

# 2. Probability and Statistical Inference

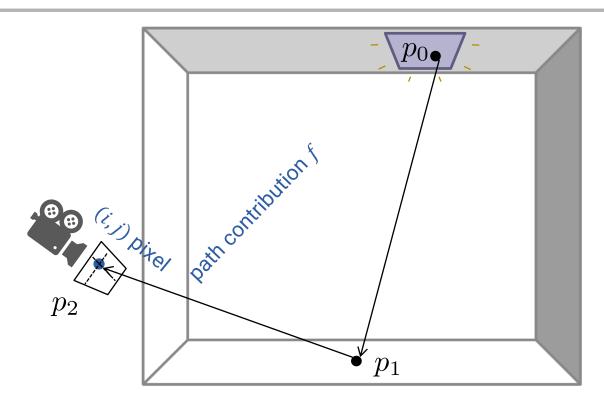
Physically Based Rendering

Shinyoung Yi (이신영)



# Preview: a simple path tracing





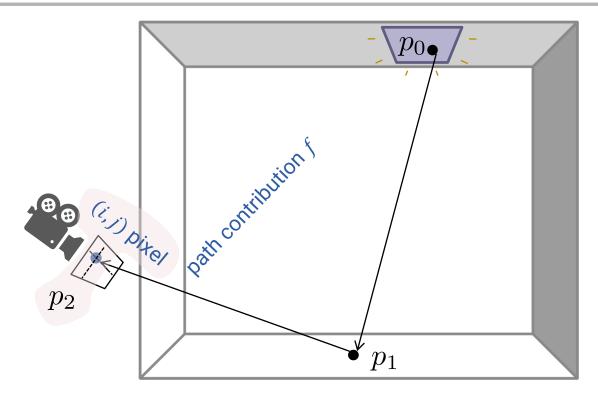
#### **Terminology and convention:**

- Notation / actual light:  $p_0 \rightarrow p_1 \rightarrow p_2$
- Computation:  $p_2 \rightarrow p_1 \rightarrow p_0$
- # of bounces < depth < # of vertices</li>

2

3



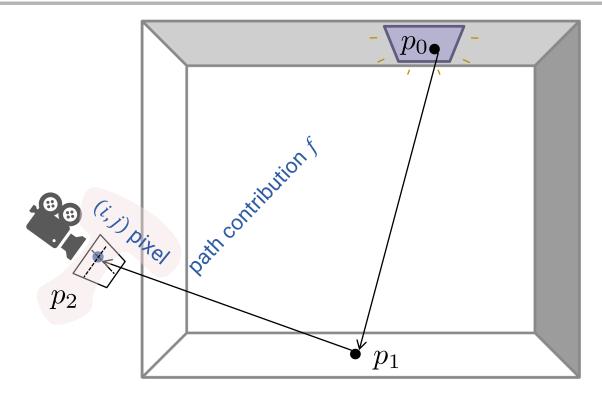


#### **Notation:**

$$\bullet \quad \widehat{\omega}_{p_0 p_1} \coloneqq \frac{p_1 - p_0}{\|p_1 - p_0\|}$$

$$I^{(i,j)} = \int_{\mathbb{S}} W_e^{(i,j)} (\hat{\omega}) L(p_2, \hat{\omega}) d\hat{\omega}$$



