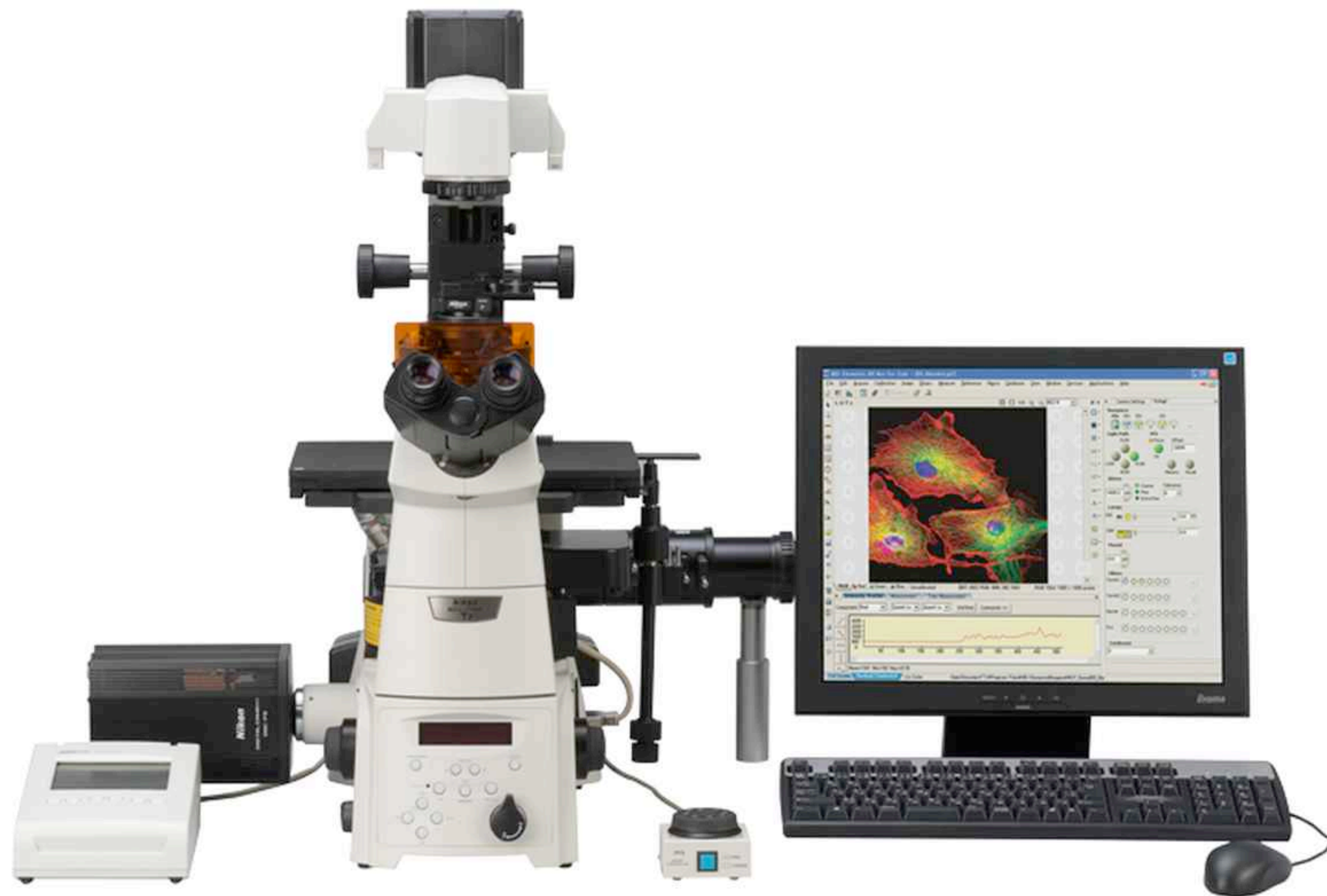
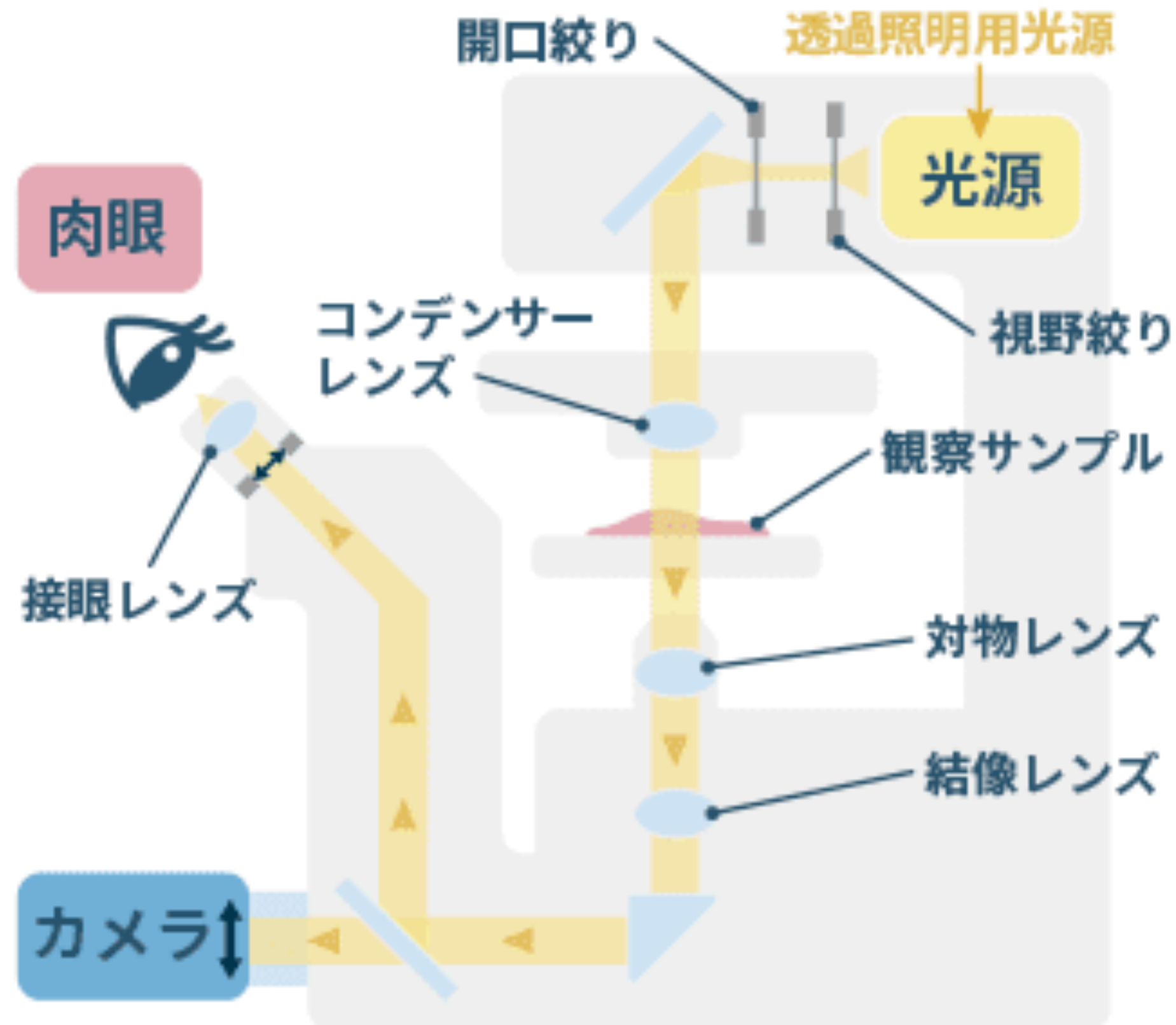


