

# Deducing cellular processes from cell-size distributions

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My group focuses on

The gas tank (a.k.a.  
adipose tissue)  
and the generator  
(a.k.a. mitochondria)

Avoid the difficult questions of appetite control  
and addiction

# The gas tank

- Mammals cannot excrete much excess fat directly. Apparently, there was no evolutionary pressure to include such a process.
- Mammals can sequester fat in adipose tissue and use this dense energy source as needed.
- Insulin is a signal for substrate switching, shutting off lipolysis and facilitating glucose entry into tissues that require active transport.

# What happens when the gas tank overflows?

- Can adipose tissue grow indefinitely?
- Apart from mechanical stress on joints and the heart, does stored (= sequestered) fat lead to any pathology?

# Working hypothesis

A dysfunction in fat storage results in an underutilization of glucose and a loss in metabolic flexibility, partly from ineffective insulin signaling. High serum glucose leads to damage in tissues that are unable to control glucose entry.

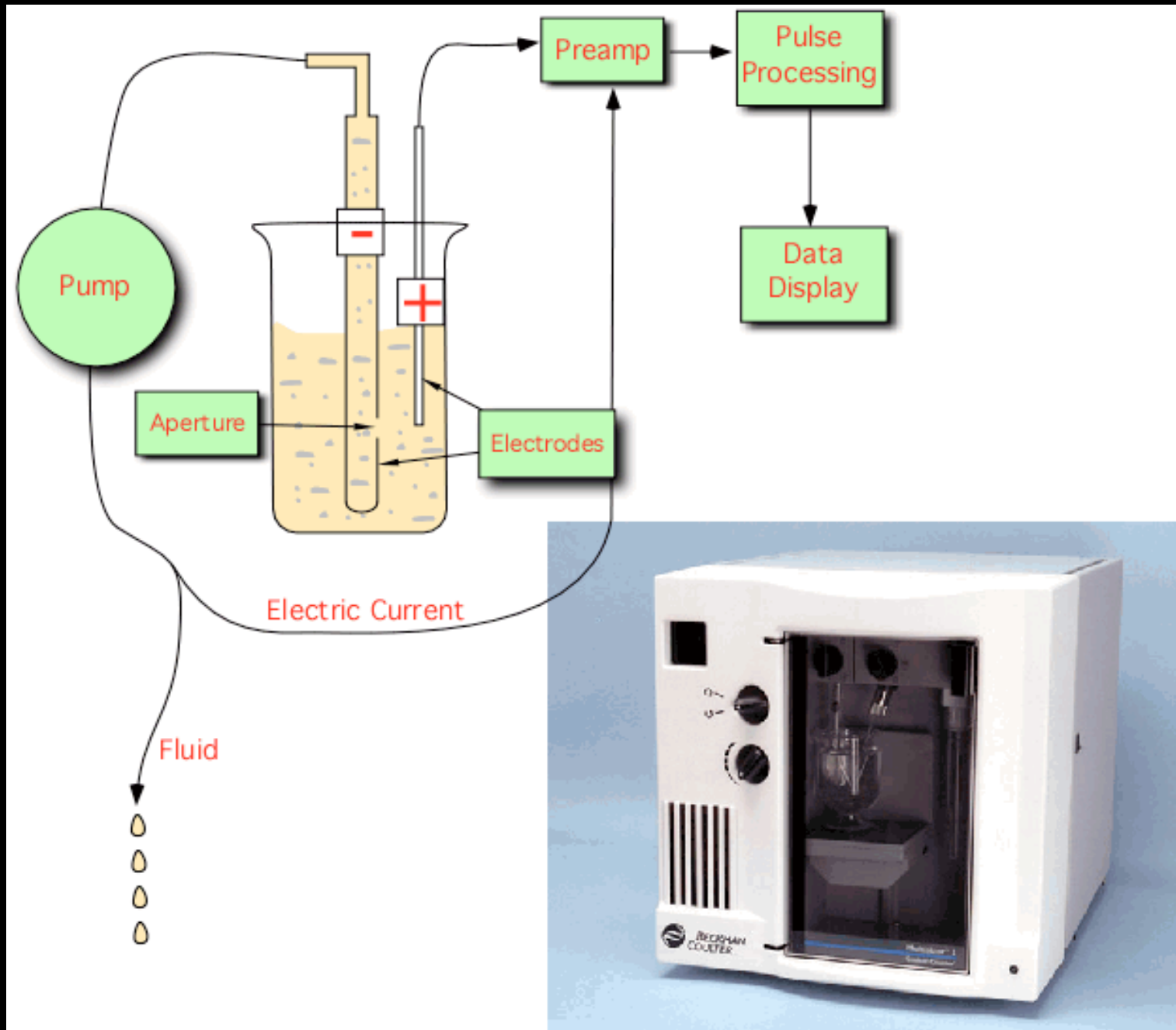
# Adipose tissue growth

The dynamical aspects of adipose tissue growth have been studied but experimental limitations preclude direct observations. We use mathematical modeling of cross-sectional and longitudinal data to understand the processes by which adipose tissue grows.

How can our body regulate energy storage capacity (fat pads) for a given diet?

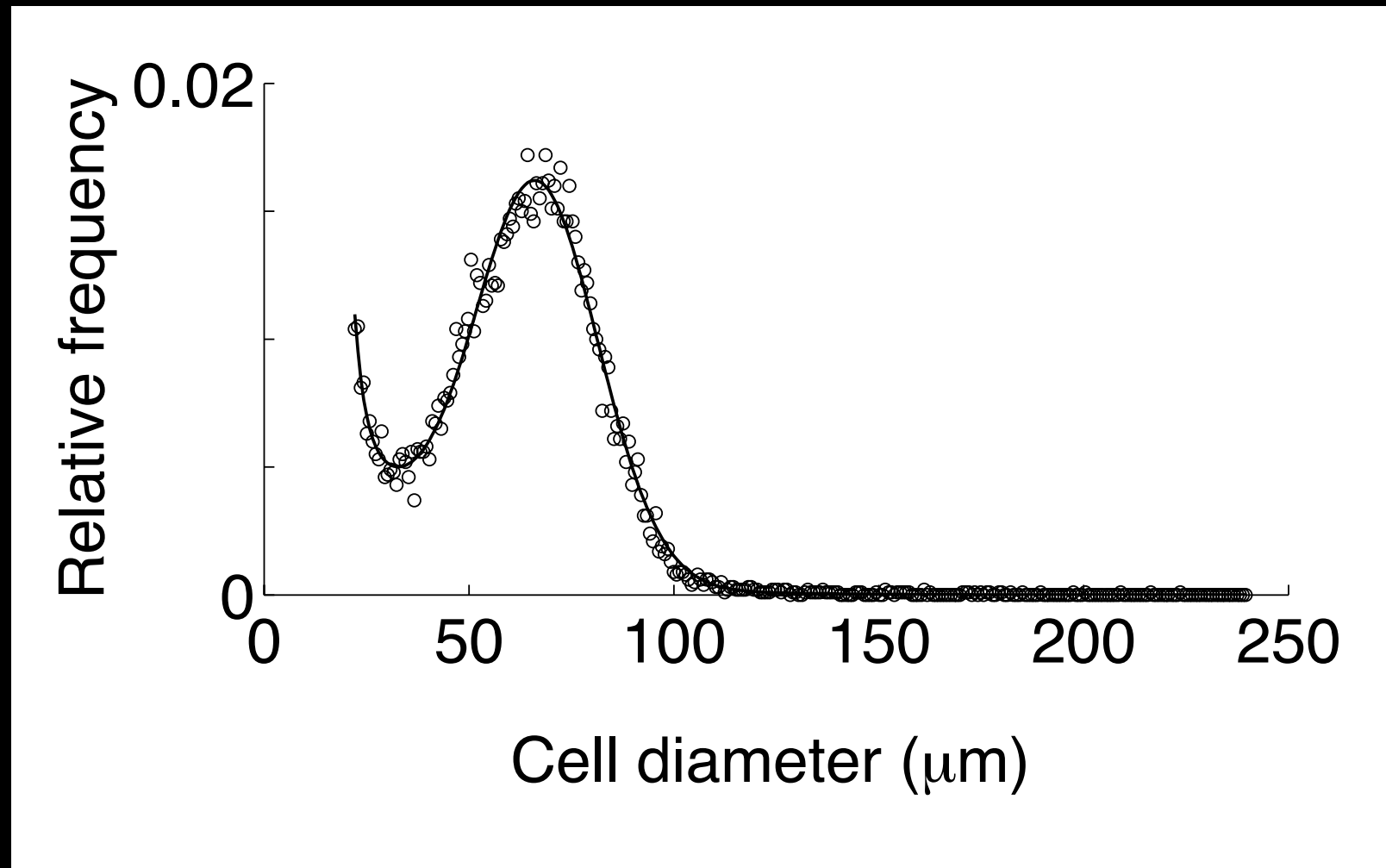
- **plasticity** of adipose cell number (under weight gain and loss)
- **differences** between fat depots (visceral vs. subcutaneous fat)