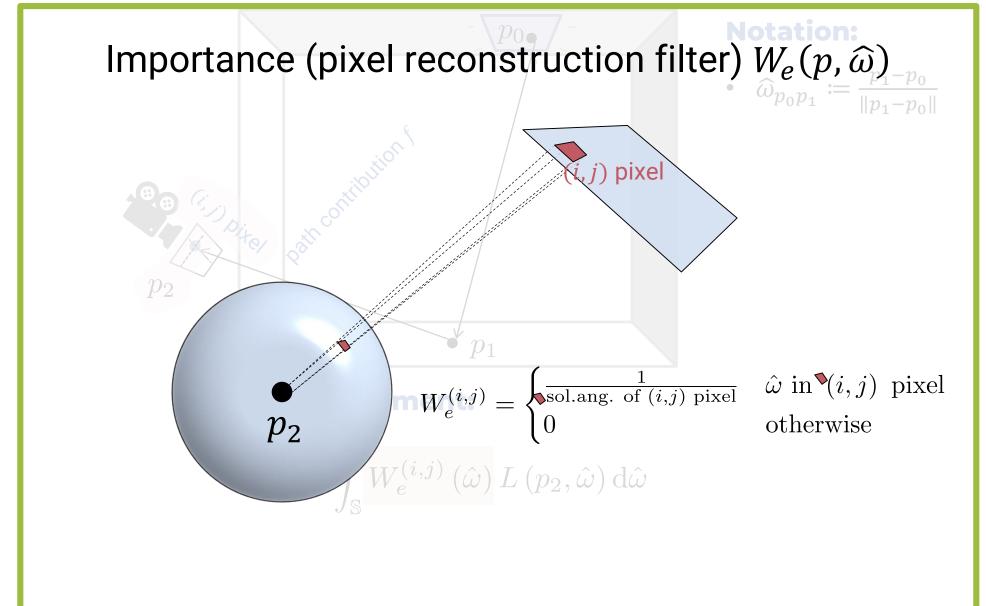


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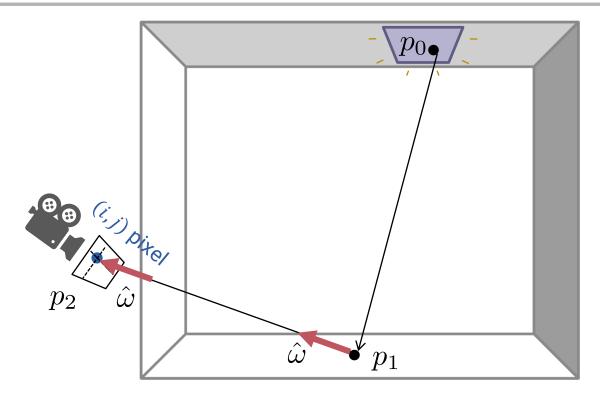
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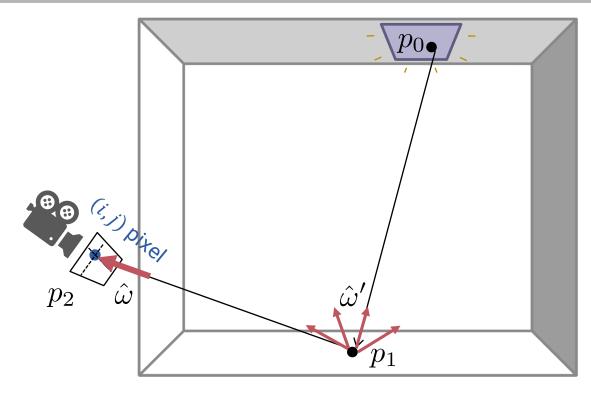
- $\bullet \quad \widehat{\omega}_{p_0 p_1} \coloneqq \frac{p_1 p_0}{\|p_1 p_0\|}$
- $r: \mathbb{R}^3 \times \mathbb{S}^2 \to \mathbb{R}^3$ 
  - ray intersection

$$I^{(i,j)} = \int_{\mathbb{S}} W_e^{(i,j)}(\hat{\omega}) L(p_2, \hat{\omega}) d\hat{\omega}$$

$$L(p_2, \hat{\omega}) = L(p_1, \hat{\omega})$$

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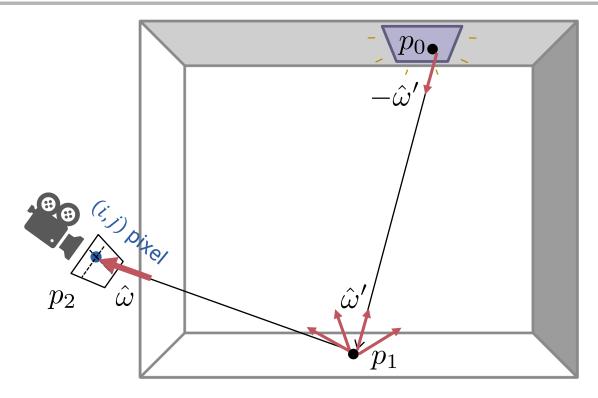
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$$I^{(i,j)} = \int_{\mathbb{S}} W_e^{(i,j)}(\hat{\omega}) L(p_1, \hat{\omega}) d\hat{\omega}$$

$$L(p_1, \hat{\omega}) = L_e(p_1, \hat{\omega}) + \int_{\mathbb{S}} L^{(\text{in})}(p_1, \hat{\omega}') \rho(p_1, \hat{\omega}', \hat{\omega}) |\hat{n} \cdot \hat{\omega}'| d\hat{\omega}'$$