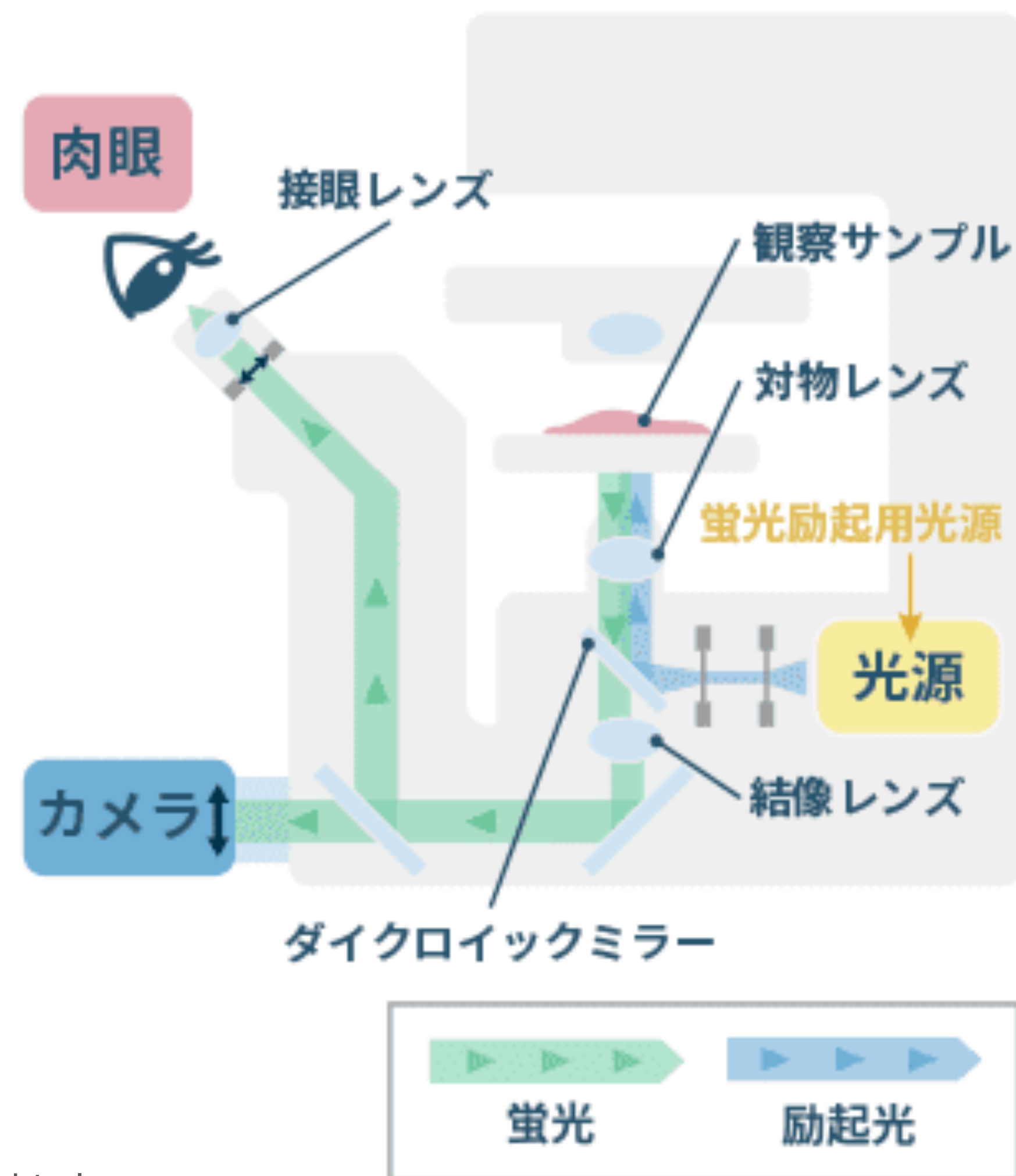
螢光顯微鏡



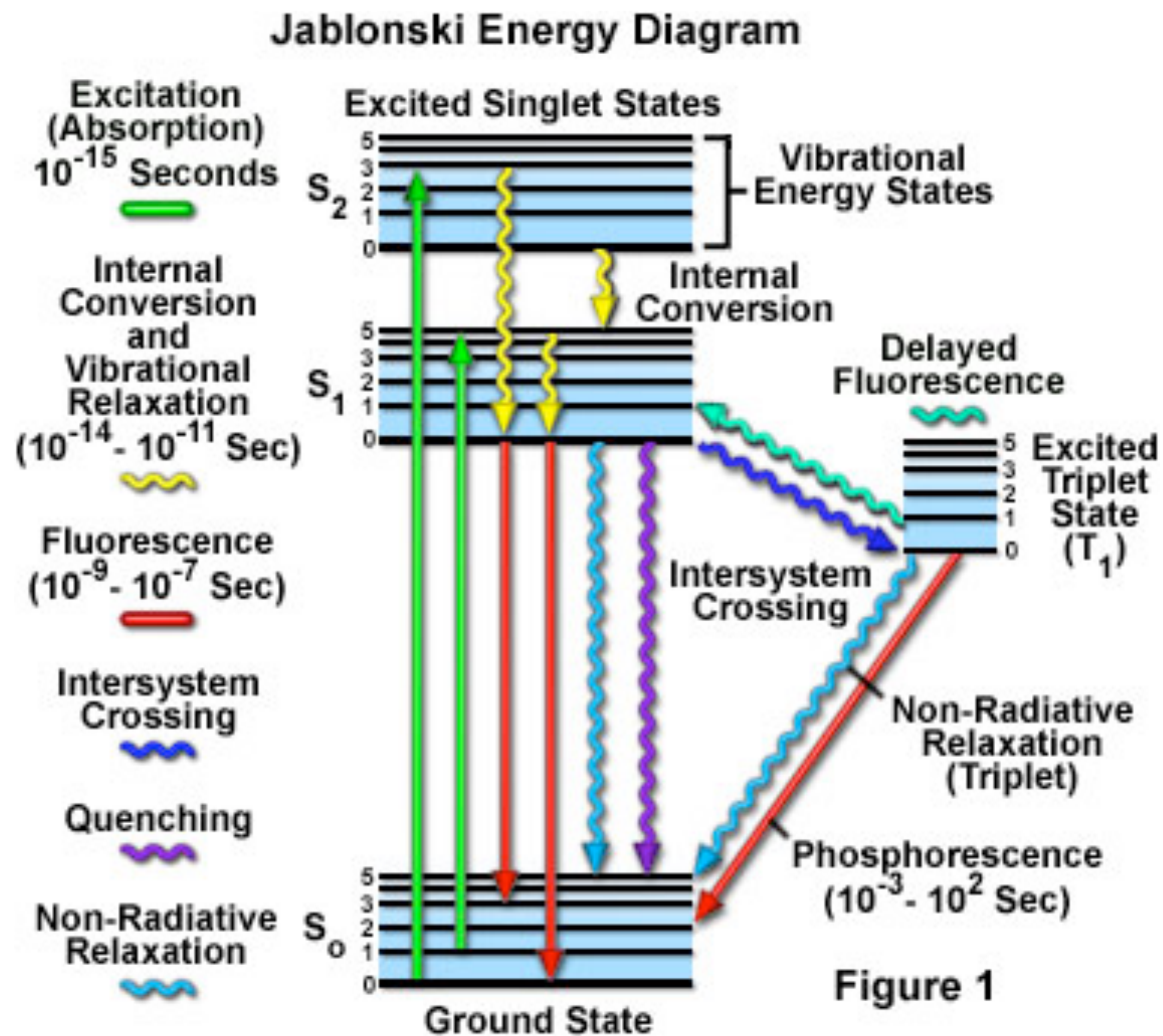
透過観察の原理



蛍光観察の原理

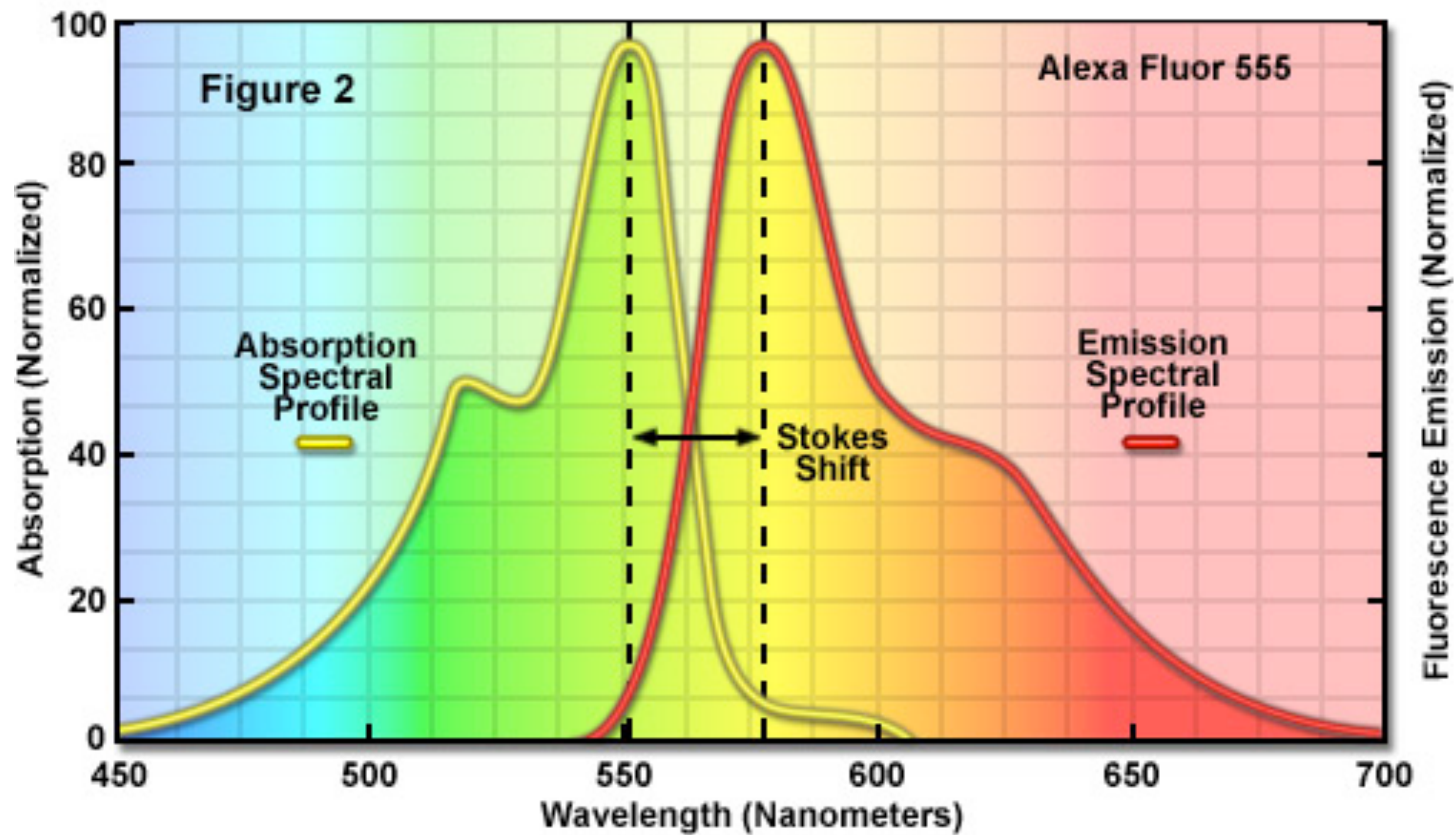


蛍光



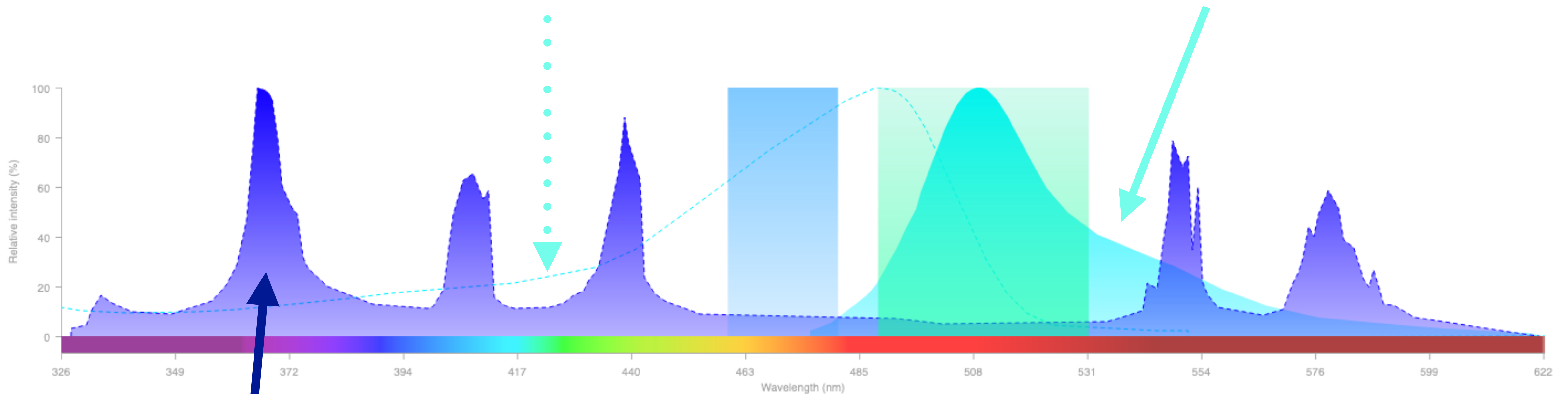
<https://www.olympus-lifescience.com/ja/microscope-resource/primer/techniques/confocal/fluoroexciteemit/>

Fluorophore Absorption and Emission Profiles



Excitation spectrum

Emission spectrum

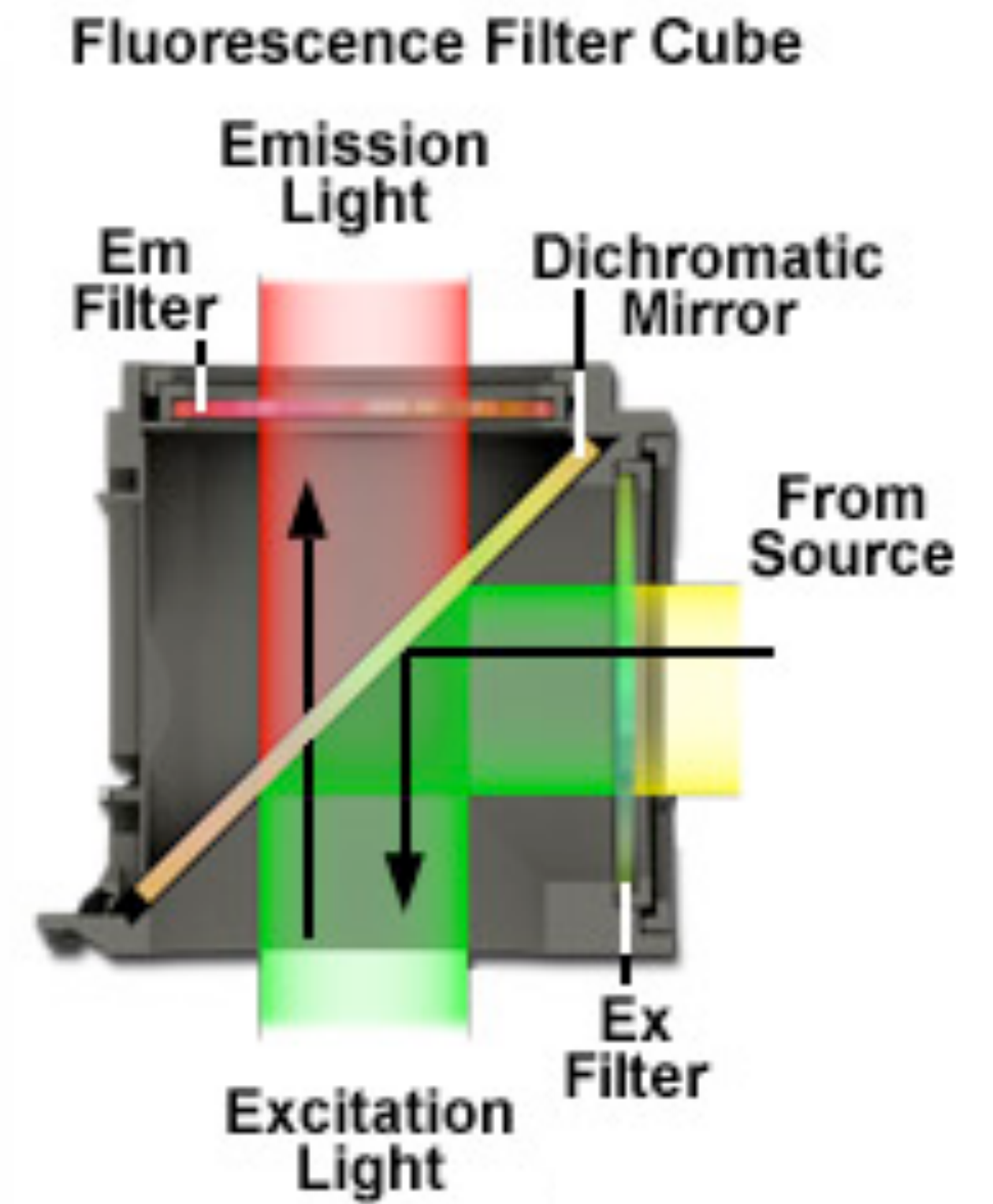
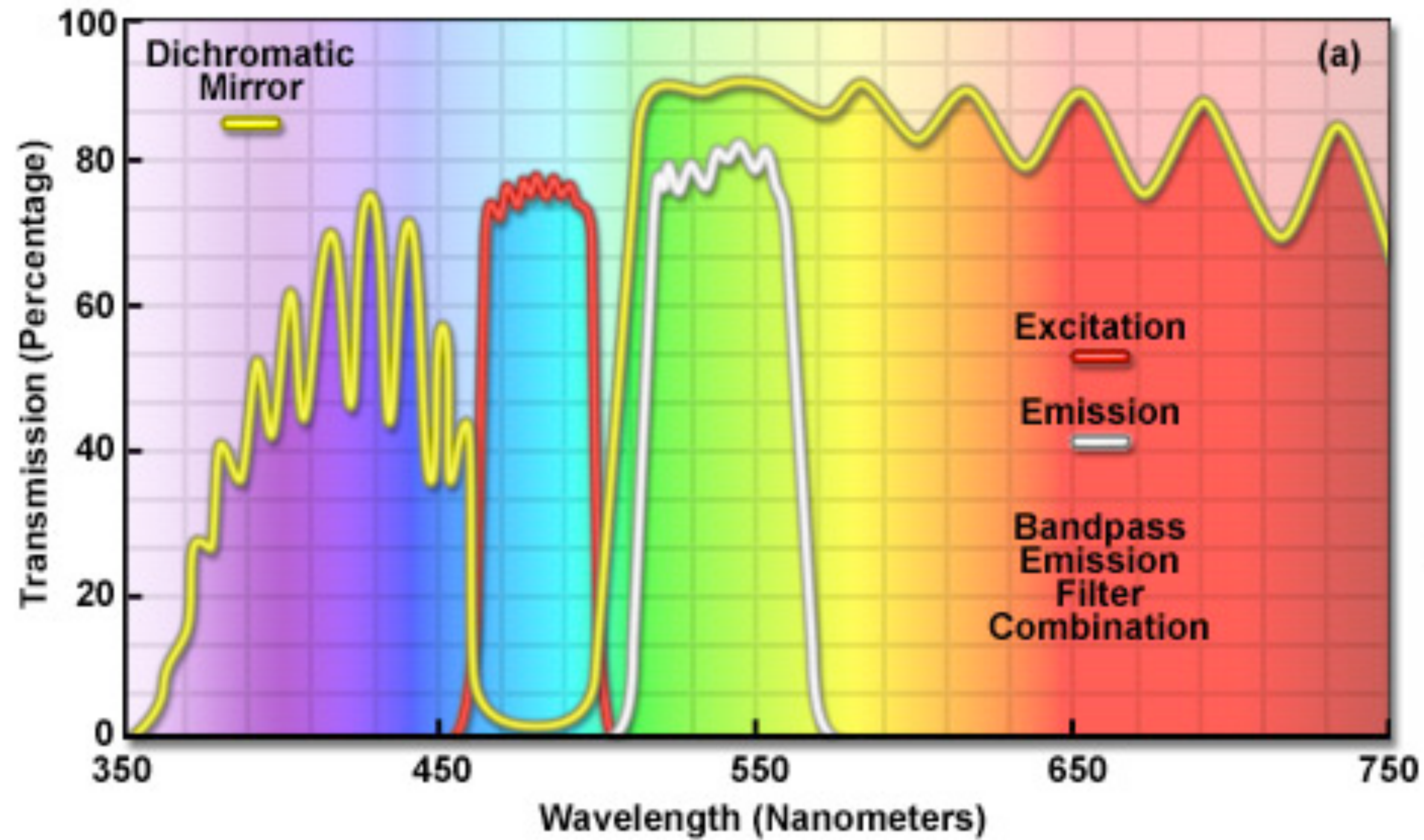