# Coulter counter





# Cell-size distributions



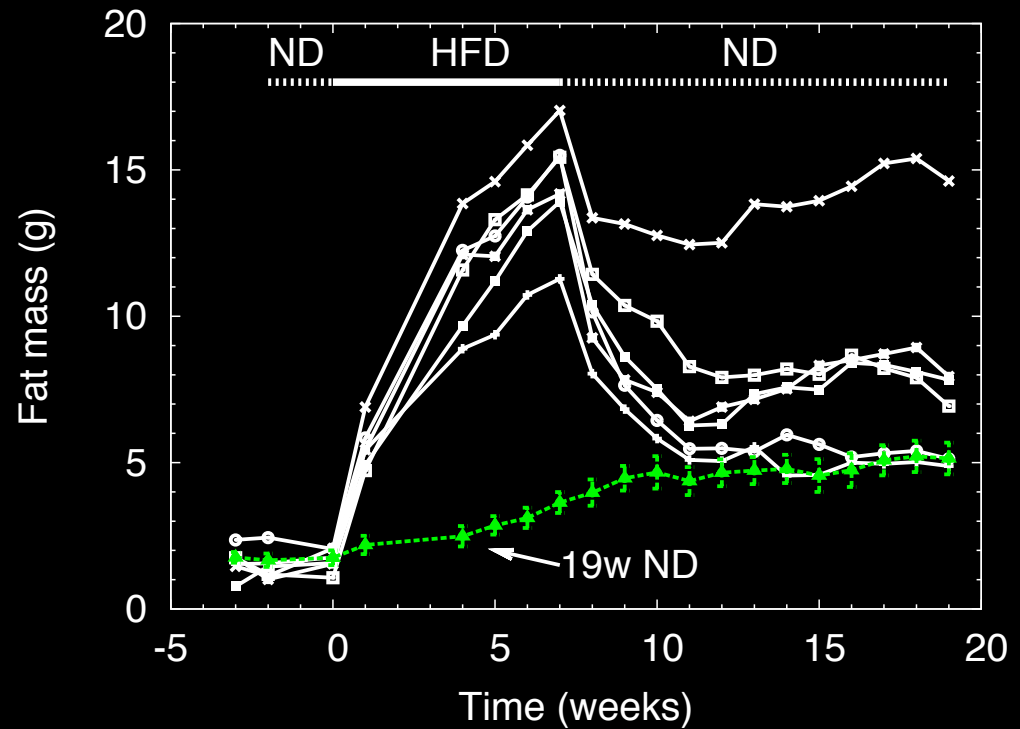
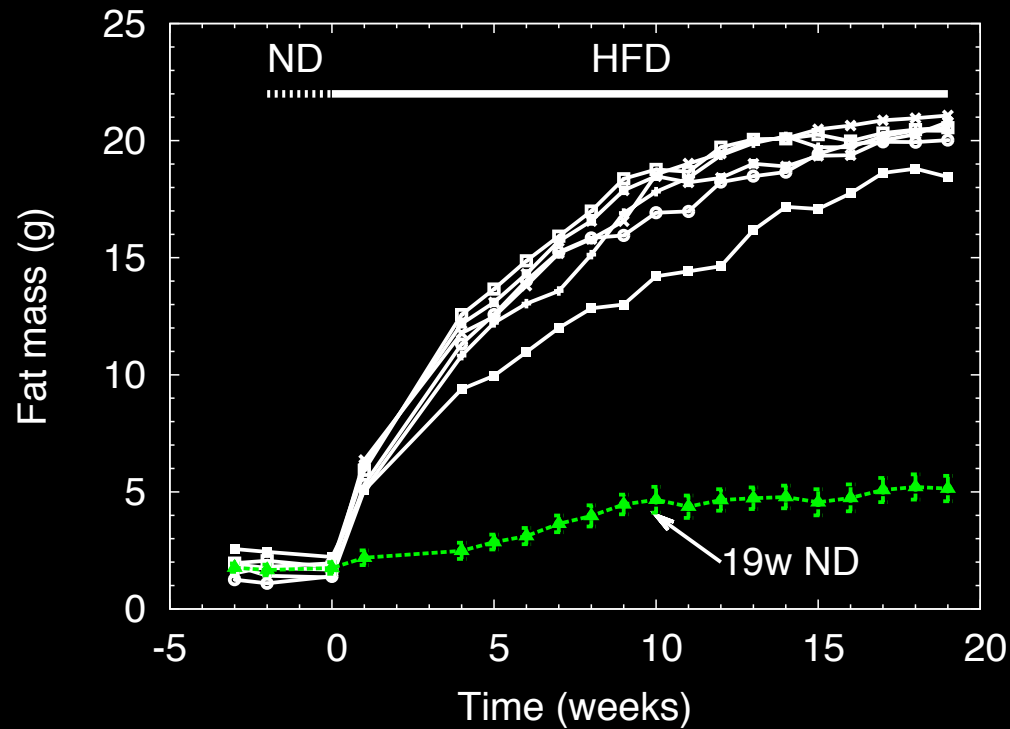
# Data

- Cell-size distributions are probability distributions of cell number as a function of cell diameter.
- The overall cell number can be deduced from the cell-size distribution and the weight of the fat pad, using the density of adipose tissue.

# Moduli problem

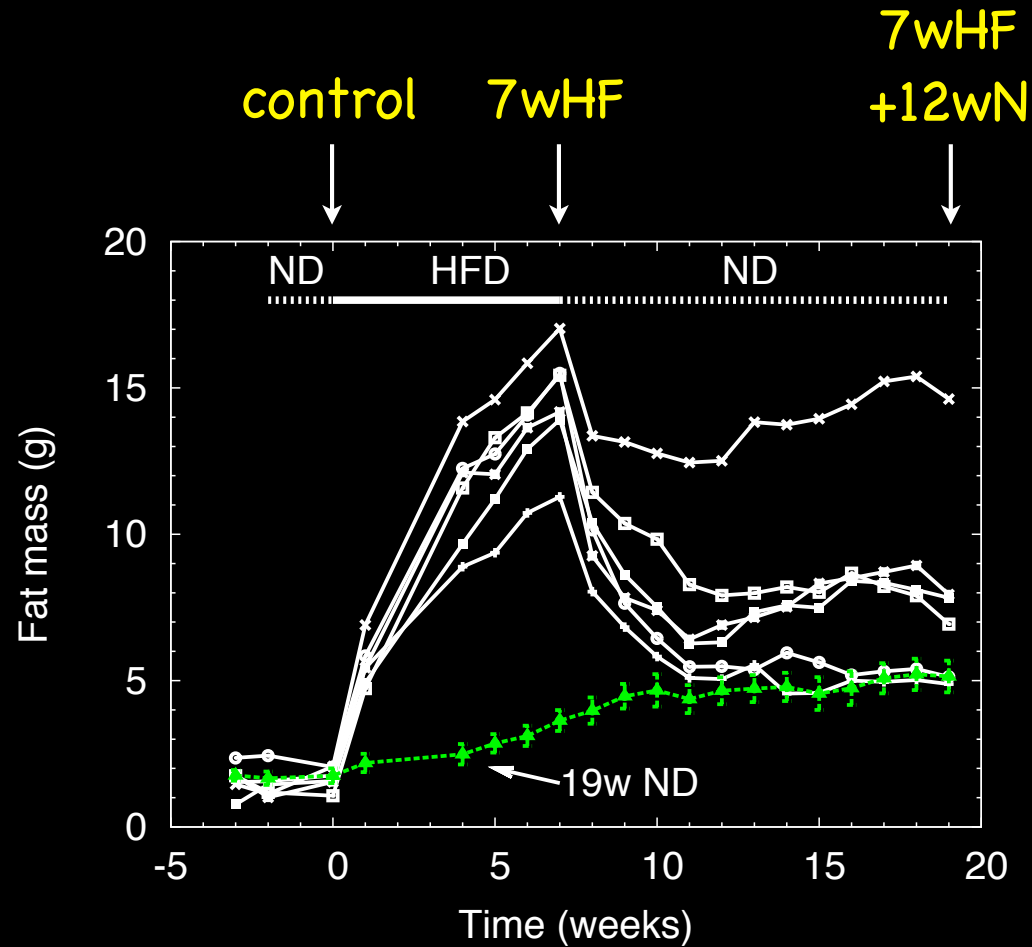
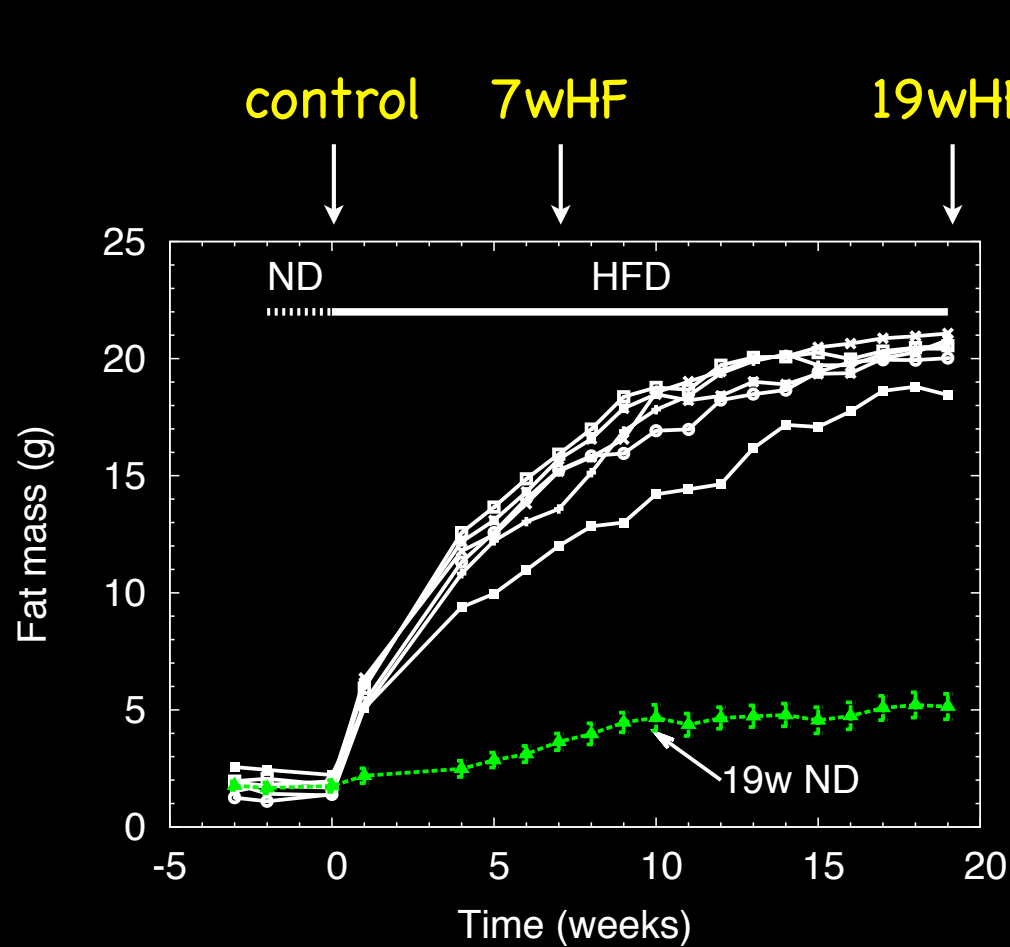
- Smooth parametrization of cell-size distributions
- Kinetics of cell-size distributions
- Dynamics of cell-size distributions

# Energy storage capacity adapts dynamically to diet



(Guo et al., 2009)