#### **Notation:**

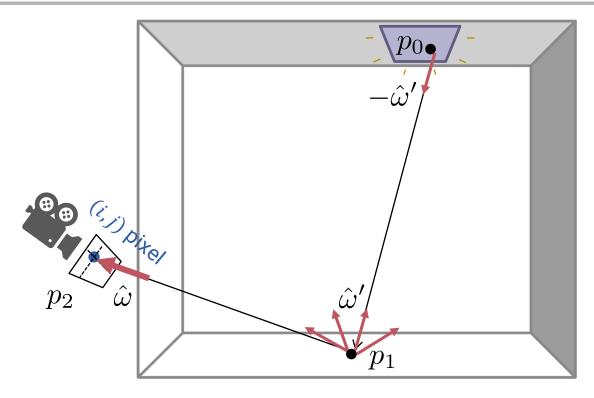
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$$L^{(\text{in})}(p_1, \hat{\omega}') = L(p_0 - \hat{\omega}')$$





#### **Notation:**

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- $r: \mathbb{R}^3 \times \mathbb{S}^2 \to \mathbb{R}^3$ 
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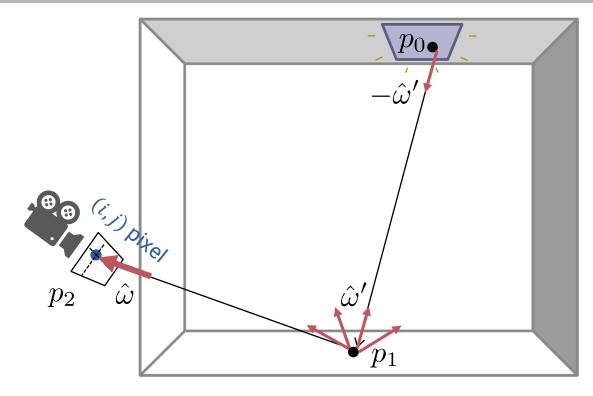
$$I^{(i,j)} = \int_{\mathbb{S}} W_e^{(i,j)}\left(\hat{\omega}\right) L\left(p_1,\hat{\omega}\right) d\hat{\omega}$$

$$L\left(p_1,\hat{\omega}\right) = L_e\left(p_1,\hat{\omega}\right) + \int_{\mathbb{S}} L^{(\mathrm{in})}\left(p_1,\hat{\omega}'\right) \rho\left(p_1,\hat{\omega}',\hat{\omega}\right) \left|\hat{n}\cdot\hat{\omega}'\right| d\hat{\omega}'$$

$$L\left(p_0,-\hat{\omega}'\right) = L_e\left(p_0,-\hat{\omega}'\right) + \int_{\mathbb{S}} L^{(\mathrm{in})}\left(p_0,\hat{\omega}''\right) \rho\left(p_0,\hat{\omega}'',-\hat{\omega}'\right) \left|\hat{n}\cdot\hat{\omega}''\right| d\hat{\omega}''$$

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#### **Notation:**

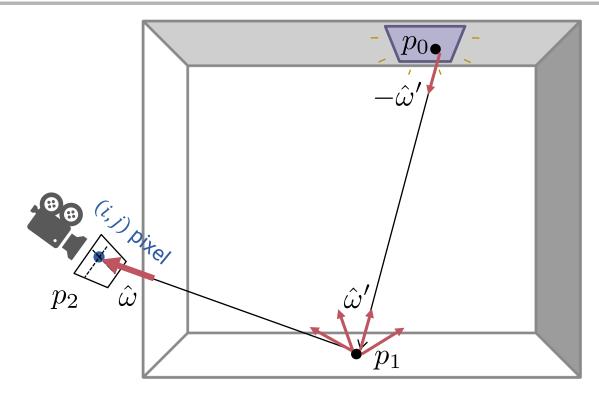
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$$L(p_0, -\hat{\omega}') = L_e(p_0, -\hat{\omega}')$$