Excitation filter

Emission filter

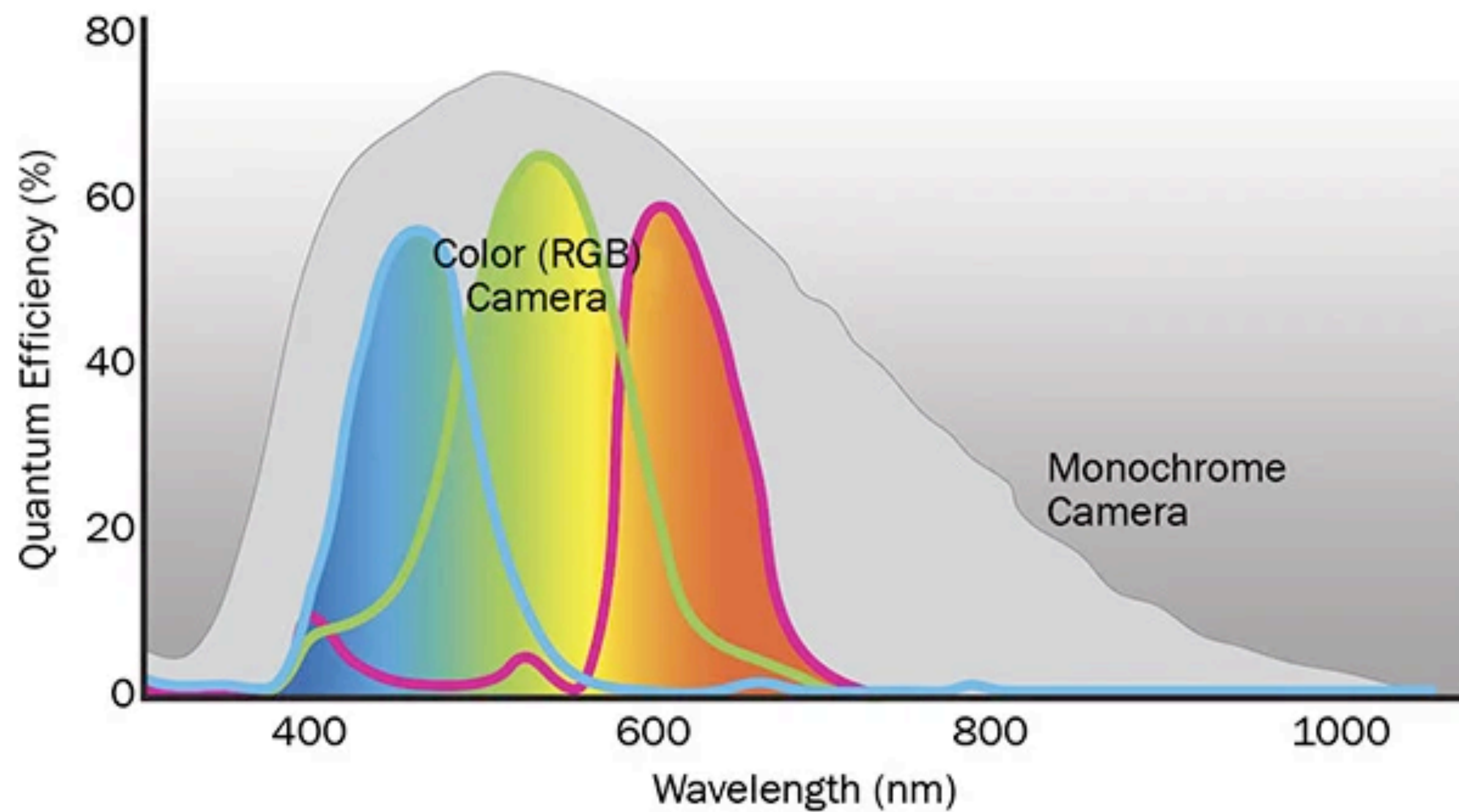
Mercury lamp spectrum

FITC / GFP Fluorescence Filter Combination

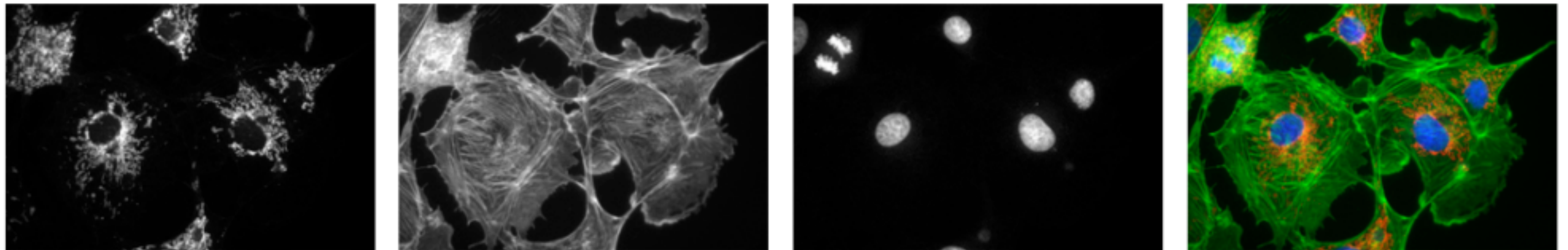


Sample

カメラ

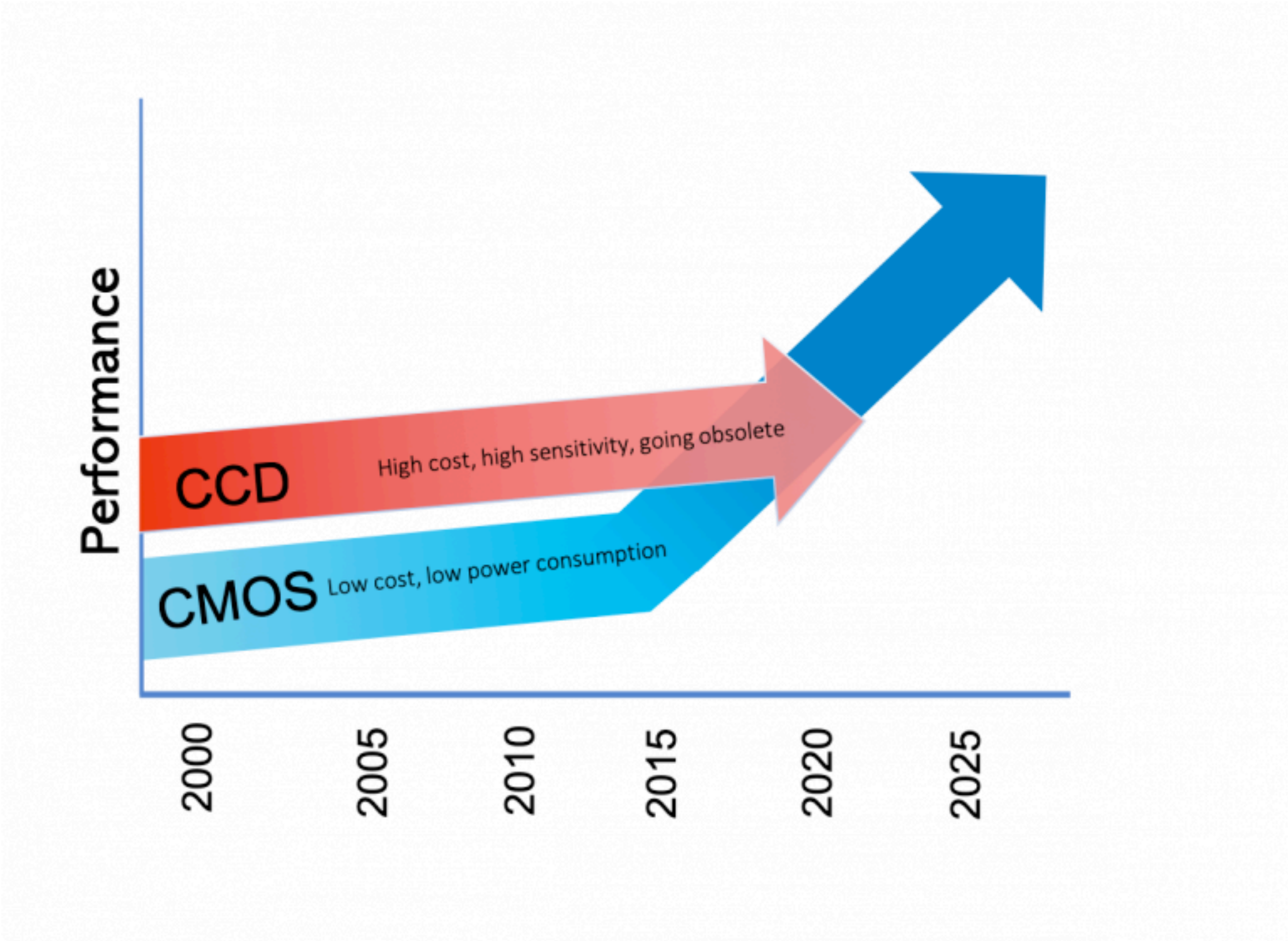
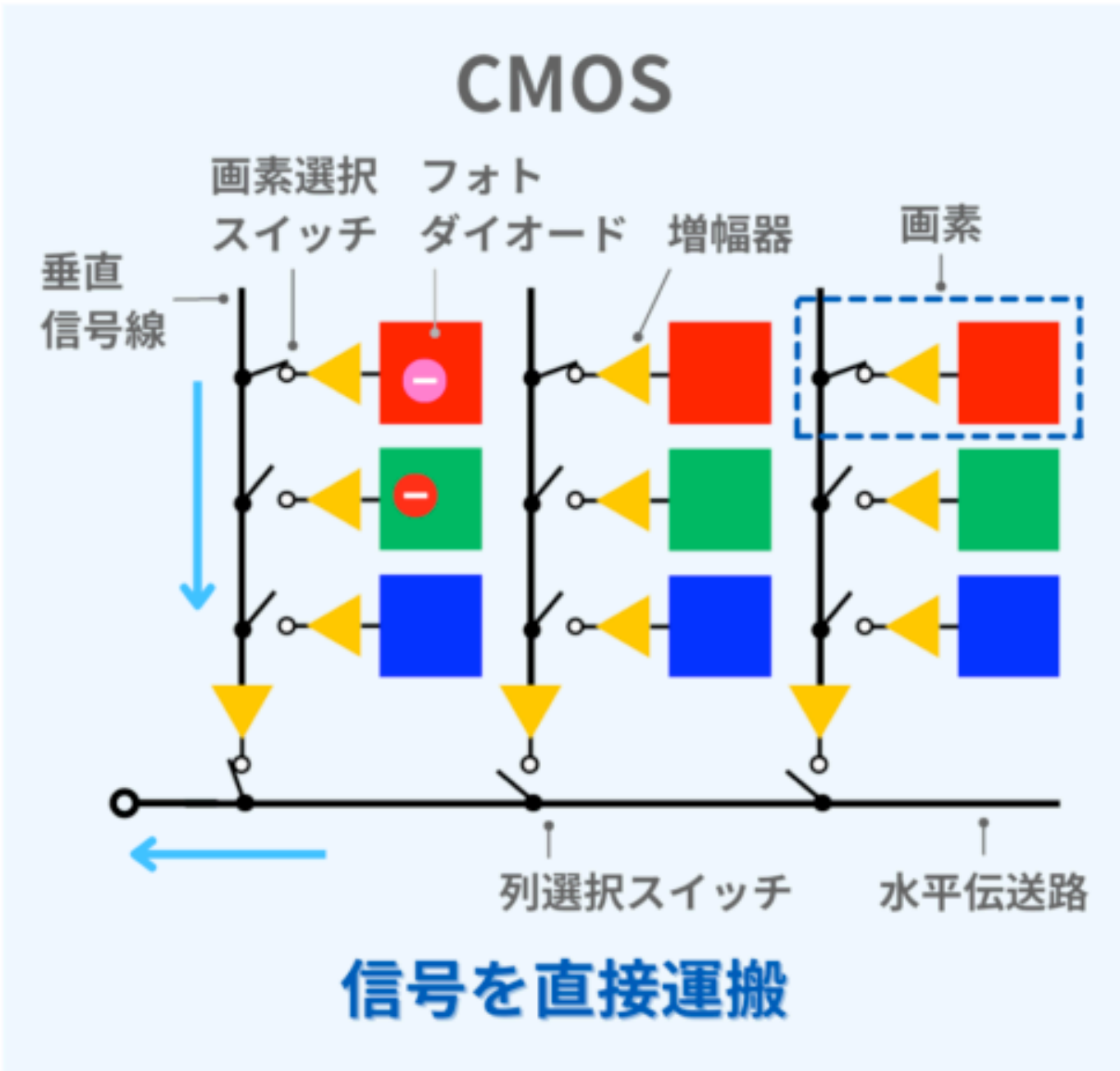
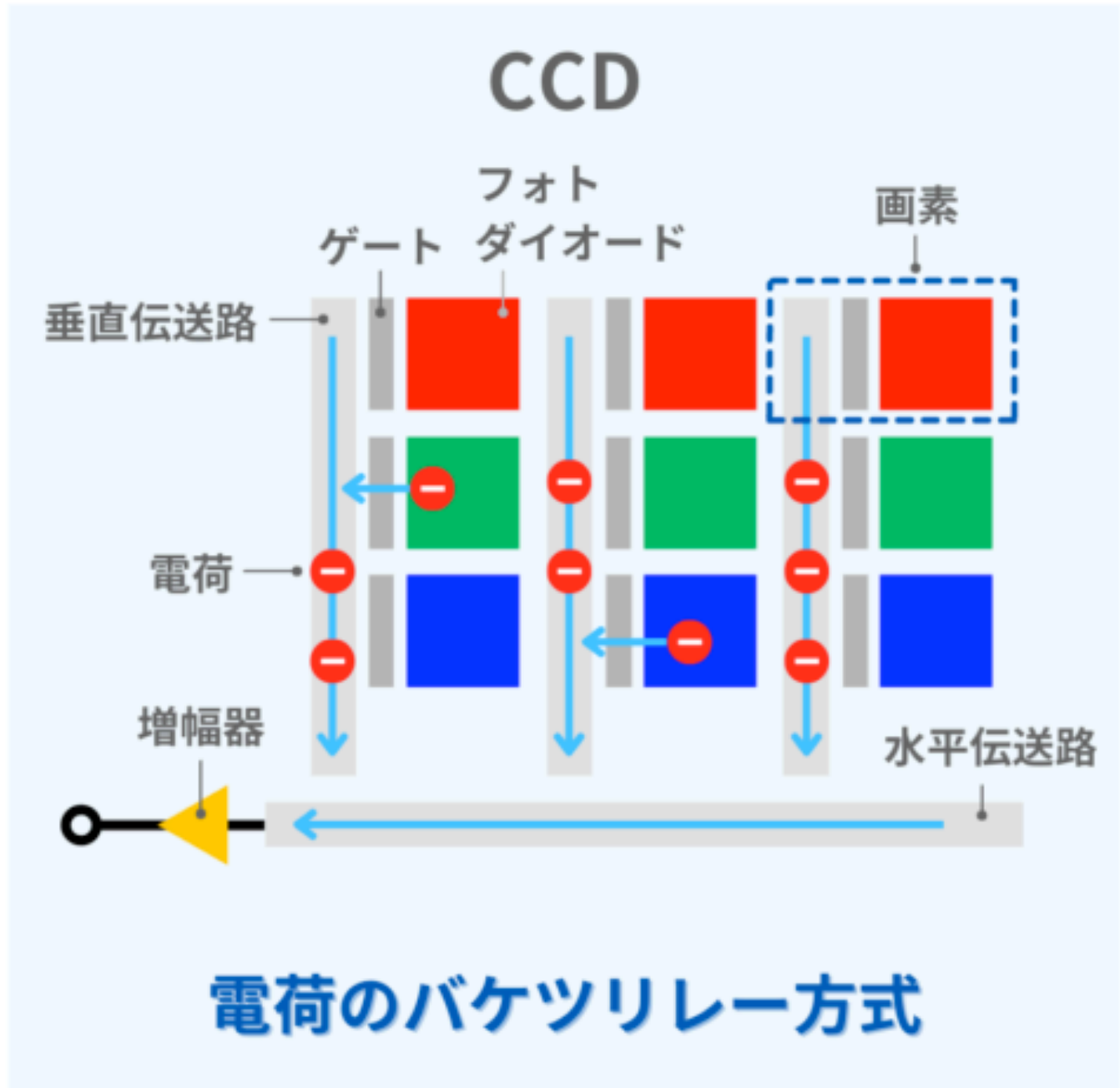