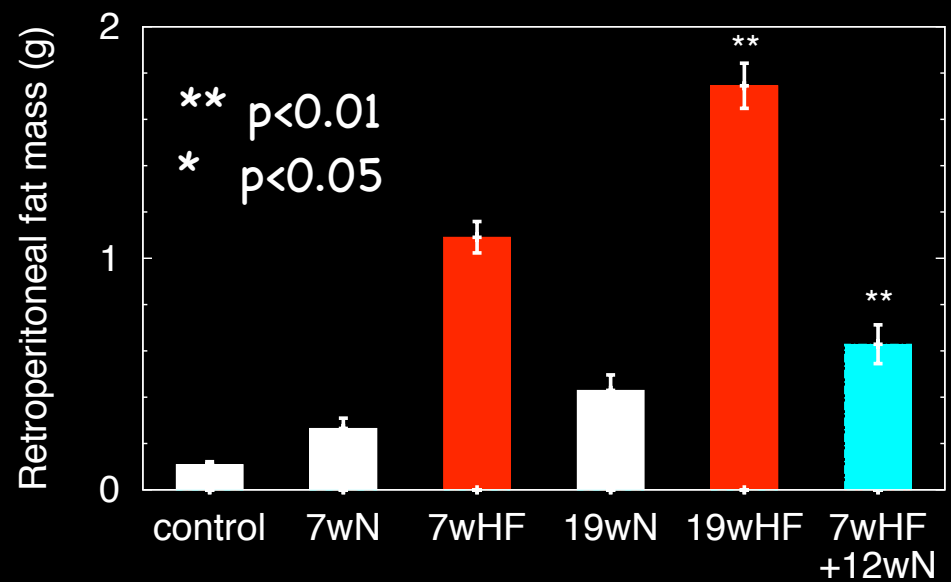
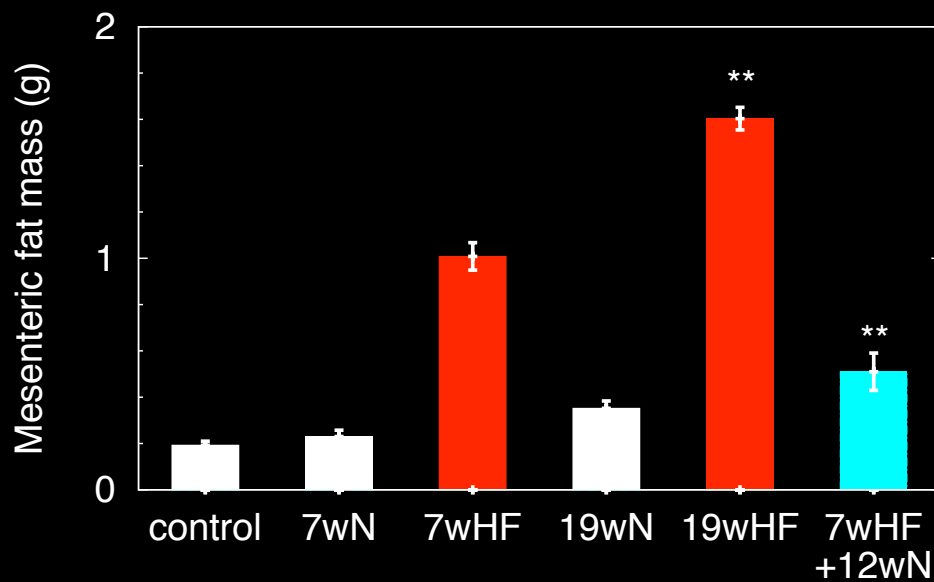
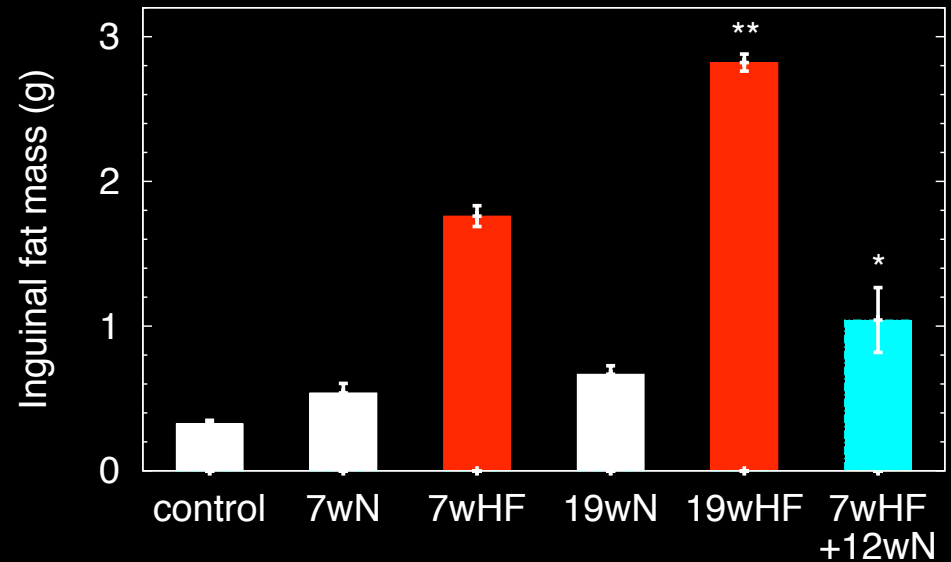
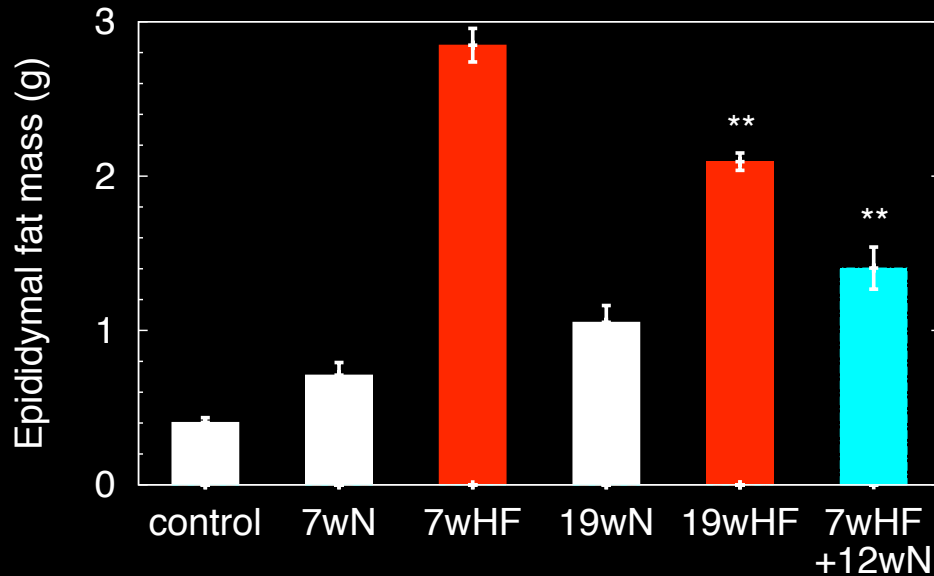
# Energy storage capacity adapts dynamically to diet



fat pad mass and cell-size distributions

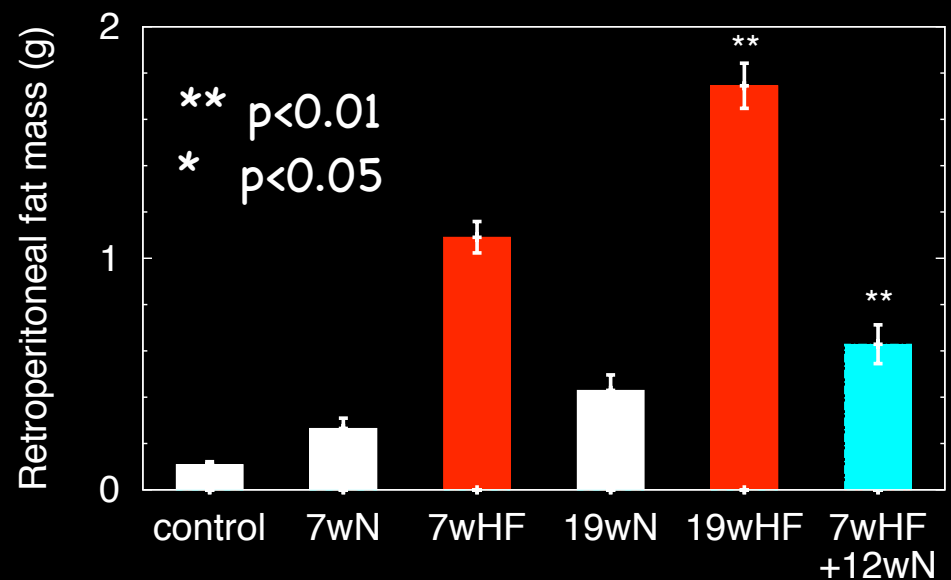
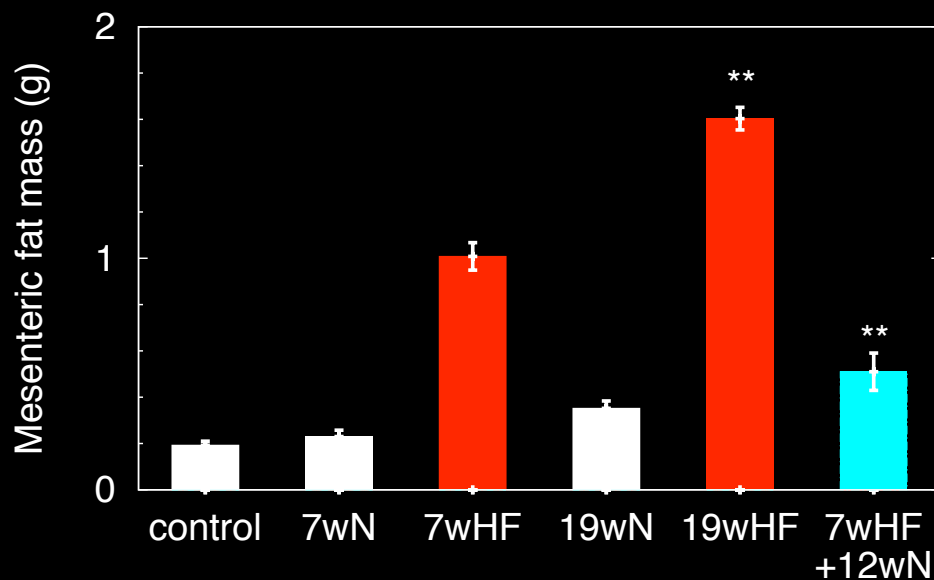
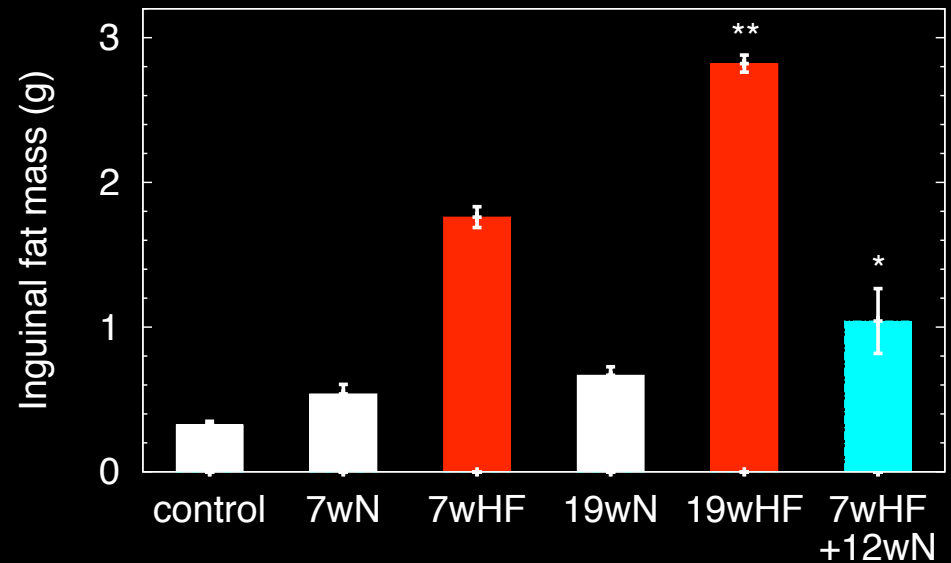
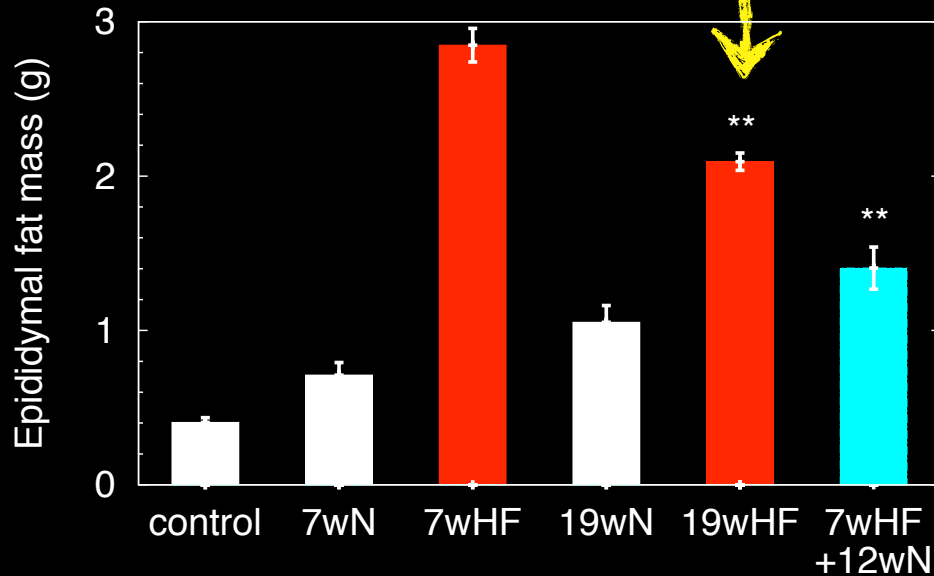
(Guo et al., 2009)