#### **Notation:**

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- $r: \mathbb{R}^3 \times \mathbb{S}^2 \to \mathbb{R}^3$ 
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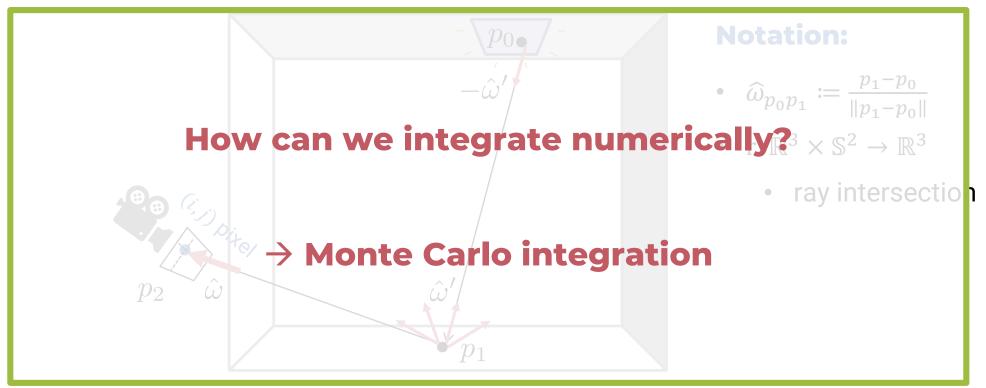
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$$p_1 = r(p_2, -\hat{\omega})$$

$$p_0 = r(p_1, \hat{\omega}')$$





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# **Basic Math: Sets and Functions**

#### Sets and functions



Suppose  $A, B, X, \cdots$  are sets and  $f, g, \cdots$  are functions

- $f: X \to Y$ ,  $A \subset X$  then  $f(A) := \{f(x) | x \in A\}$  is the **image of** A **under** f
- $B \subset Y$  then  $f^{-1}(B) := \{x | f(x) \in B\}$  is the **preimage of** B **under** f
  - Even if an inverse function does not exists, preimages are well defined.



# Probability

# What is probability?

## random sample

probability

observation

distribution

random variable

### Probability overview



- Sample space (probability space)  $\Omega$
- Event *E*
- Probability P

- Random variable X
- CDF of *X*
- PDF of *X*

### **Probability overview**



- Sample space (probability space)  $\Omega$ : a set of all things that the outcome can be.
- Event  $E \subset \Omega$
- Probability  $P(E) \in \mathbb{R}_{\geq 0}$ 
  - $P(E_1 \cup E_2) = P(E_1) + P(E_2) \text{ if } E_1 \cap E_2 = \phi$
  - $-P(\Omega)=1$
- Random variable  $X: \Omega \to \mathbb{R}$  (codomain  $\mathbb{R}^n$  or  $\mathbb{S}^2$  is also okay)
- CDF of  $X: F_X(x) = P(X \le x) := P(X^{-1}(\{y \in \mathbb{R} | y \le x\}))$
- PDF of X: some function  $p_X: \mathbb{R} \to \mathbb{R}_{\geq 0}$  s.t.  $P(a \leq x \leq b) = \int_a^b p_X(x) dx$ 
  - $-p_X(x) = \frac{\mathrm{d}F_X(x)}{\mathrm{d}x}$  for continuous random variable

# Why is probability so confusing?





#### Choice 1

Sample space:  $\{h, t\}$ 

$$H = \{h\}, T = \{t\}$$

$$P[{h}] = \frac{1}{2}, P[{t}] = \frac{1}{2}$$

#### Choice 2

Sample space:  $\{\theta | 0 \le \theta \le 2\pi\}$ 

$$H = \{\theta | 0 \le \theta \le \pi\}, T = \{\theta | \pi < \theta \le 2\pi\}$$

$$P[\{\theta | 0 \le \theta_1 < \theta \le \theta_2 \le 2\pi\}] = \frac{\theta_2 - \theta_1}{2\pi}$$

Event: *H*, *T* 

Choice 3

Probability: 
$$P[H] = \frac{1}{2}$$
,  $P[T] = \frac{1}{2}$ 

Sample space:  $\{\mathbf{R} | \mathbf{R} \in SO(3)\}$ 

$$H = \left\{ \mathbf{R} \middle| 0 \le \mathbf{R}\hat{z} \le \frac{\pi}{2} \right\}, T = \left\{ \mathbf{R}\hat{z} \middle| \frac{\pi}{2} < \mathbf{R}\hat{z} \le 1 \right\}$$

$$P[A \subset SO(3)] = \frac{1}{8\pi^2} \int_A \sin\beta \, d\alpha d\beta d\gamma$$



