

6.866

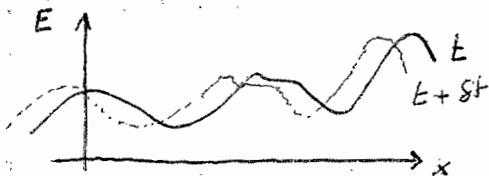
Quiz #1 Oct. 19

Quiz #2 Nov. 30

Yajun Fang

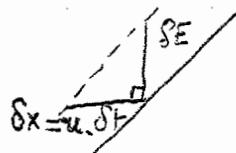
① Estimating motion

1st example, 1 vector (x, y) , 1 variable $E = \text{brightness}$
 $x = \text{pos. on the image}$



shift + meas. noise

$u = \text{velocity} = \frac{\delta x}{\delta t}$

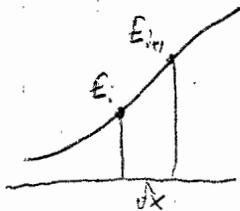


slope = $\frac{dE}{dx}$

$\frac{dE}{dx} = -\frac{\delta E}{u \delta t}$

$u \cdot E_x + E_t = 0$

or: $u \frac{dE}{dx} + \frac{\delta E}{\delta t} = 0$



$E_x \approx \frac{E_{i+1} - E_i}{\delta x}$ and $E_t \approx \frac{E_{i+1} - E_i}{\delta t}$

thus $u = -\frac{E_t}{E_x}$

but image noisy so poor method

and what if $E_t = 0$? (blank area)

and what if $E_x \approx 0$? not good either

Better method:

$u \approx \frac{-1}{x_2 - x_1} \int_{x_1}^{x_2} \frac{E_x}{E_t} dx + \text{filtering } E_t \neq 0$

$$u \approx \frac{-1}{x_2 - x_1} \int_{x_1}^{x_2} w(x) \frac{E_x}{E_t} dx / W \quad \text{where } W = \frac{1}{x_2 - x_1} \int_{x_1}^{x_2} w(x) dx$$

so $u = - \int_{x_1}^{x_2} w(x) \frac{E_x}{E_t} dx / \int_{x_1}^{x_2} w(x) dx$

let's take $w(x) = E_t$

$$u = \frac{\int_{x_1}^{x_2} E_x dx}{-E_2 - E_1} / \int_{x_1}^{x_2} E_t dx \quad \text{not good! } \%?$$

but what about $w(x) = |E_t|$

$$u = - \int_{x_1}^{x_2} E_x \text{sign}(E_t) dx / \int_{x_1}^{x_2} |E_t| dx \quad \text{not optimal... discontinuities}$$

so $w(x) = E_t^2$

$$u = - \int_{x_1}^{x_2} E_x E_t dx / \int_{x_1}^{x_2} E_t^2 dx$$

⊕ takes info from all the image

Another way:

$$u E_x + E_t = 0 \quad \text{constraint equation}$$

$$\int_{x_1}^{x_2} (u E_x + E_t)^2 dx \begin{cases} \rightarrow 0 & \text{perfect case} \\ \rightarrow \varepsilon & \text{error real case} \end{cases}$$

$u = \text{min value for which } \varepsilon$

$$\frac{d}{du} (\varepsilon) = 0 \quad \Rightarrow \quad \int_{x_1}^{x_2} 2E_x (uE_x + E_t) E_x dx = 0$$

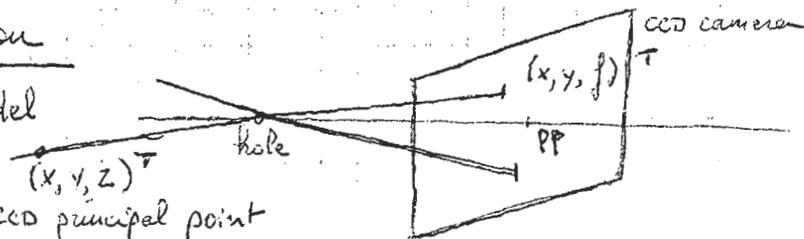
$$u \int_{x_1}^{x_2} E_x^2 dx + \int_{x_1}^{x_2} E_x E_t dx = 0$$

$$u = - \frac{\int_{x_1}^{x_2} E_x E_t dx}{\int_{x_1}^{x_2} E_x^2 dx}$$

Least Square Solution
"optical flow"