擬似カラー Pseudocolor



Gray scale images

CCD vs CMOS



Fiji のダウンロード

<https://imagej.net/software/fiji/downloads>



Fiji Downloads

Fiji is a distribution of ImageJ which includes many useful plugins [contributed by the community](#).

~ Download Fiji for your OS ~



64-bit

or



macOS



64-bit

*Mexican Hat Filter Plug-in*のダウンロード

<https://imagej.nih.gov/ij/plugins/mexican-hat/index.html>

[home](#) | [news](#) | [docs](#) | [download](#) | [plugins](#) | [macros/dev](#) | [list](#) | [links](#)

Mexican Hat Filter

Author: Dimitar Prodanov (dimiterpp at gmail.com)

History: 2012/11/10: First release
2012/11/24: Added "Separable" option; requires v1.47g
2012/11/26: Multi-threaded
2015/09/12: Bug fixes
2017/12/31: Fixed kernel calculation bug

Source: [Mexican_Hat_Filter.java](#)

Installation: Drag and drop [Mexican_Hat_Filter.class](#) onto the "ImageJ" window.

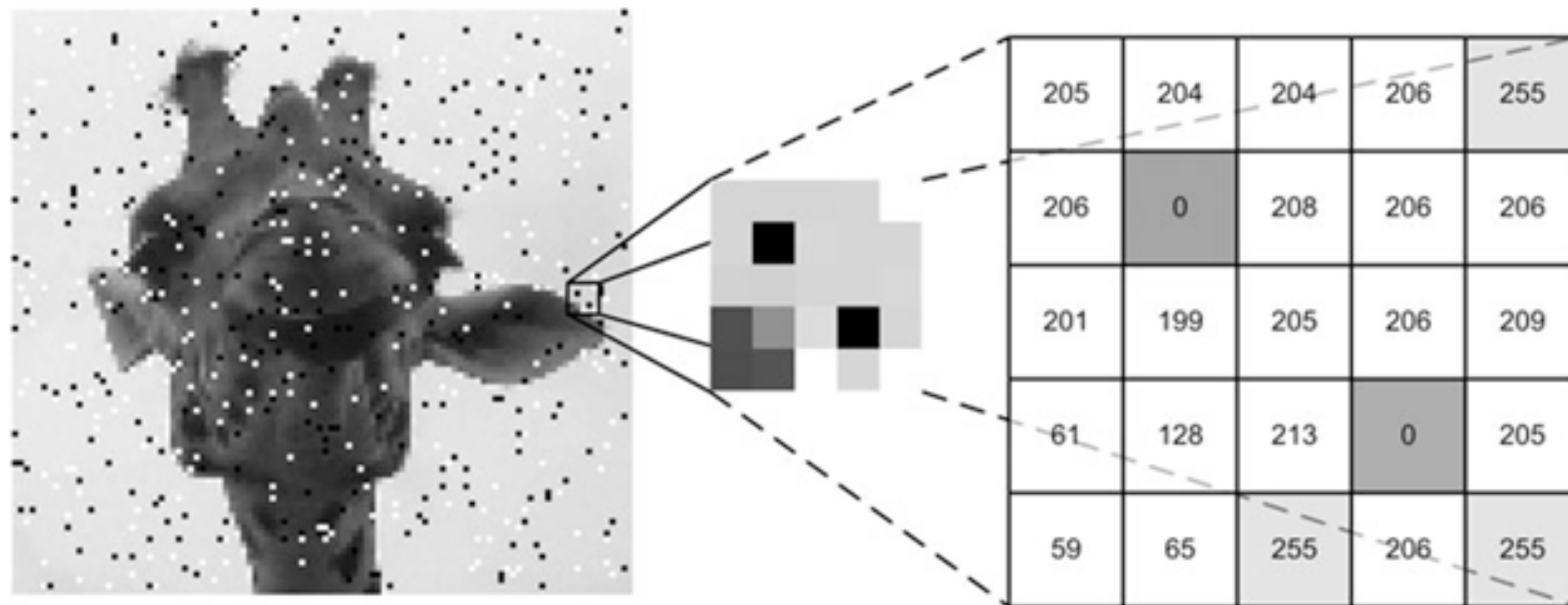
Description: This plugin applies a Laplacian of Gaussian (Mexican Hat) filter to a 2D image.

See Also: [3D Laplacian of Gaussian \(LoG\) plugin](#)
[Difference of Gaussians plugin](#)