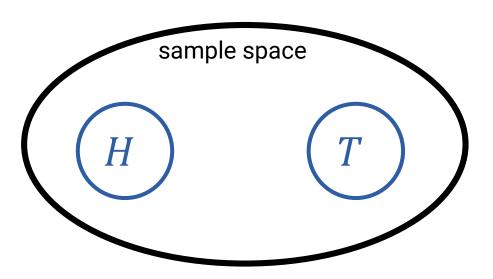


Choice 1

Sample space: {h, t}

$$H = \{h\}, T = \{t\}$$

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sample space

**-** *H* 

Choice 1

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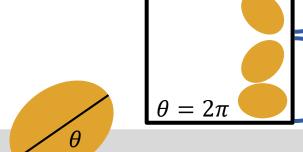
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Sample space:  $\{\theta | 0 \le \theta \le \pi\}$ 

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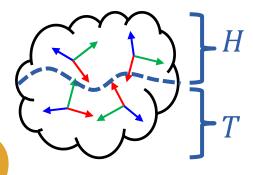
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sample space = SO(3)



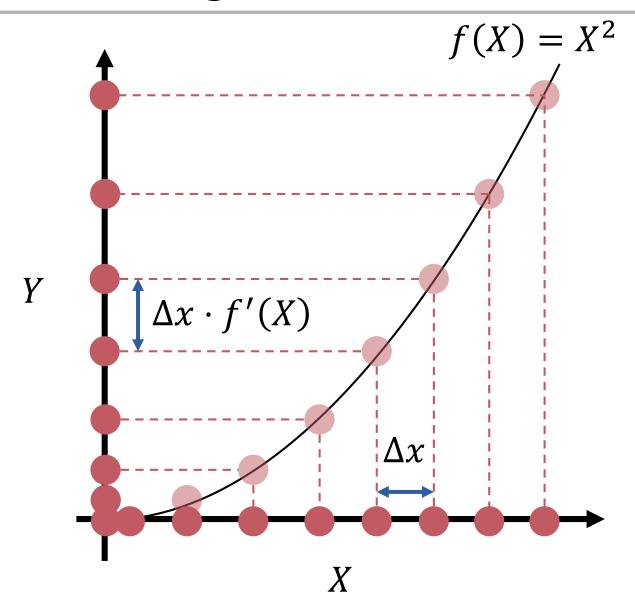
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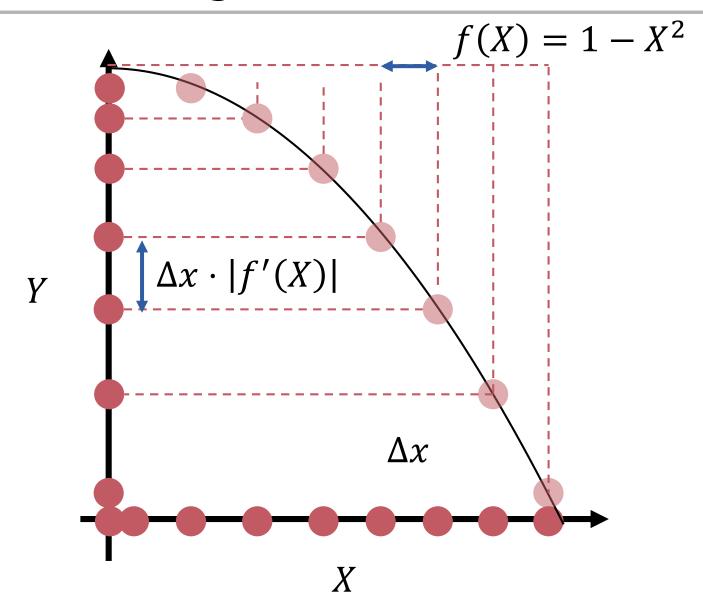




$$Y = f(X)$$

$$p_Y(y) = \frac{p_X(f^{-1}(y))}{f'(f^{-1}(y))}$$



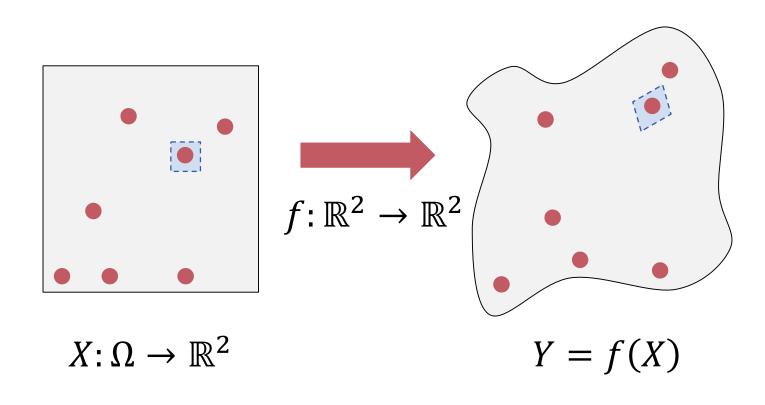


$$Y = f(X)$$

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PDFs should be nonnegative!