# Only epididymal fat mass decreases under 19w high-fat diet



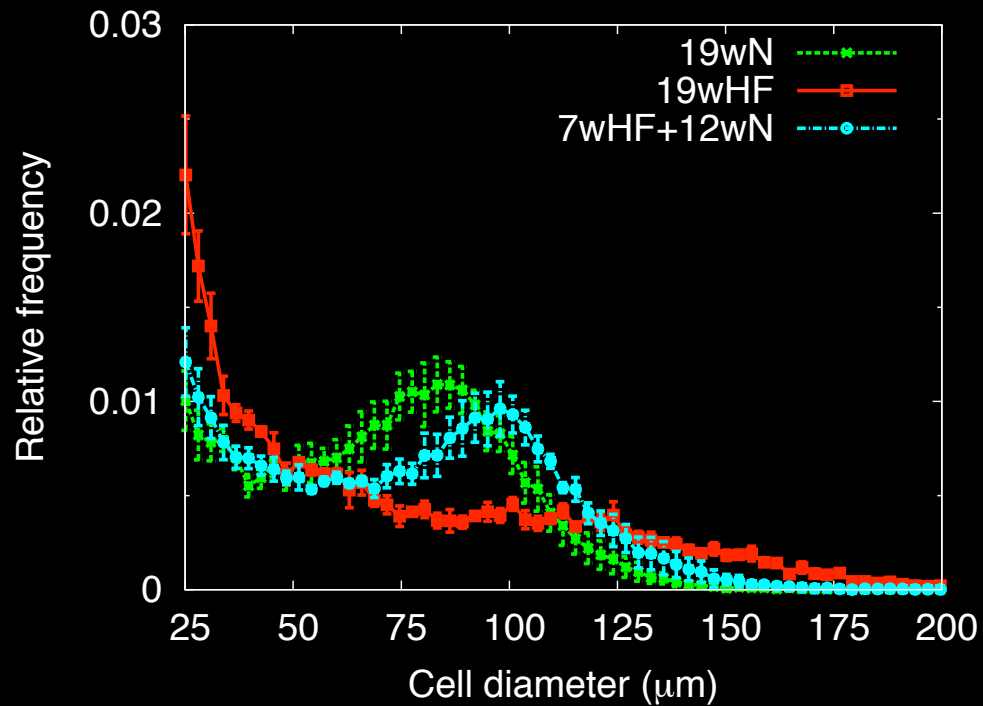
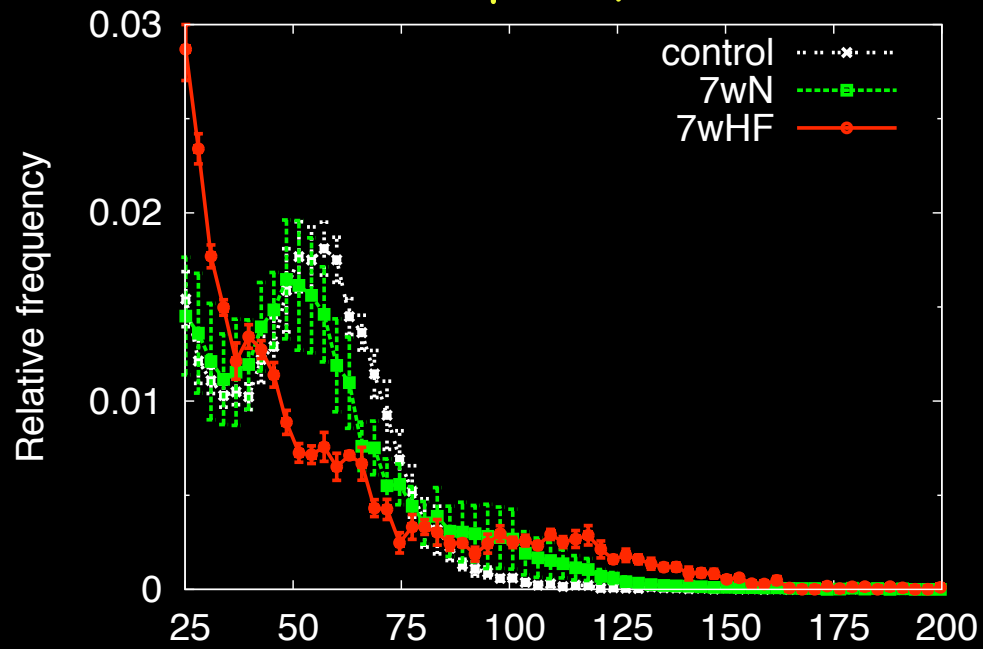
# Only epididymal fat mass decreases under 19w high-fat diet

(Streissel et al., 2007)