画像処理

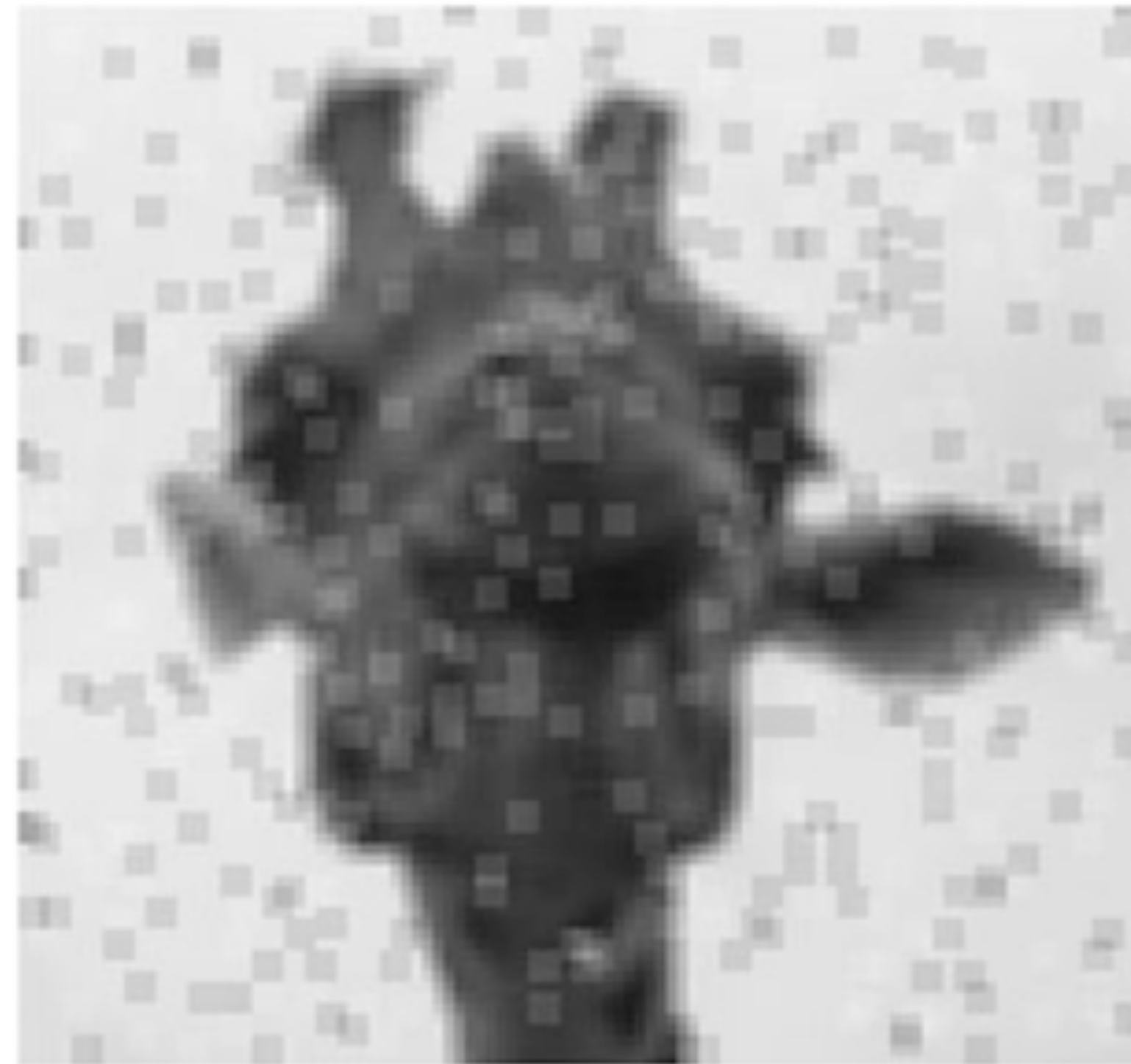
— *Filter* —



ノイズ (外れ値) を除去したい

画像処理

— *Filter* —



Mean filtered

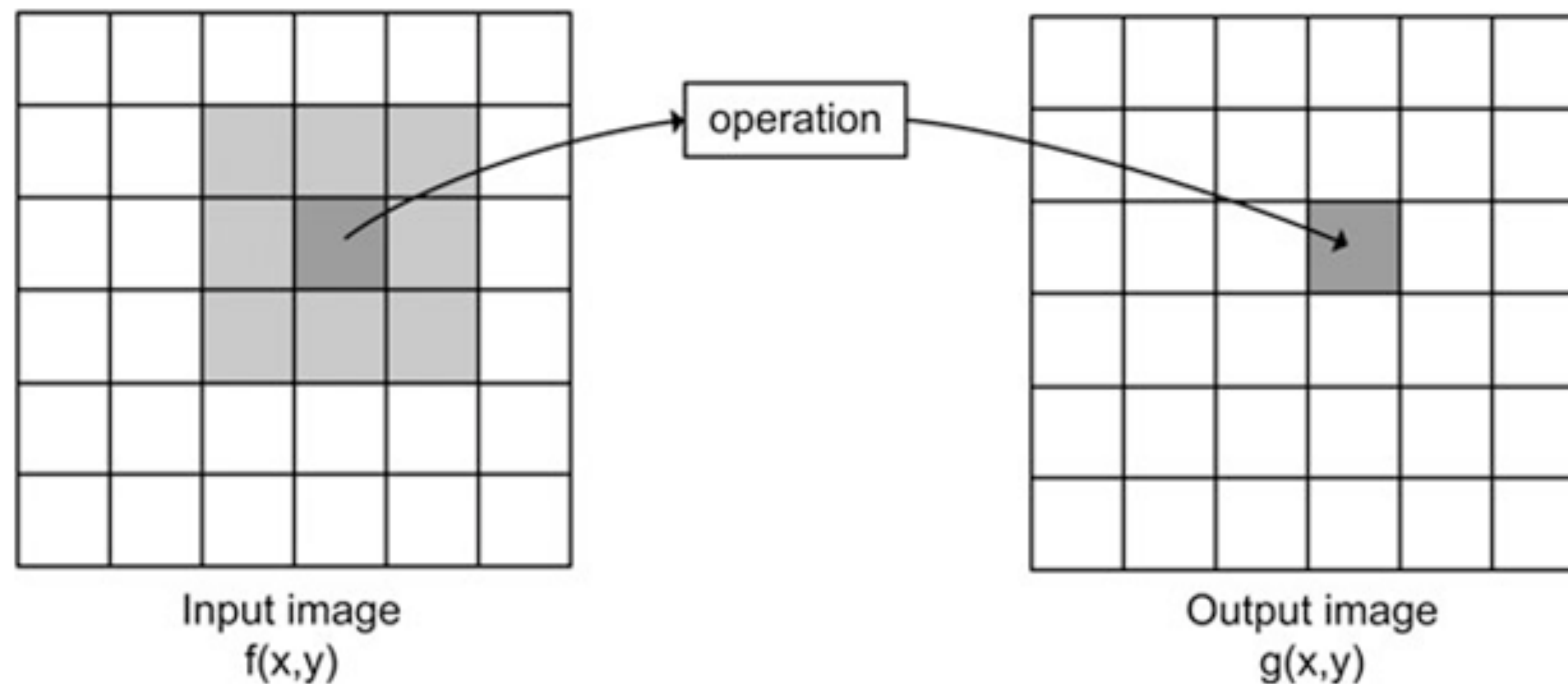


Median filtered

ノイズ (外れ値) を除去したい

画像処理

— *Filter* —

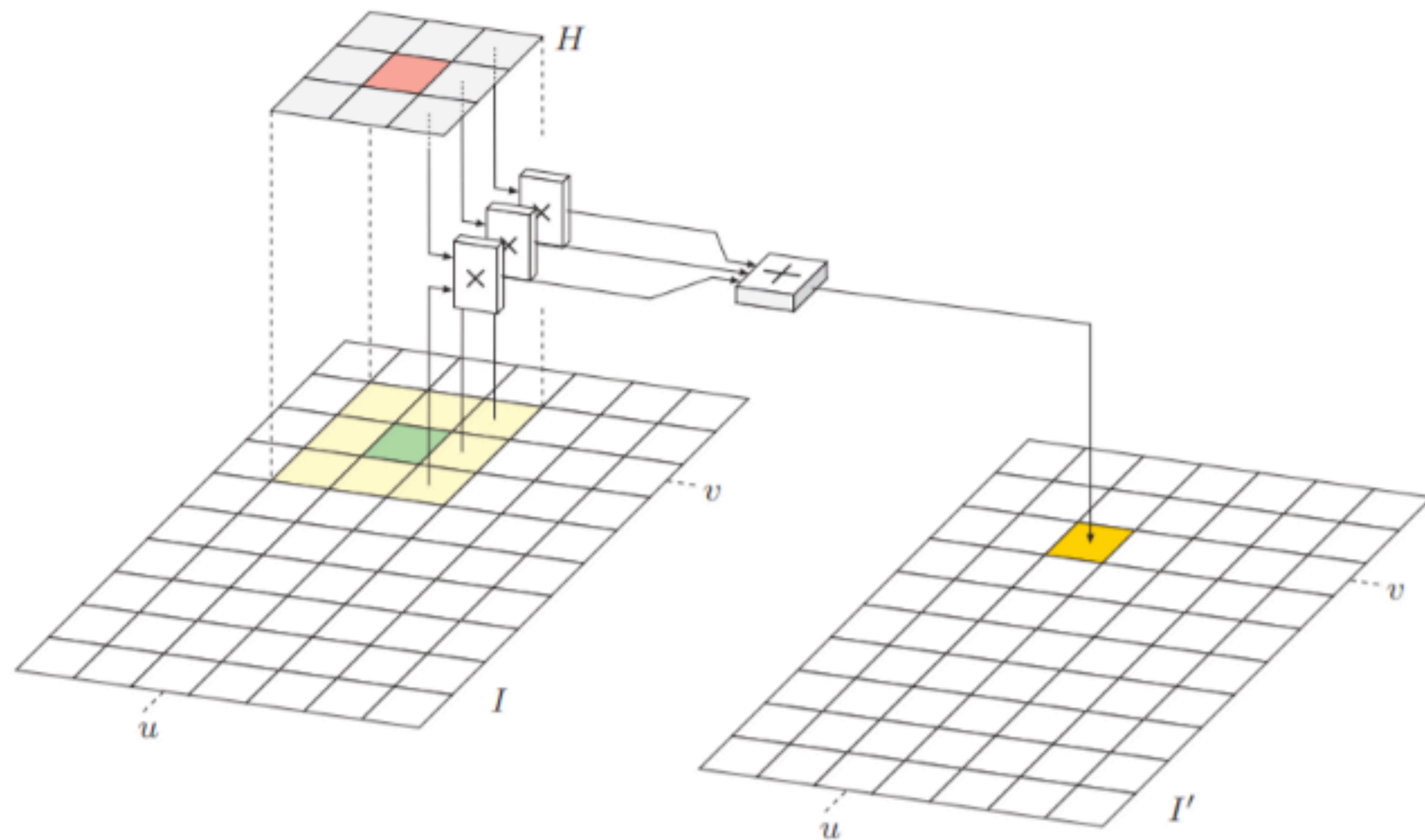


Radius = 1

Kernel width/height = (Radius x 2) + 1 = 3

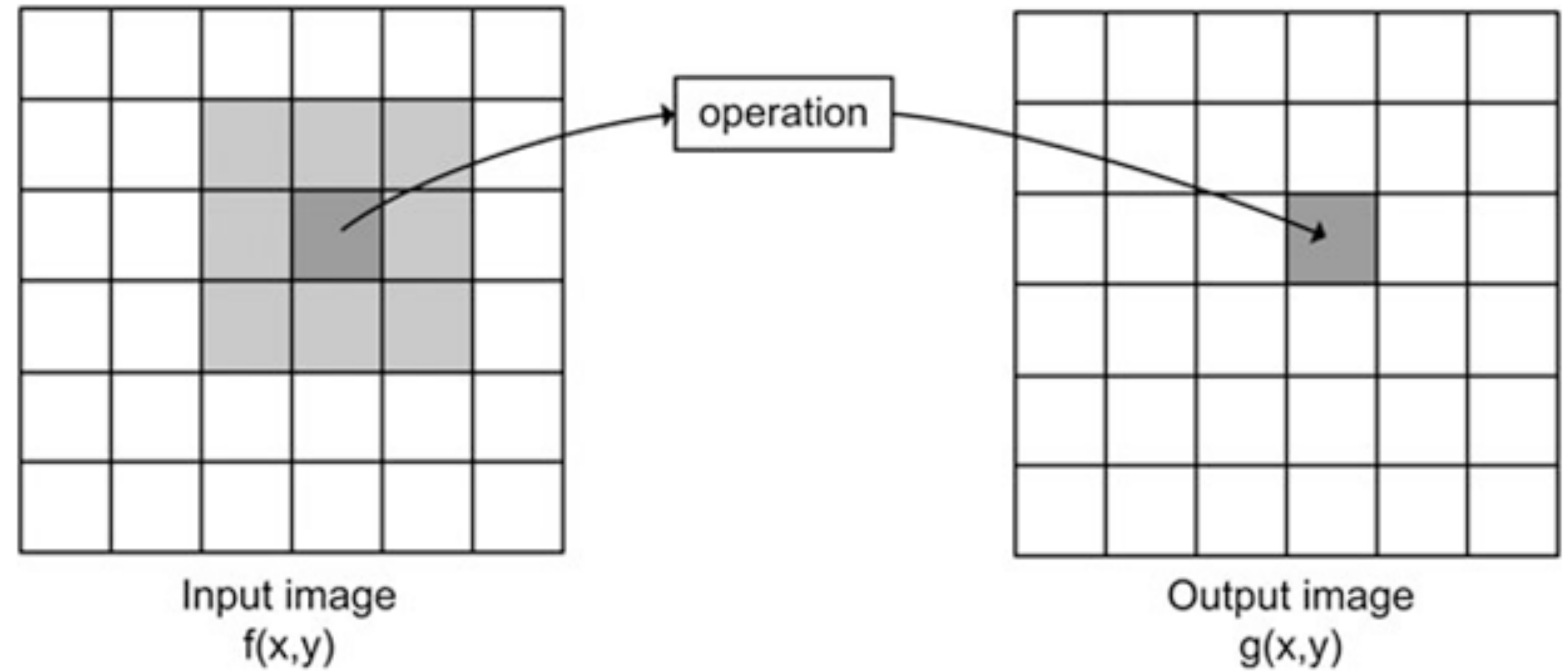
画像処理

— *Filter* —



图像处理

— *Filter* —

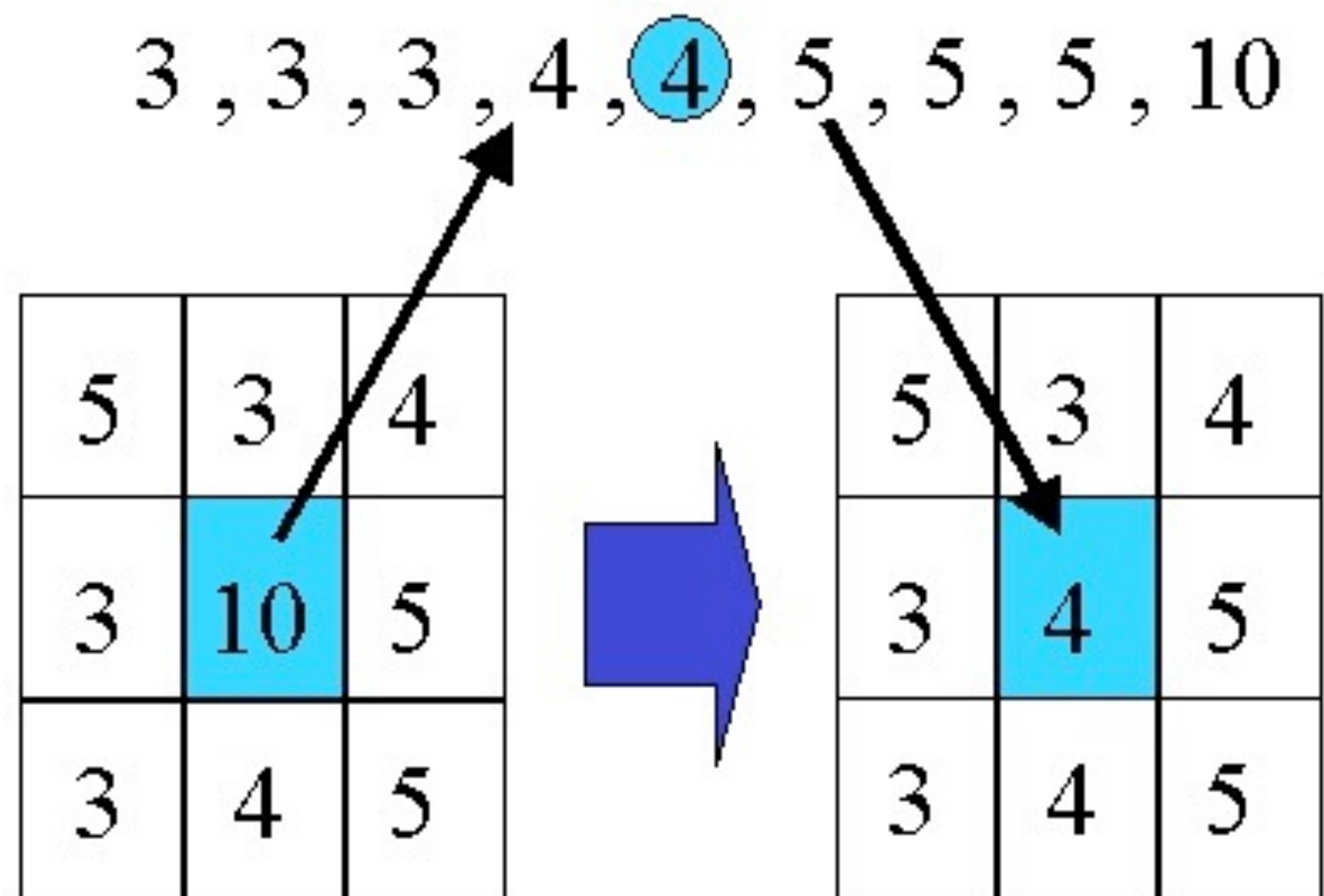


$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$
$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$
$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$

