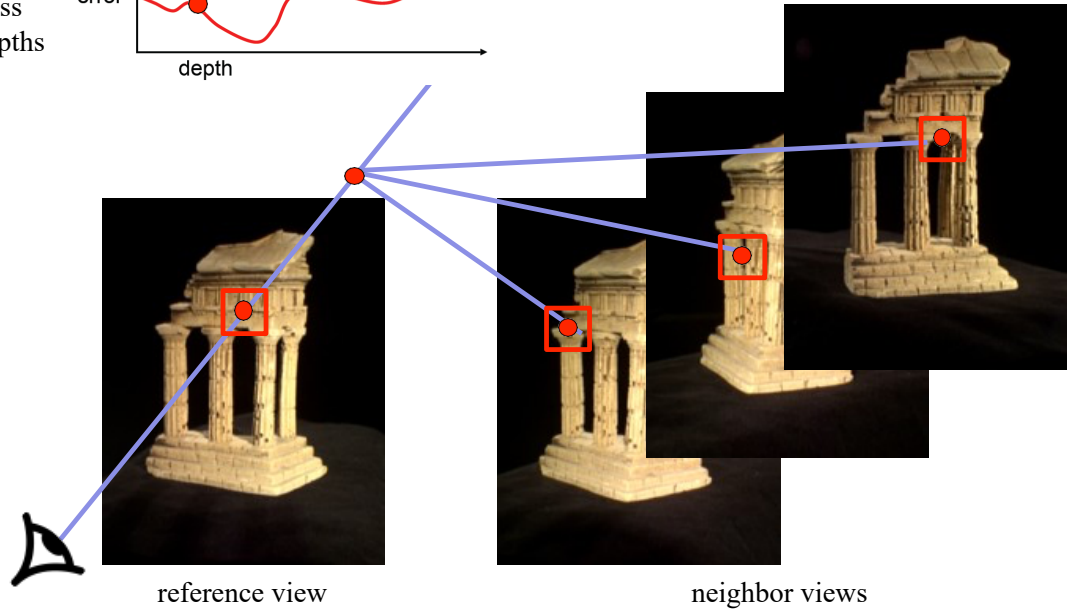
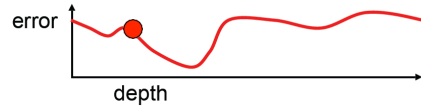
Evaluate the likelihood of geometry at a particular depth for a particular reference patch:



Corresponding patches at depth guess in other views

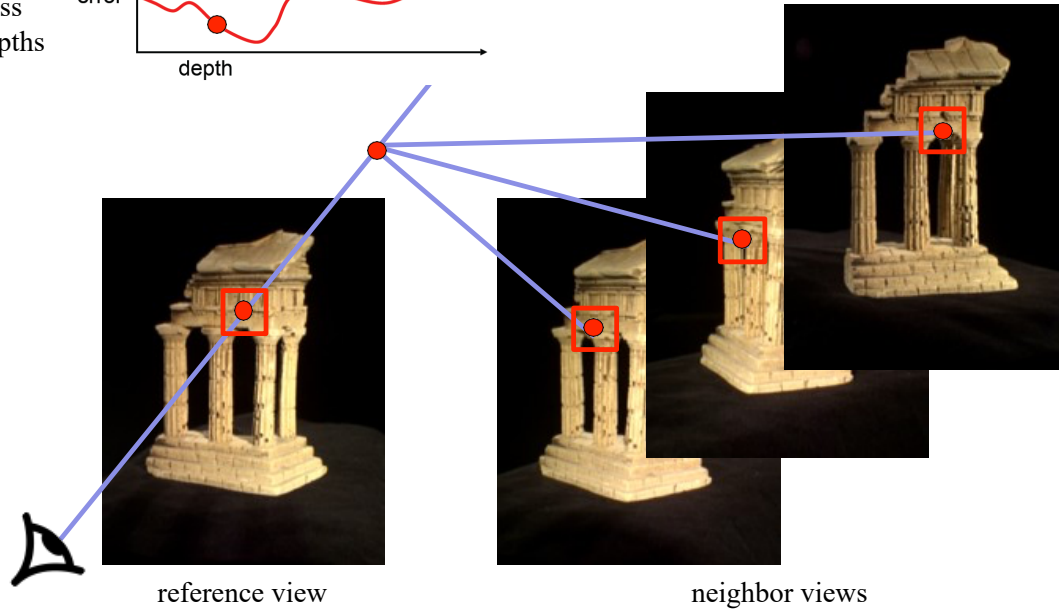
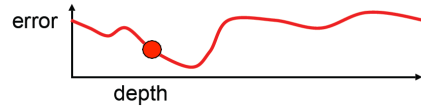