# Adipose cell-size distributions

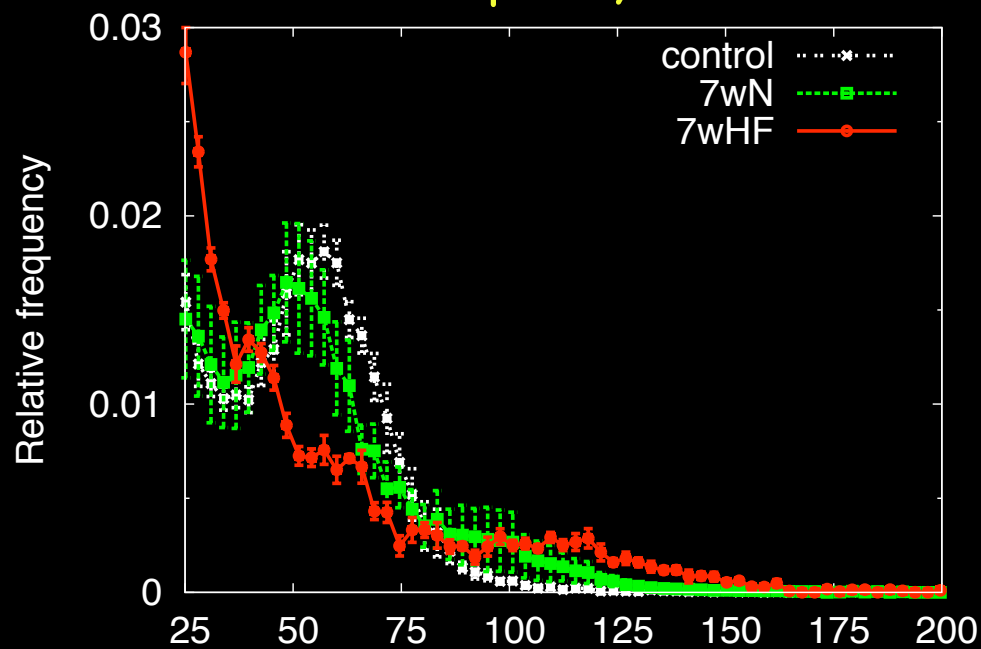
epididymal



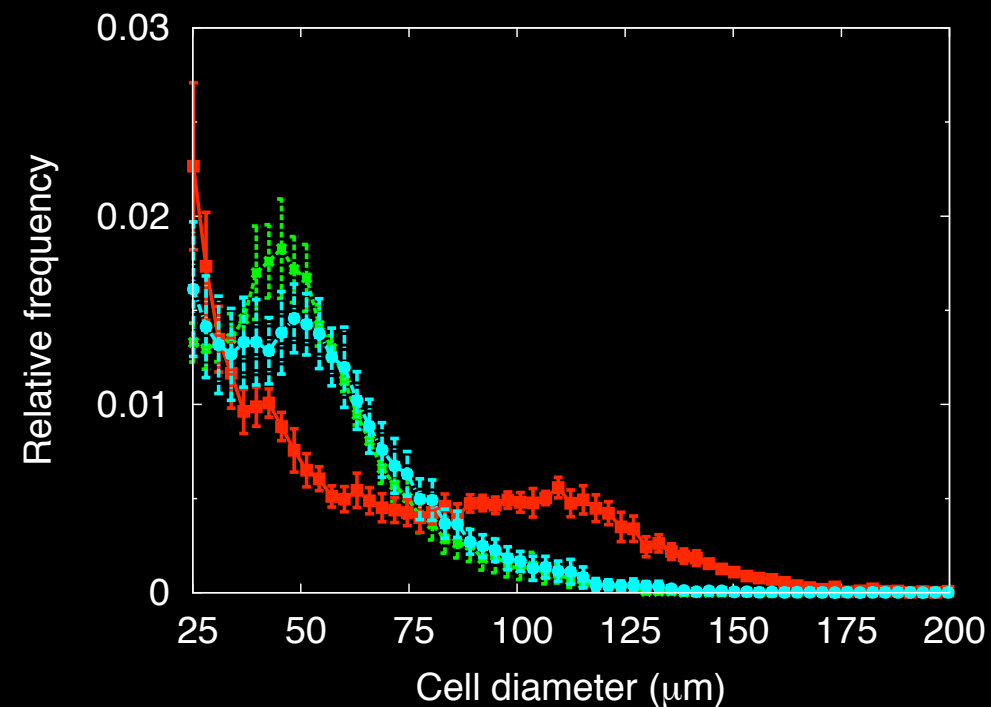
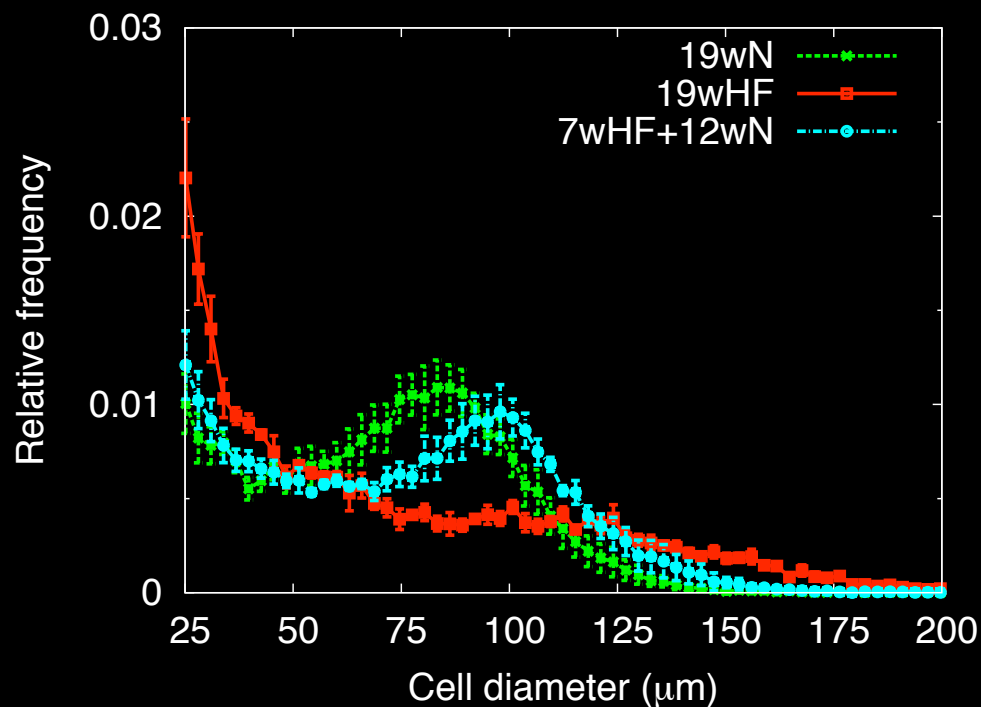
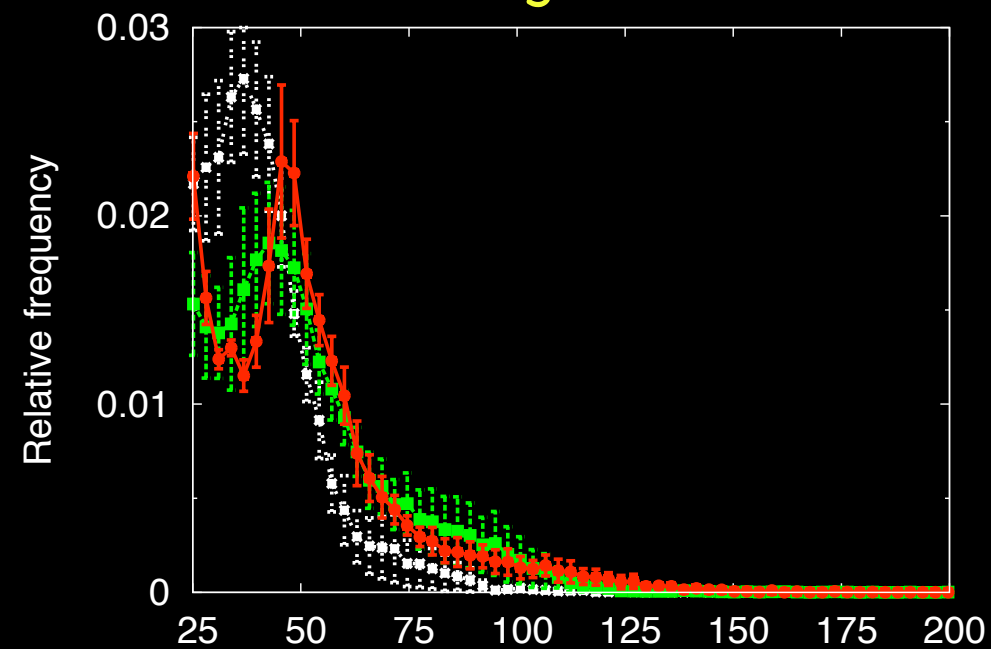


# Adipose cell-size distributions

epididymal

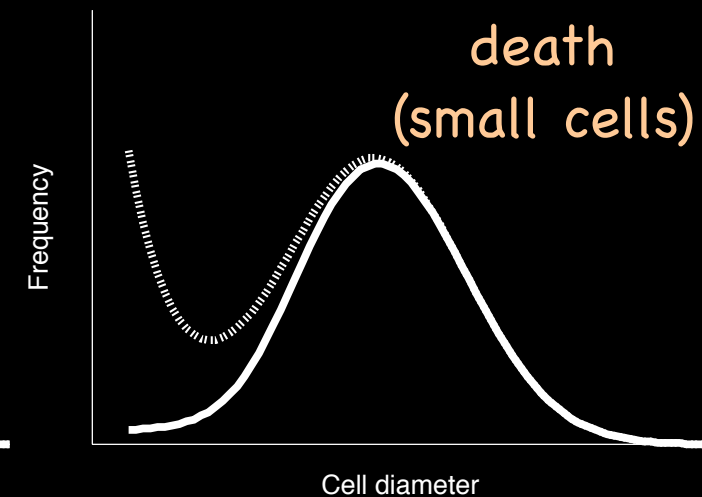
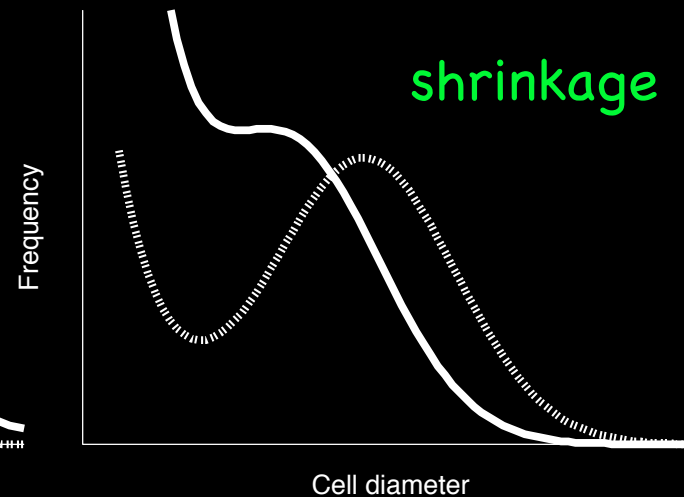
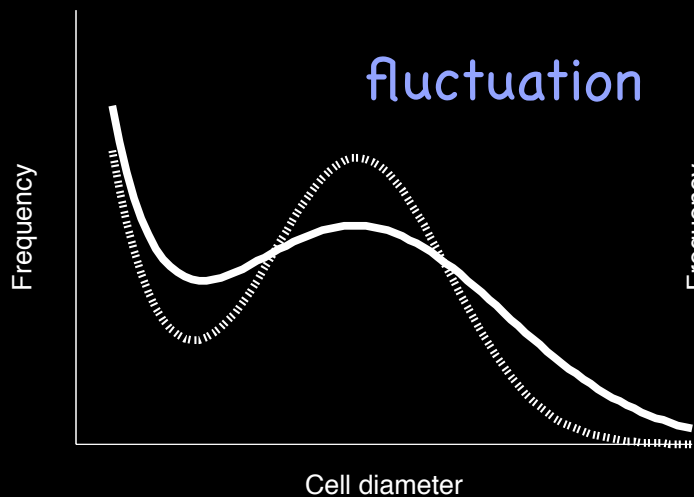
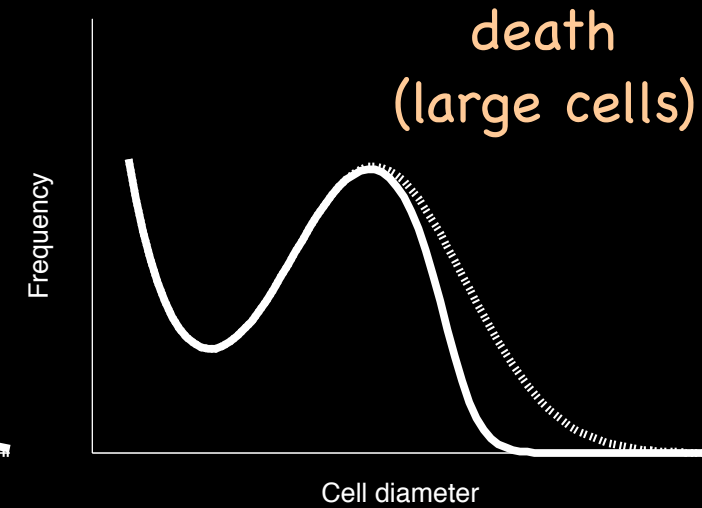
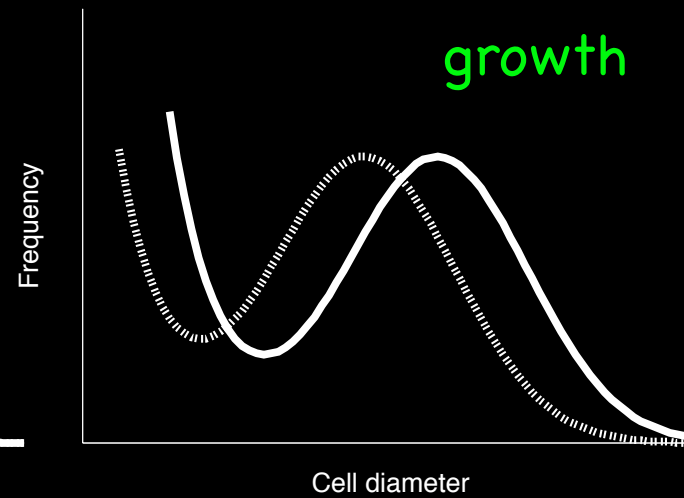
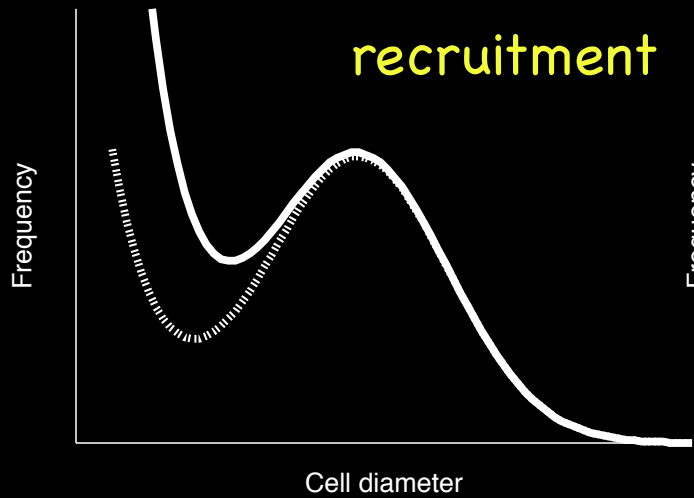