Kernel for 3 x 3 mean filter

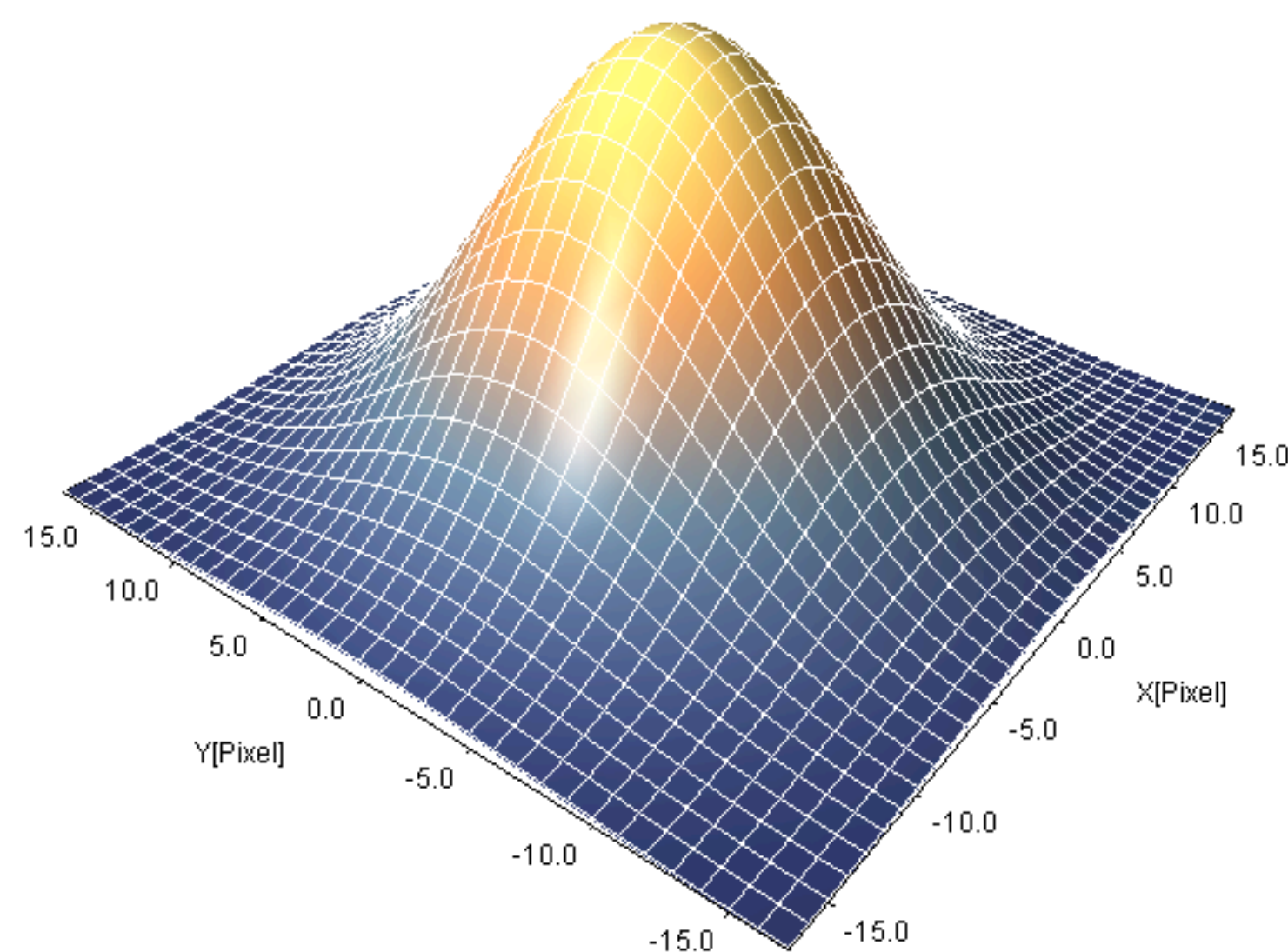
图像处理

— *Filter* —



3 x 3 median filtering

Gaussian Filter



$$Gauss(x, y) = \frac{1}{2\pi\sigma^2} \exp\left(-\frac{x^2 + y^2}{2\sigma^2}\right)$$

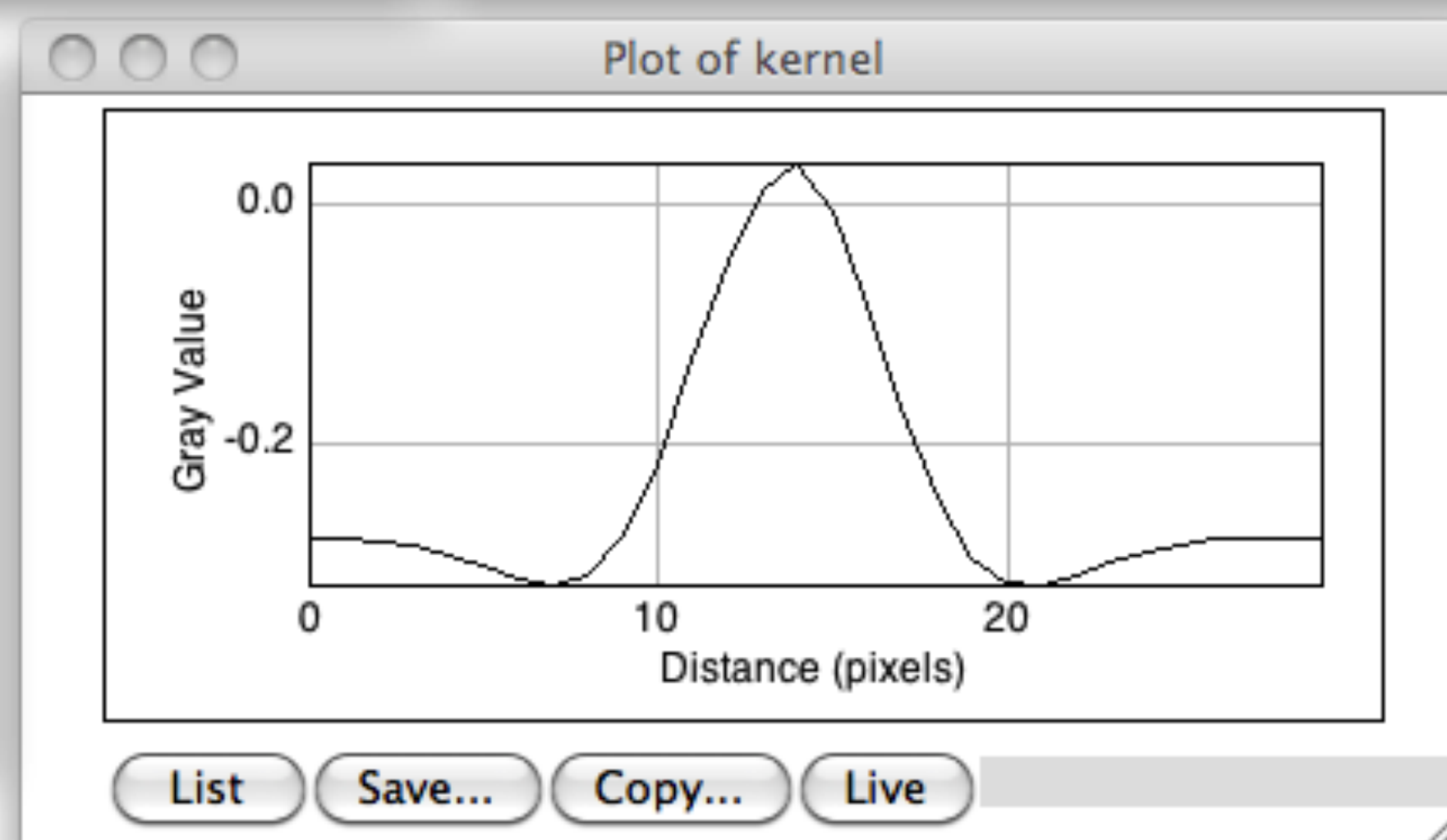
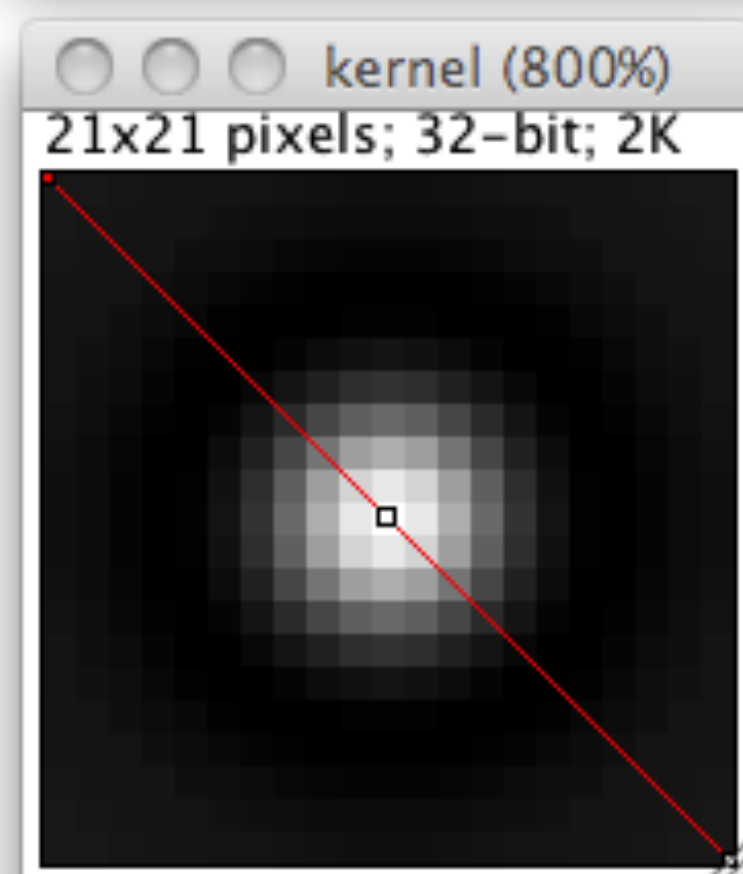
1/16	2/16	1/16
2/16	4/16	2/16
1/16	2/16	1/16

3×3の重みづけ

1/256	4/256	6/256	4/256	1/256
4/256	16/256	24/256	16/256	4/256
6/256	24/256	36/256	24/256	6/256
4/256	16/256	24/256	16/256	4/256
1/256	4/256	6/256	4/256	1/256