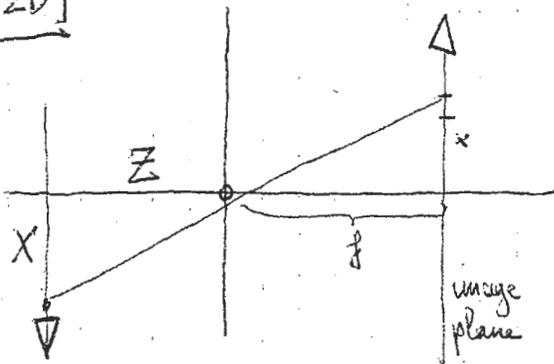
if motion = constant then average on time also
→ more accurate

② Image projection
pinhole model



$f = \text{dis. bet. hole \& CCD principal point}$

2D



3D

$$\frac{x}{f} = \frac{X}{Z}$$

$$\frac{y}{f} = \frac{Y}{Z}$$

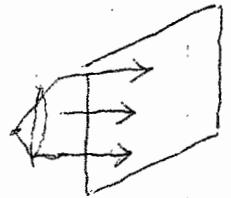
perspective projection equation

$f = \text{zoom factor}$
+ scale ambiguity

H] z constant ($z = \text{optical axis}$)

$$\frac{x}{f} = \frac{X}{Z_0} \Rightarrow x = \frac{f}{Z_0} \cdot X \quad y = \frac{f}{Z_0} Y.$$

case $f = Z_0$: $x = X, y = Y$ orthographic projection
 good approx. for telephoto lenses.

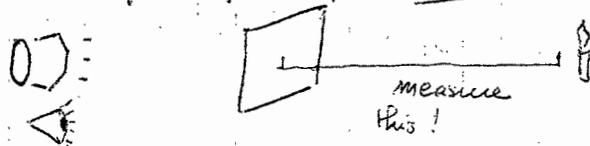


③ Brightness

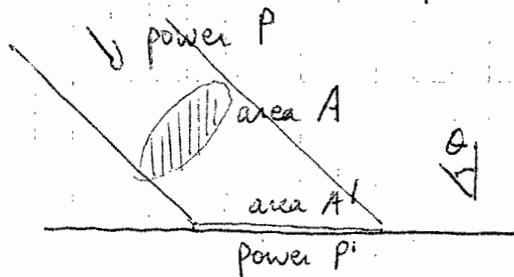
- Parameters :
- (i) material of surface
 - (ii) illumination : power & distribution
 - (iii) viewer direction

Model of a particular type of surface : "Lambertian" (ideal)

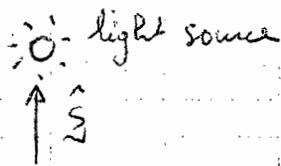
Lambert:



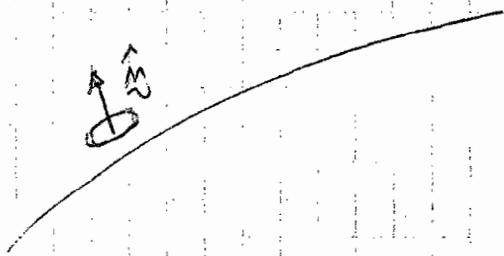
- (i) reflects all incident light
- (ii) appears equally bright from all viewpoints (mat)



$$\frac{P'}{A'} = \frac{P}{A} \cos \theta$$



Power received = $P \times \hat{n} \cdot \hat{s}$
 where $\hat{n} \cdot \hat{s} = \|\hat{n}\| \cdot \|\hat{s}\| \cdot \cos \theta$

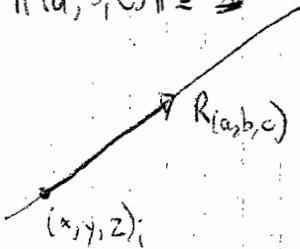


1. A line in 3D becomes a line in 2D camera
2. Two parallel lines in 3D are generally not parallel in 2D.
3. The common point is the vanishing point.

Equation of a line:

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x_i \\ y_i \\ z_i \end{pmatrix} + \begin{pmatrix} R_a \\ R_b \\ R_c \end{pmatrix}$$

$$\|(a, b, c)\| = 1$$



$$\begin{cases} \frac{x}{f} = \frac{X}{Z} = \frac{x_i + R_a}{z_i + R_c} \\ \frac{y}{f} = \frac{Y}{Z} = \frac{y_i + R_b}{z_i + R_c} \end{cases}$$

if $R \rightarrow \infty$: $\frac{x}{f} \rightarrow \frac{a}{c}$ $\frac{y}{f} \rightarrow \frac{b}{c}$

vanishing point