inguinal



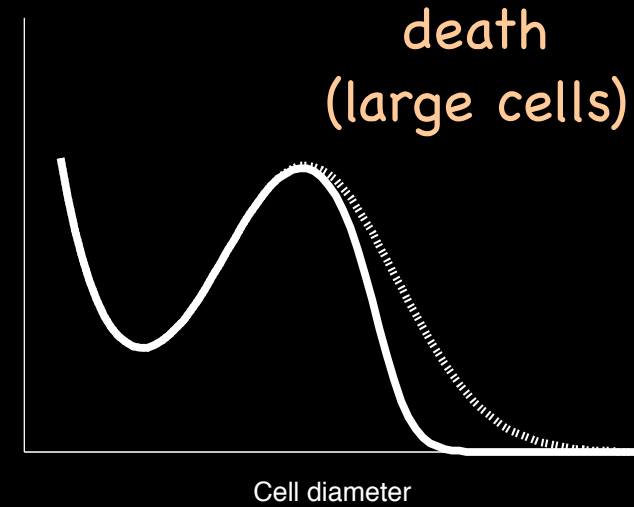
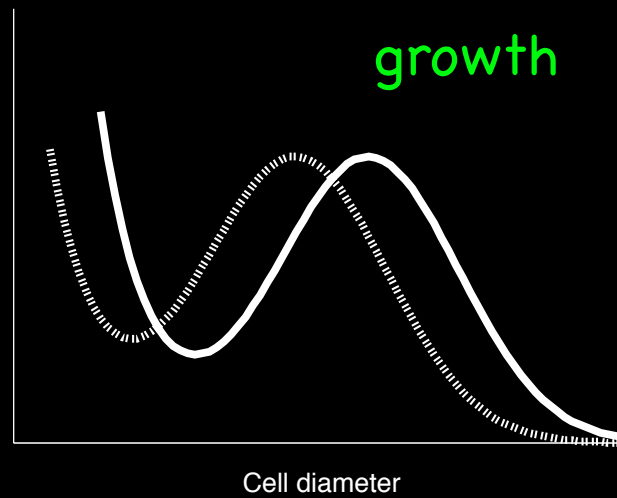
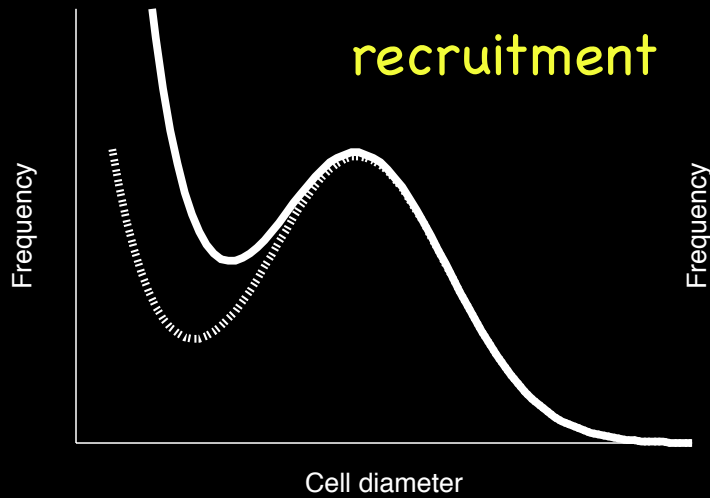
# Multiple biological processes lead to cell-size distributions

(Jo et al., 2009)

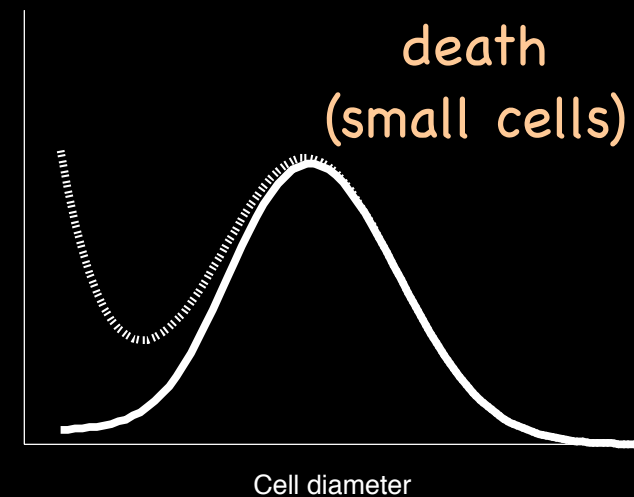
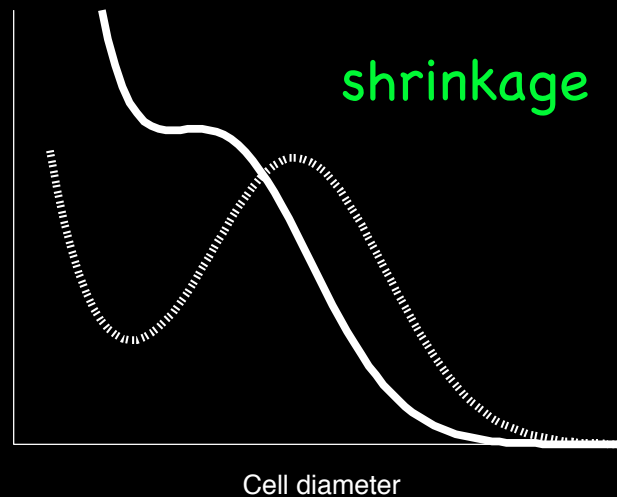
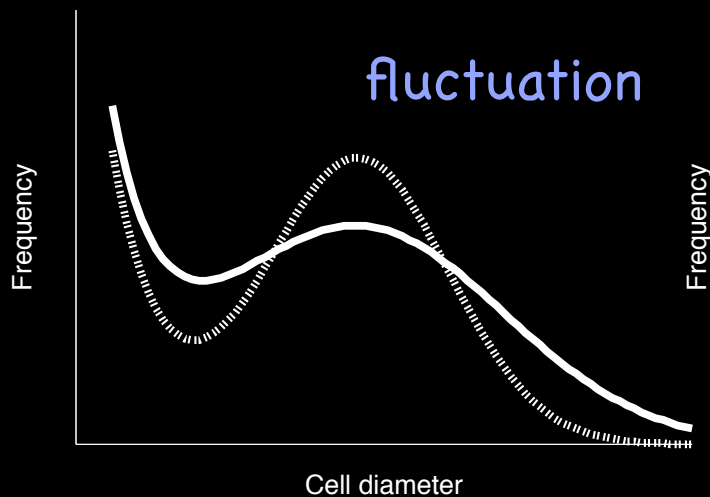


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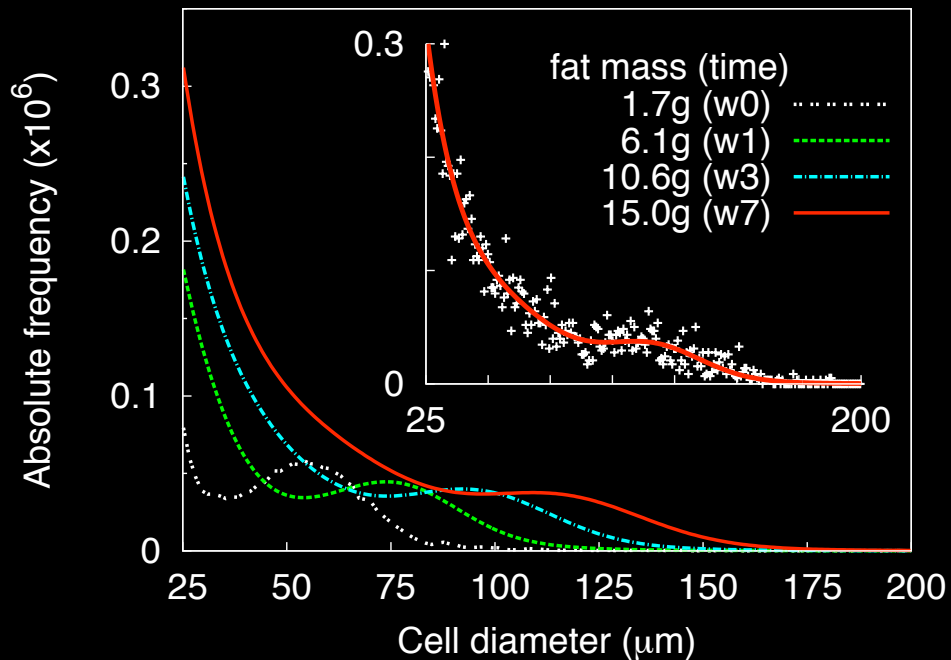
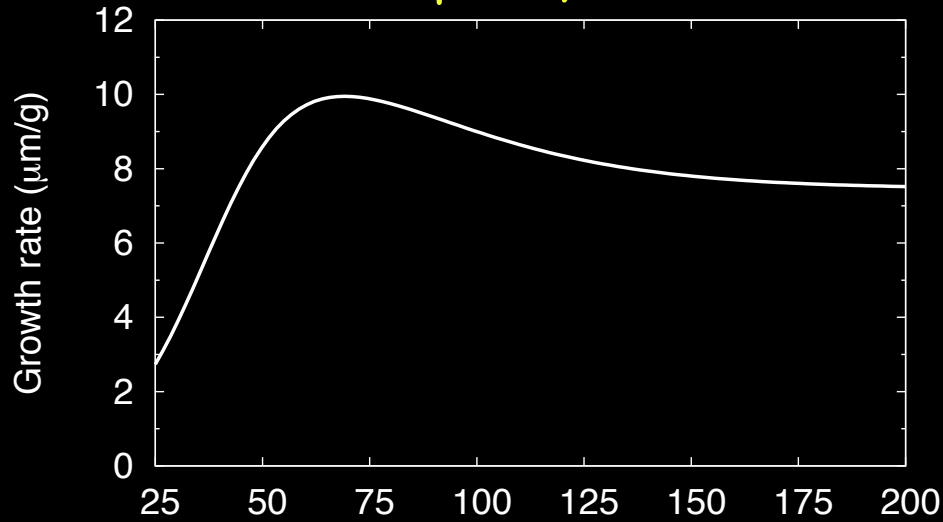


$$\frac{\partial n}{\partial m} = \beta n_p \delta(s - s_0) - v(s) \frac{\partial n}{\partial s} + D \frac{\partial^2 n}{\partial s^2} - k(s) \phi(m) n$$