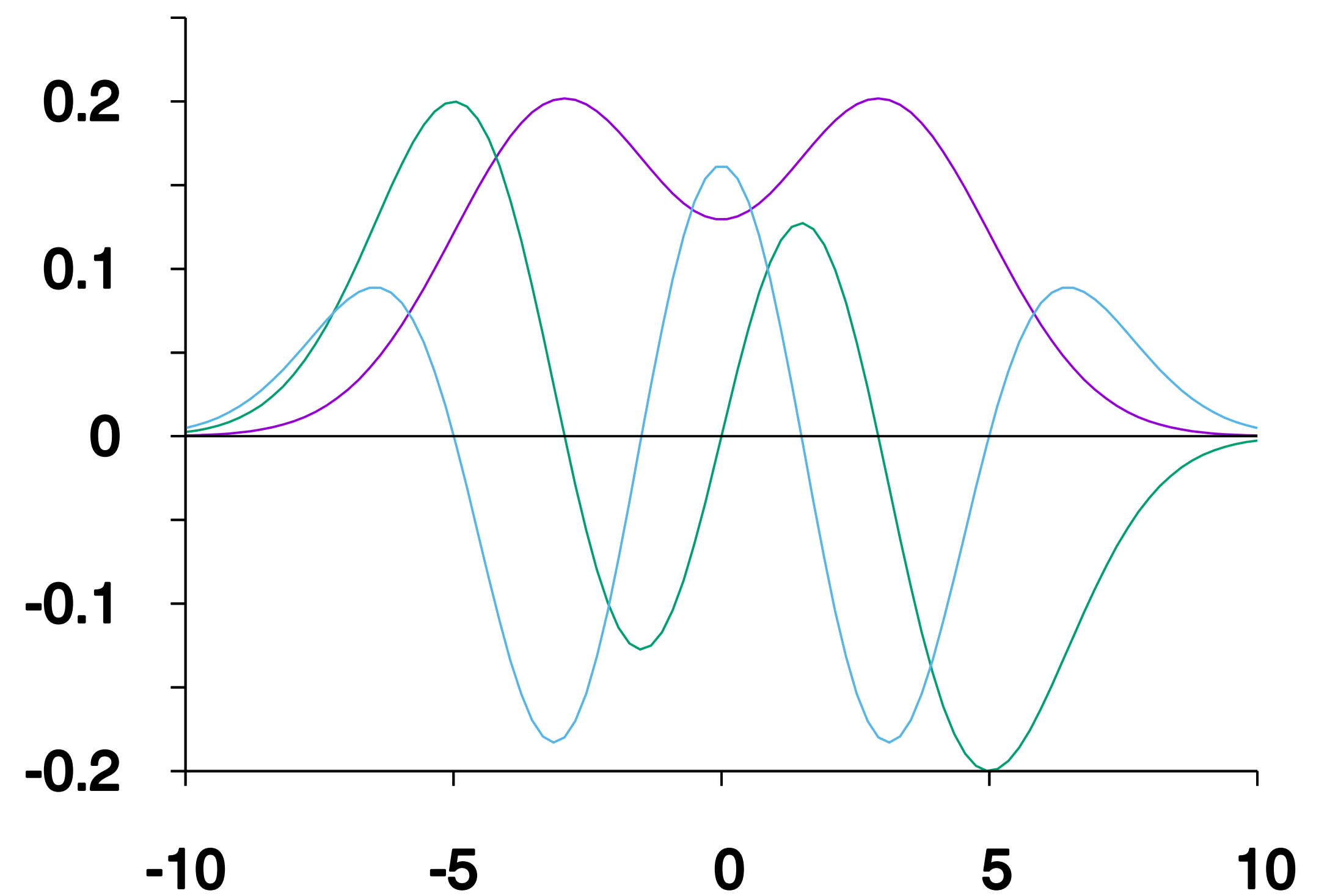
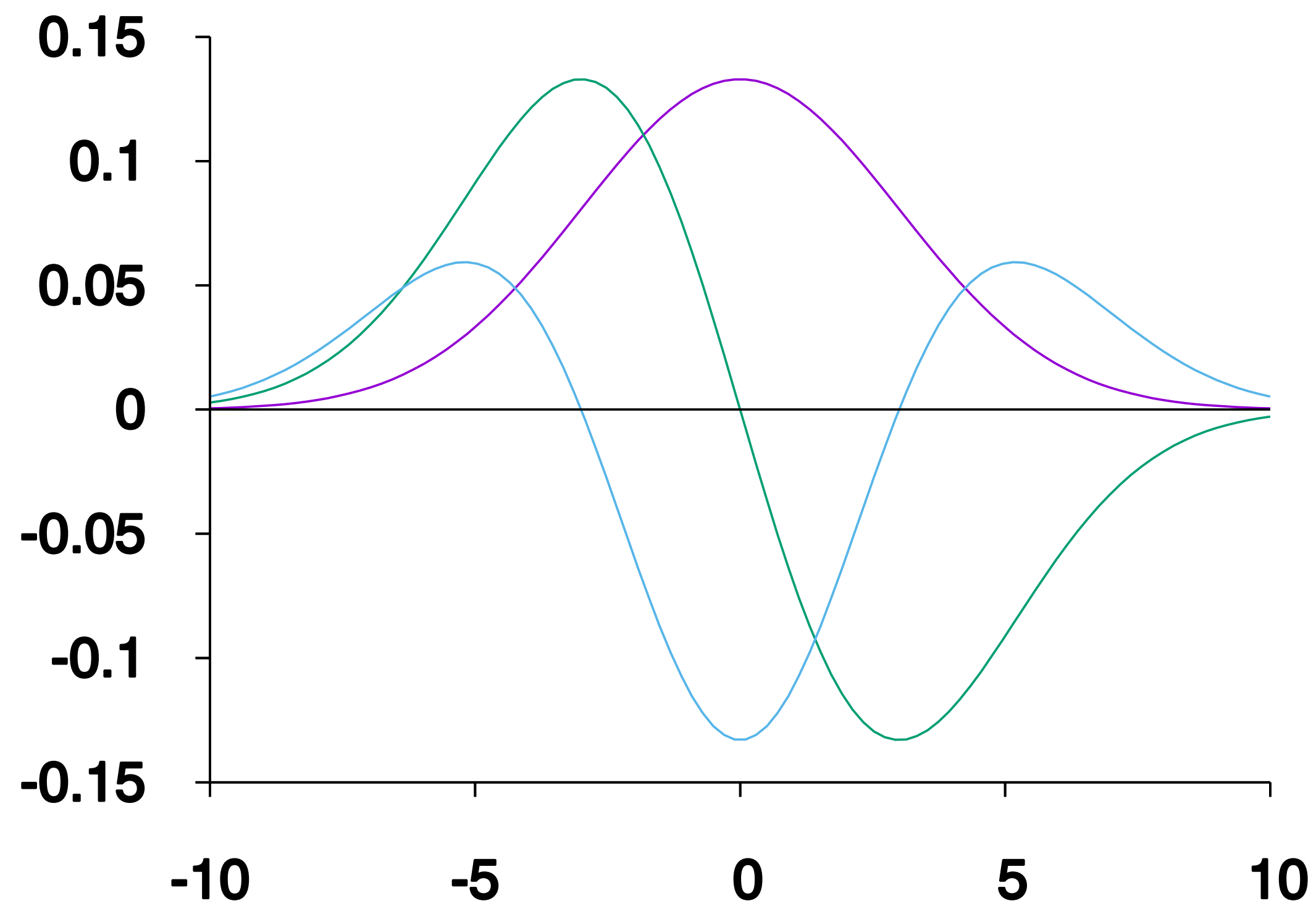
5×5の重みづけ