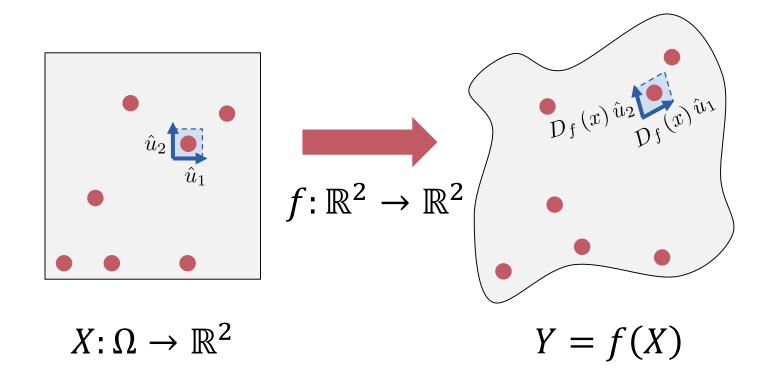




$$Y = f(X)$$

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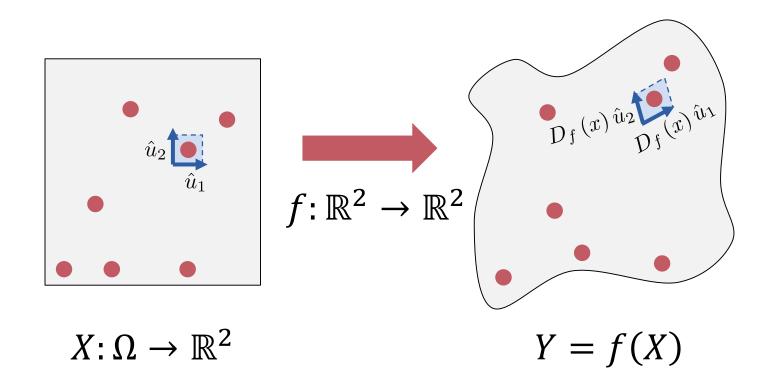
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$$| D_f(x) \hat{u}_1 \times D_f(x) \hat{u}_2 |$$

$$= |\det D_f(x)| ||\hat{u}_1|| ||\hat{u}_2||$$





$$Y = f(X)$$

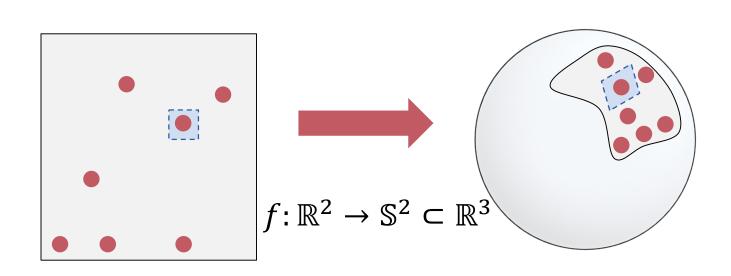
$$p_Y(y) = \frac{p_X(f^{-1}(y))}{|\det D_f(f^{-1}(y))|}^*$$

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 $X:\Omega\to\mathbb{R}^2$ 





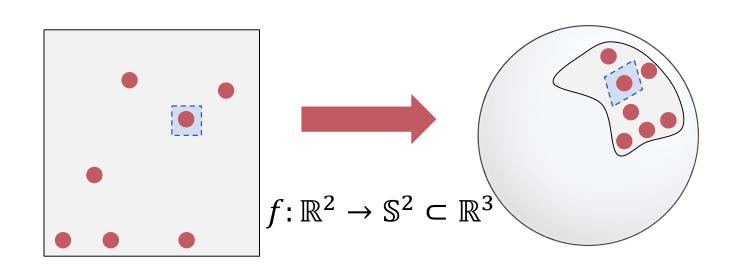
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No determinant for  $3 \times 2$  matrices

 $X:\Omega\to\mathbb{R}^2$ 





Y = f(X)

#### Proposition

$$Y = f(X)$$

$$p_Y(y) = \frac{p_X(f^{-1}(y))}{J_f(f^{-1}(y))}$$

$$\star J_f(x) = \sqrt{\det\left(D_f(x)^T D_f(x)\right)}$$

#### PDF change of variables: same object, different coordinates