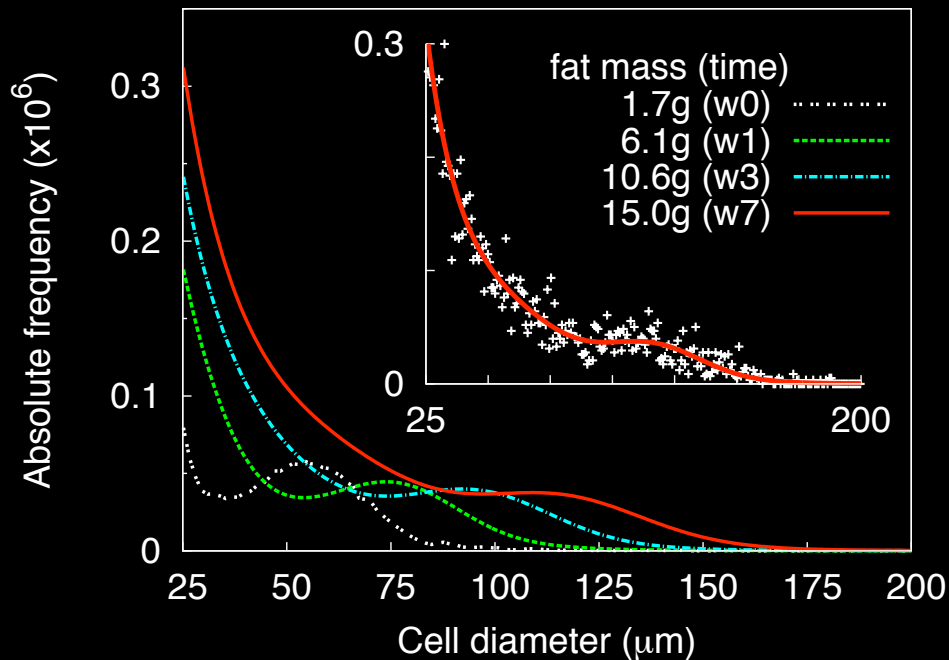
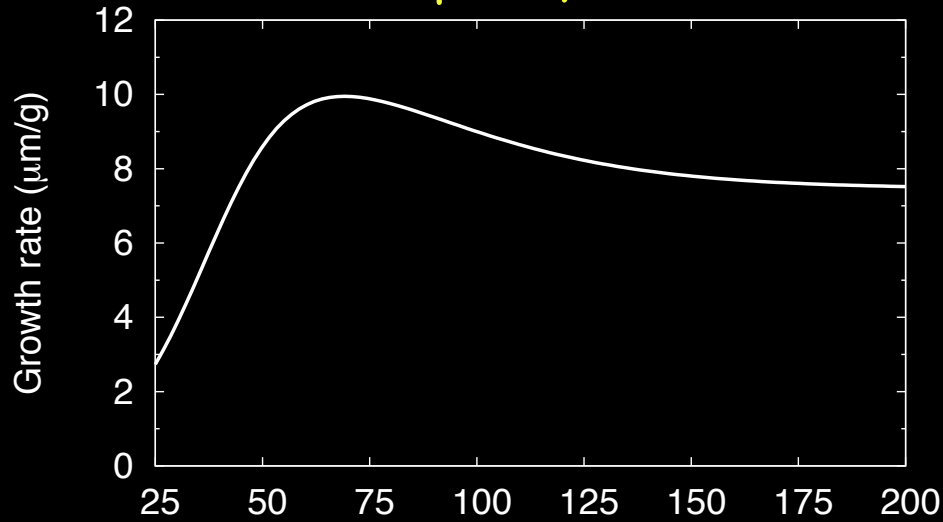
The mathematical model can extrapolate changes of adipose cell-size distributions under **7w HFD**

epididymal

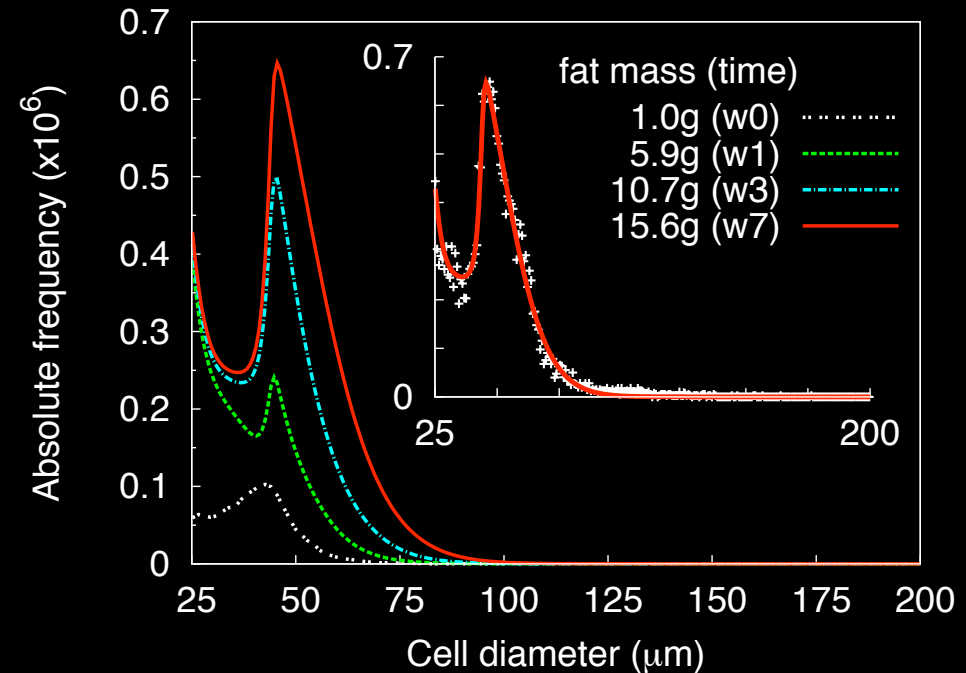
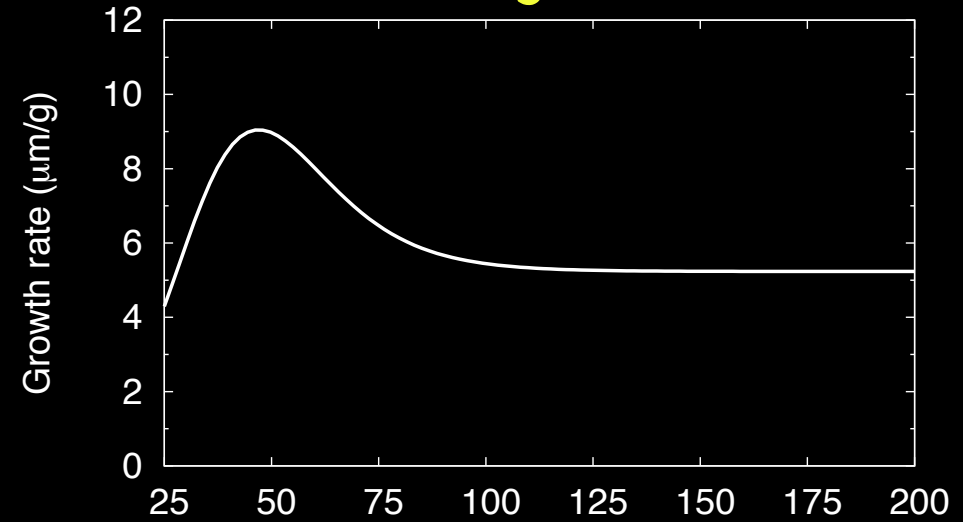


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## epididymal

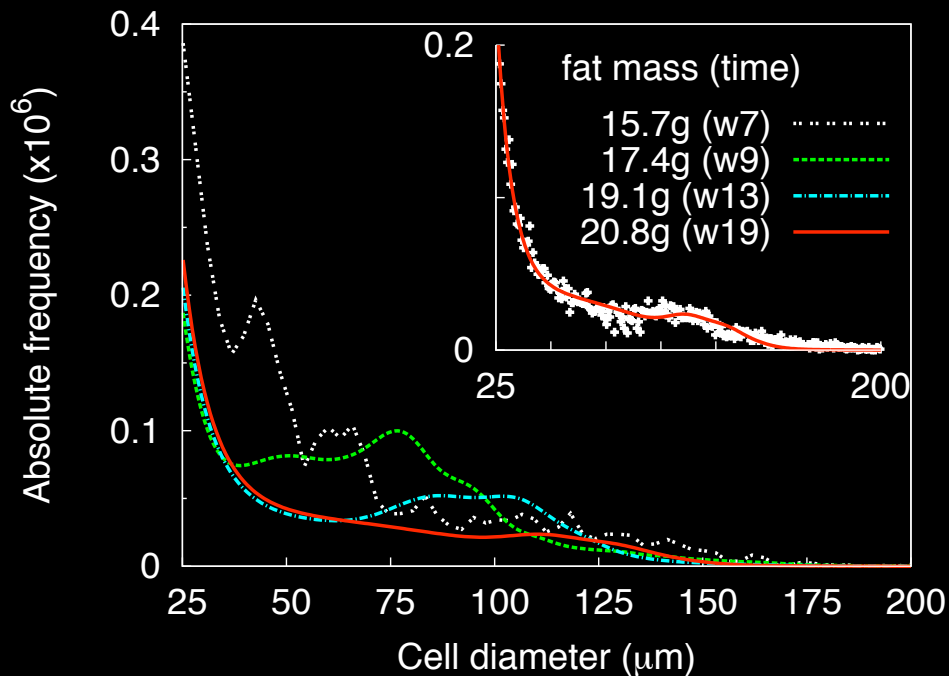
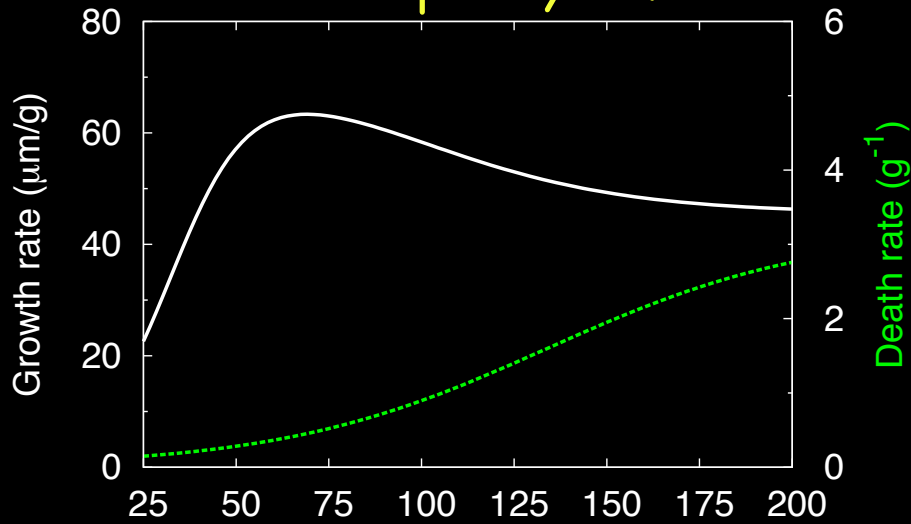