Photometric error across different depths



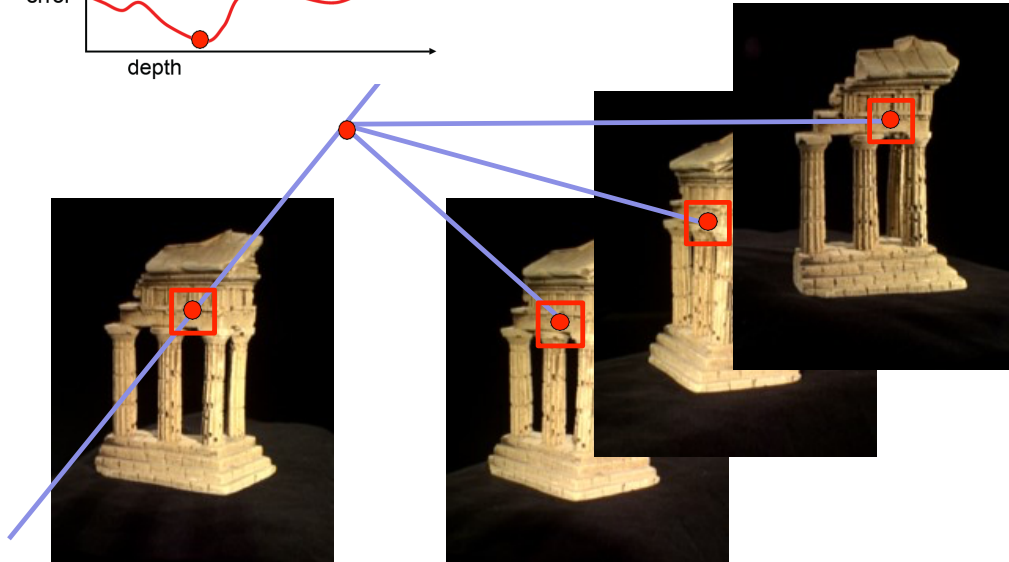
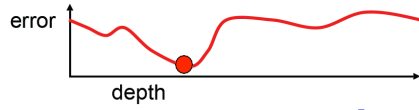
Source: Y. Furukawa

Photometric error across different depths



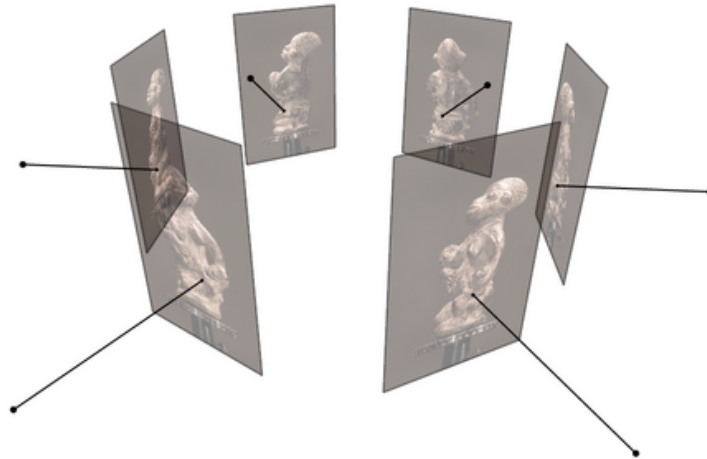
Source: Y. Furukawa

Photometric error across different depths



Depth map fusion

- Compute depth maps for multiple cameras, then fuse them into a 3D model



Figures by Carlos Hernandez

Result

- <https://www.youtube.com/watch?v=N6Douyfa7I8>
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