Mexican hat filter

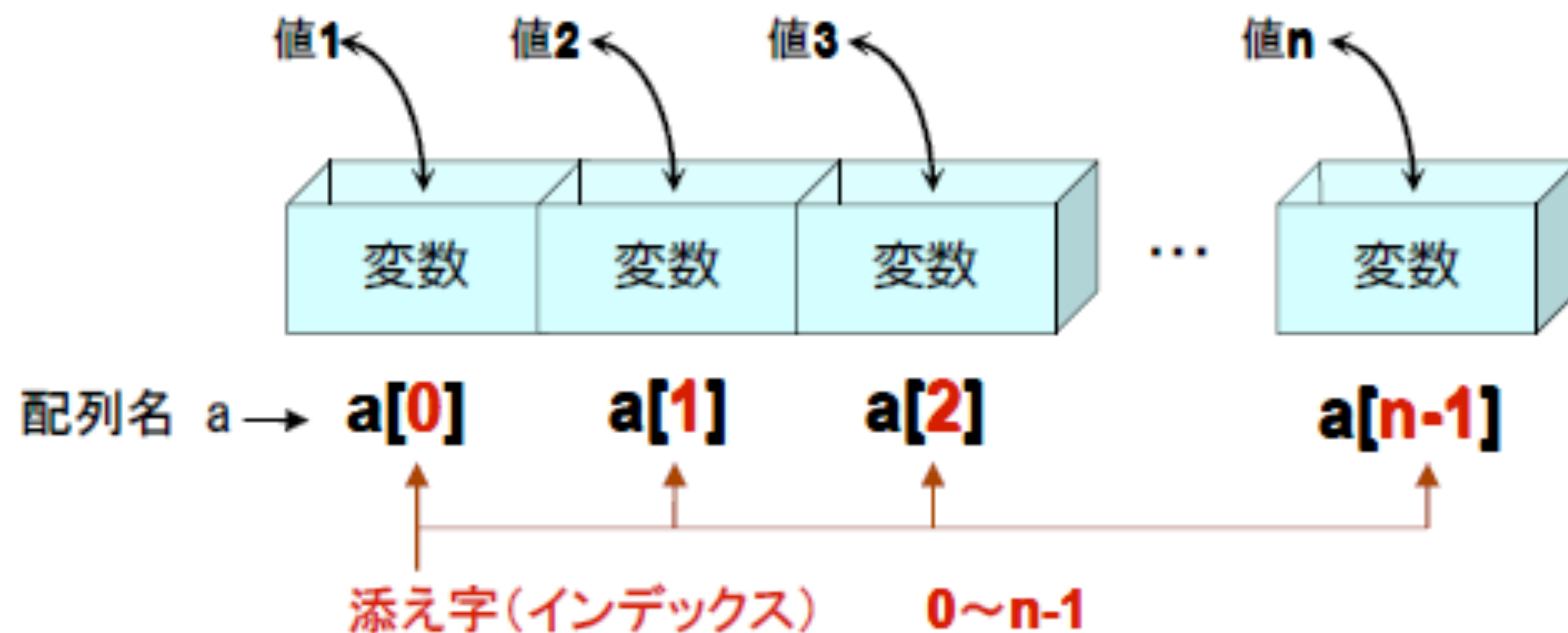


Edge detection

Original
1st derivative
2nd derivative



プログラミング： 配列

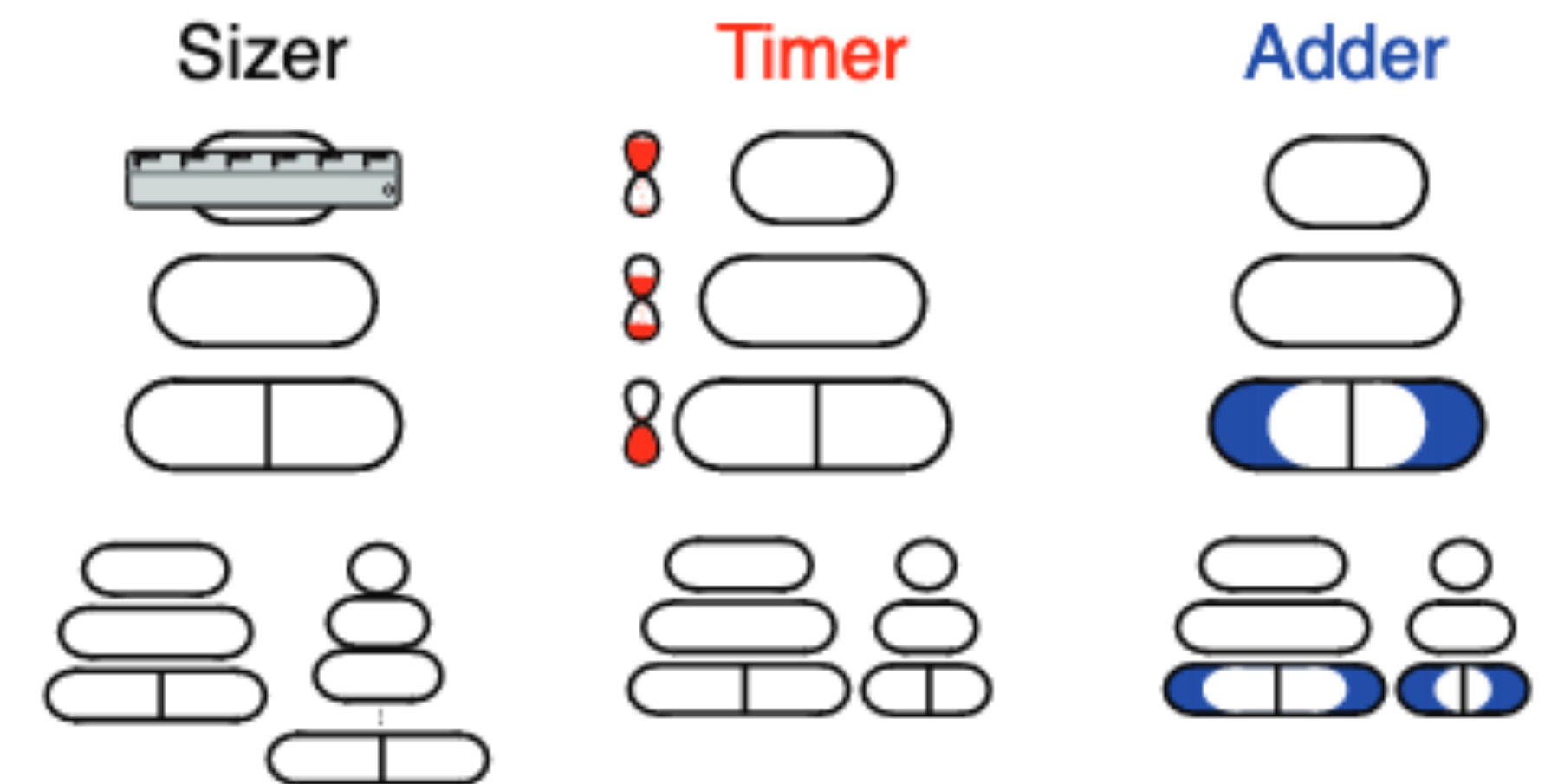


細胞分裂の3つのモデル

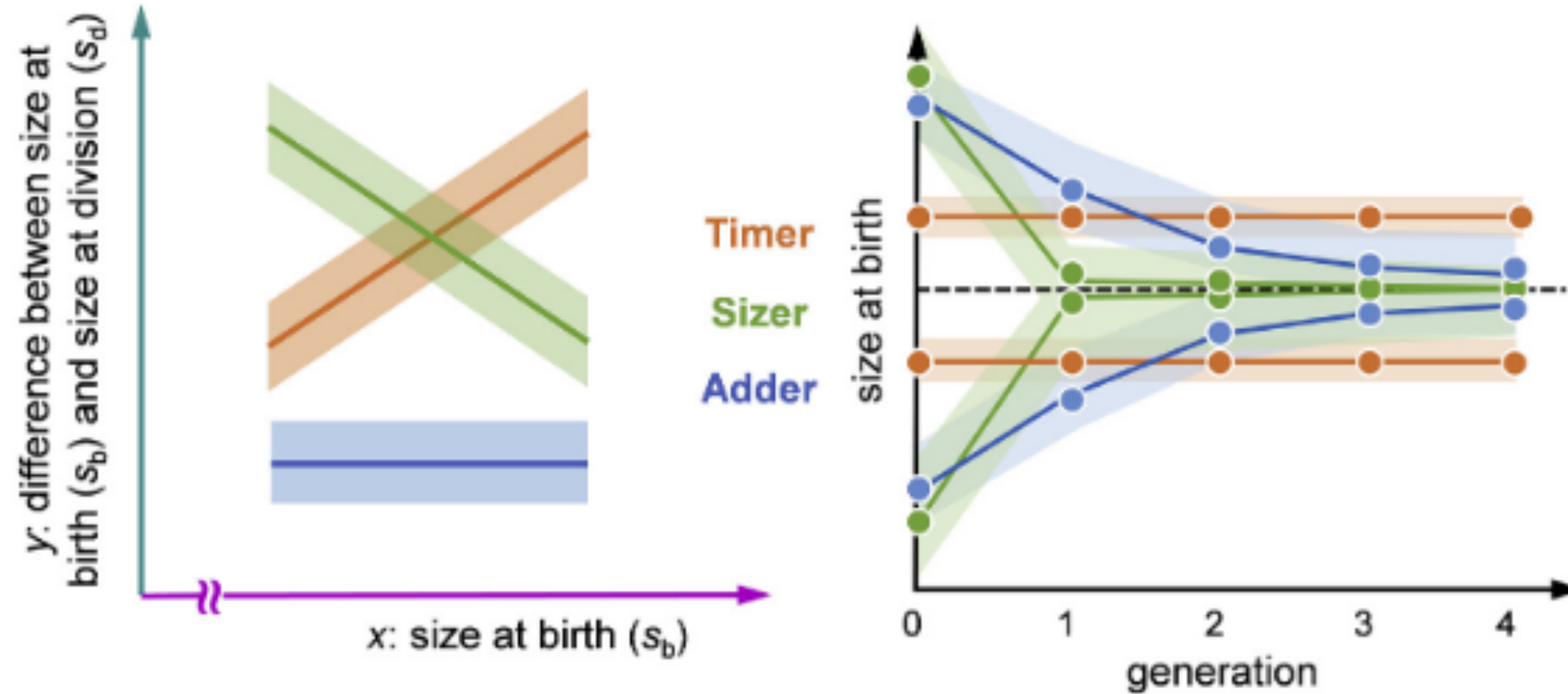
Sizer : ある決まった大きさになったら分裂する

Timer : ある決まった時間が経過したら分裂する

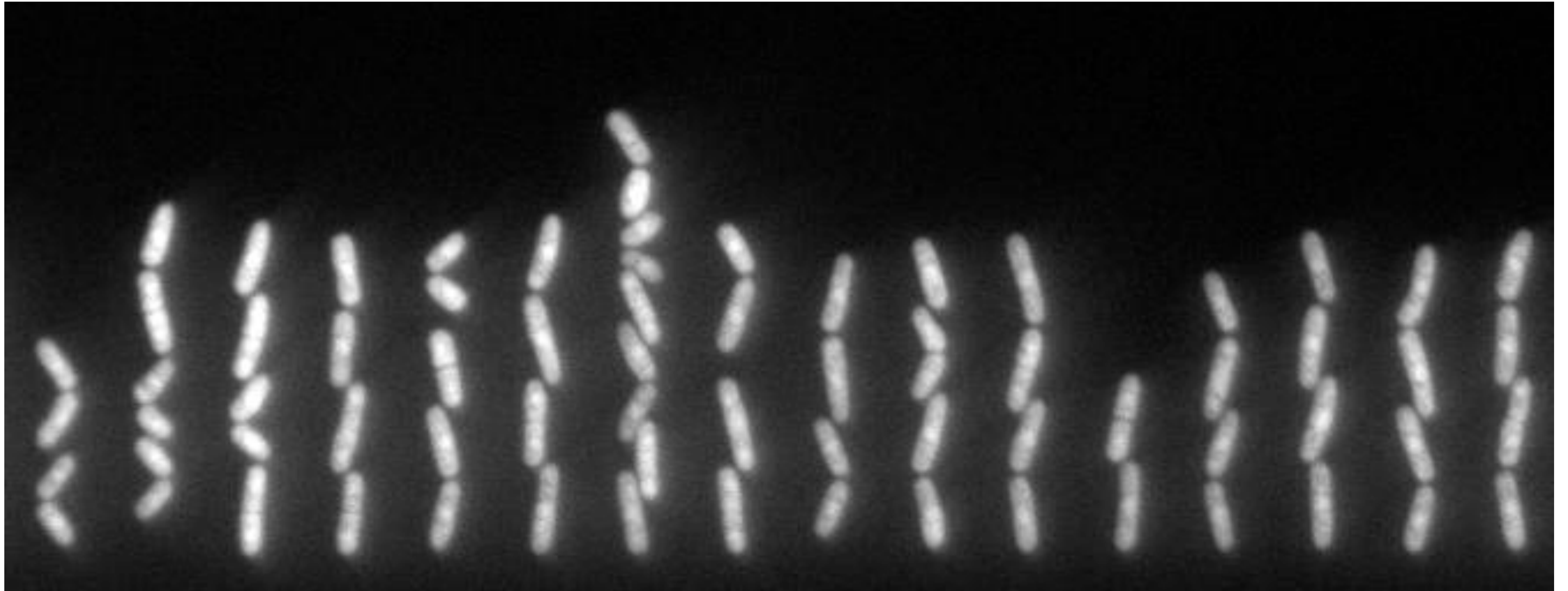
Adder : ある決まった体積分成長したら分裂する



細胞分裂の3つのモデル



マイクロ流体デバイスを用いた 分裂酵母のイメージング



自己相関係数の計算

オリジナルデータ $x[0]$ $x[1]$ $x[2]$ $x[3]$ $x[4]$ $x[5]$ ($N = 6$)

$\tau = 1$ $x[0]$ $x[1]$ $x[2]$ $x[3]$ $x[4]$ $x[5]$ ($x[0]x[1] + x[1]x[2] + x[2]x[3]$) / 3



$\tau = 2$ $x[0]$ $x[1]$ $x[2]$ $x[3]$ $x[4]$ $x[5]$ ($x[0]x[2] + x[1]x[3] + x[2]x[4]$) / 3



$\tau = 3$ $x[0]$ $x[1]$ $x[2]$ $x[3]$ $x[4]$ $x[5]$ ($x[0]x[3] + x[1]x[4] + x[2]x[5]$) / 3



自己相関係数の計算

オリジナルデータ

x[0] x[1] x[2] x[3] x[4] x[5] (N = 6)

$\tau = 1$

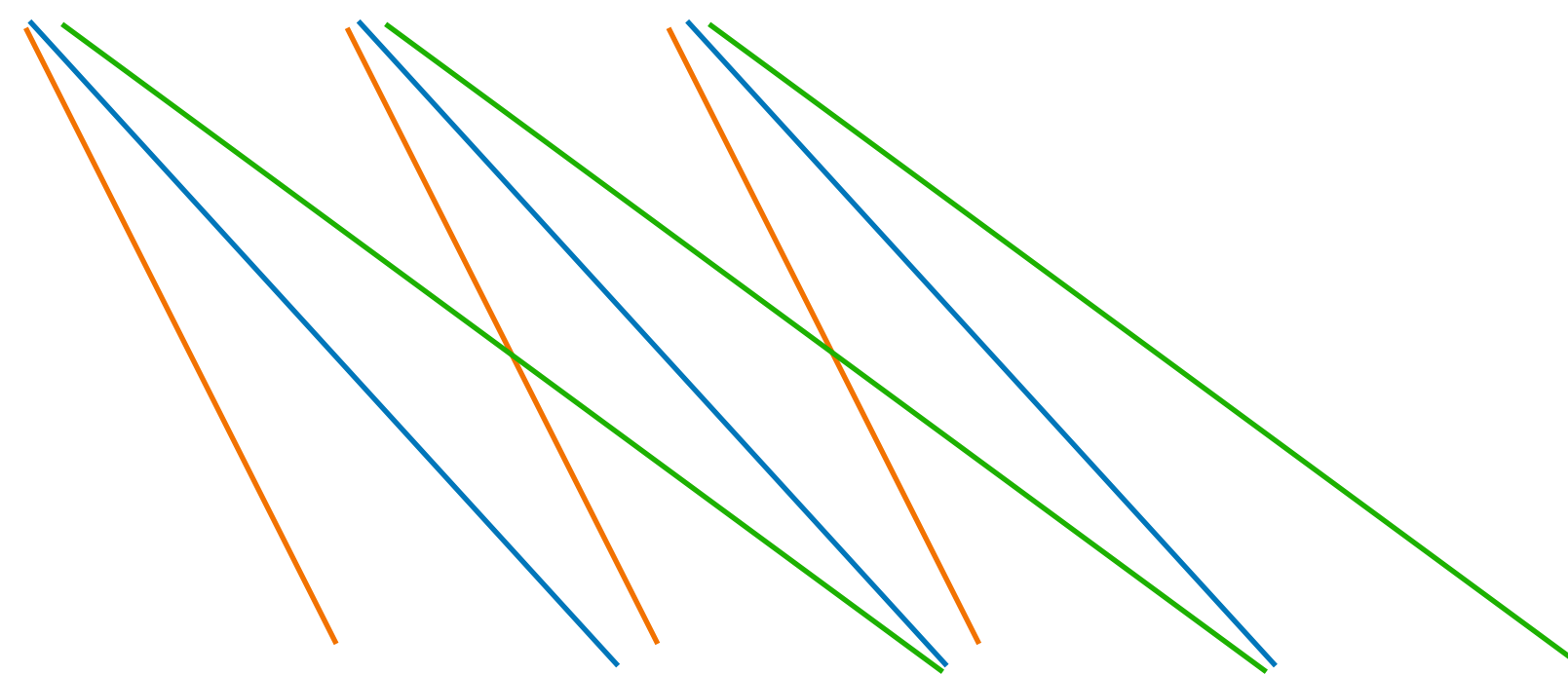
$$(x[0]x[1] + x[1]x[2] + x[2]x[3]) / 3$$

$\tau = 2$

$$(x[0]x[2] + x[1]x[3] + x[2]x[4]) / 3$$

$\tau = 3$

$$(x[0]x[3] + x[1]x[4] + x[2]x[5]) / 3$$



分散の定義と計算

Sample mean : \bar{x}

True mean : μ

Sample variance : $s^2 = \frac{1}{n} \sum_i (x_i - \bar{x})^2$

True variance : $\sigma^2 = E[(x_i - \mu)^2]$

$$\begin{aligned} s^2 &= \frac{1}{n} \sum (x_i - \bar{x})^2 \\ &= \frac{1}{n} \sum (x_i - \mu + \mu - \bar{x})^2 \\ &= \frac{1}{n} \sum (x_i - \mu)^2 - 2(\bar{x} - \mu) \frac{1}{n} \sum (x_i - \mu) + (\bar{x} - \mu)^2 \\ &= \frac{1}{n} \sum (x_i - \mu)^2 - 2(\bar{x} - \mu) \frac{1}{n} (n\bar{x} - n\mu) + (\bar{x} - \mu)^2 \\ &= \frac{1}{n} \sum (x_i - \mu)^2 - (\bar{x} - \mu)^2 \end{aligned}$$

分散の定義と計算

Sample mean : \bar{x}

True mean : μ

Sample variance : $s^2 = \frac{1}{n} \sum_i (x_i - \bar{x})^2$

True variance : $\sigma^2 = E[(x_i - \mu)^2]$

$$\begin{aligned} E[s^2] &= E\left[\frac{1}{n} \sum (x_i - \mu)^2 - (\bar{x} - \mu)^2\right] \\ &= \frac{1}{n} \sum E[(x_i - \mu)^2] - E[(\bar{x} - \mu)^2] \\ &= \sigma^2 - E[(\bar{x} - \mu)^2] \\ &= \sigma^2 - E\left[\left\{\frac{1}{n}(x_1 + x_2 + \dots - n\mu)\right\}^2\right] \\ &= \sigma^2 - \frac{1}{n^2} \sum E[(x_i - \mu)^2] \\ &= \sigma^2 - \frac{1}{n} \sigma^2 = \frac{n-1}{n} \sigma^2 \end{aligned}$$