## inguinal



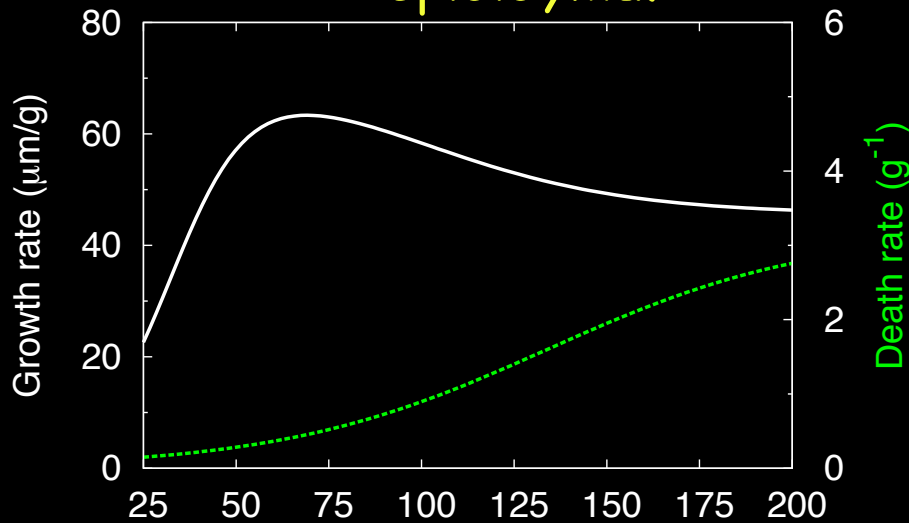
Cell loss of large cells is required to explain the changes of cell-size distributions under **19w HFD**

epididymal

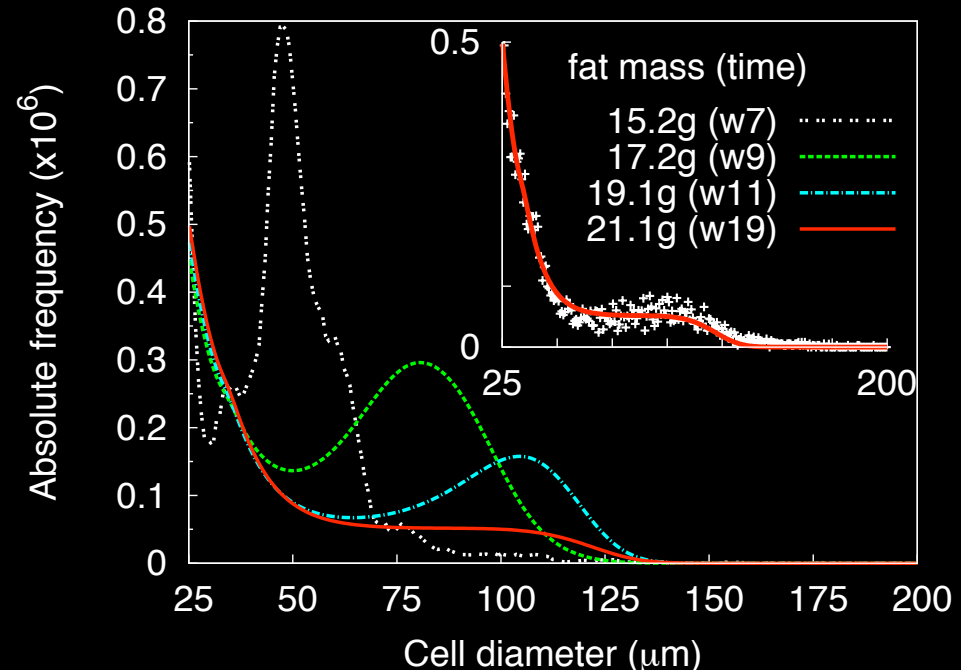
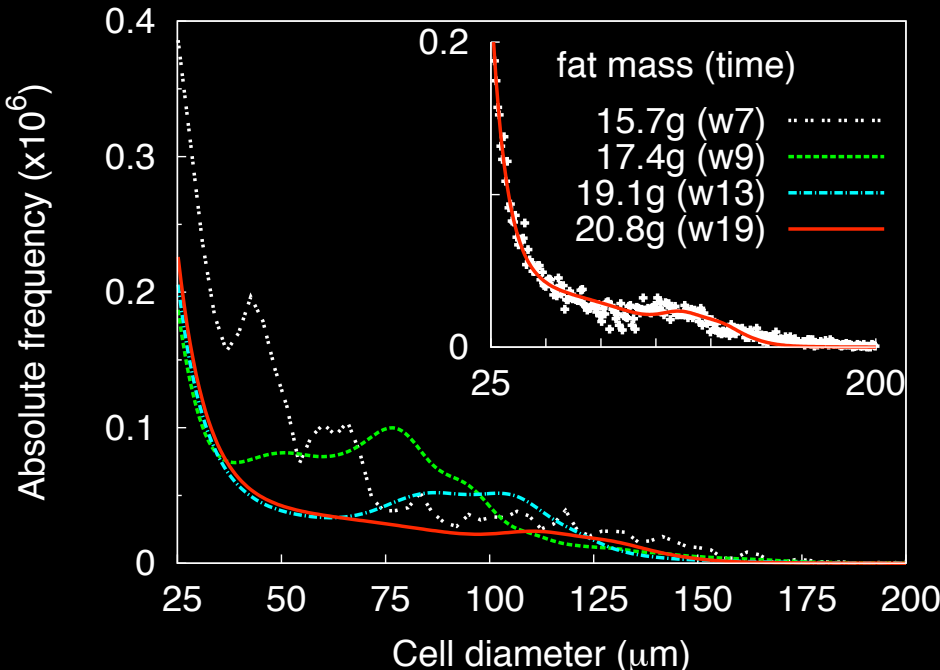
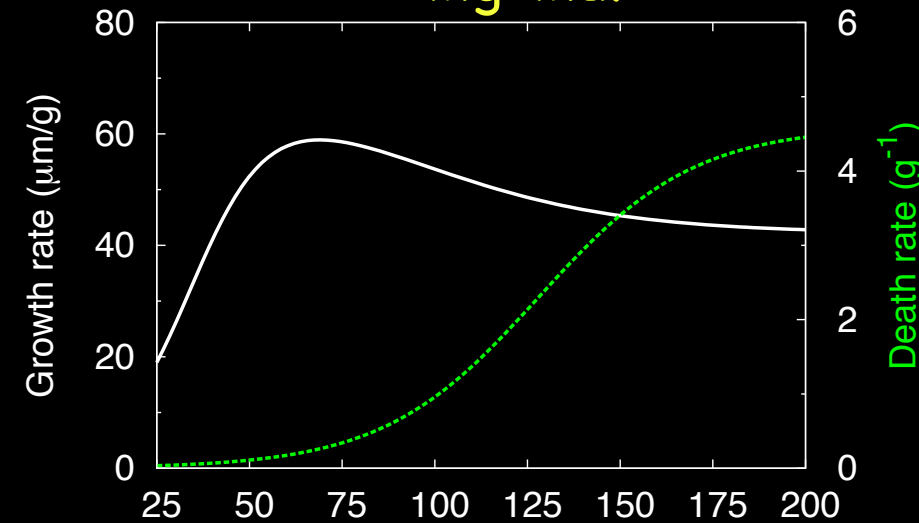


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## epididymal

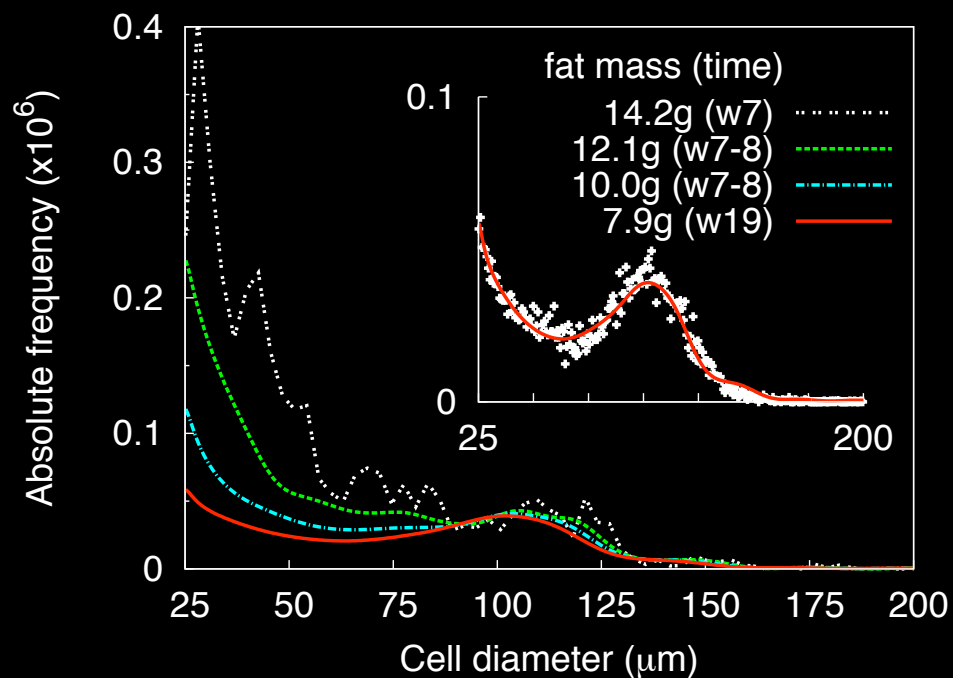
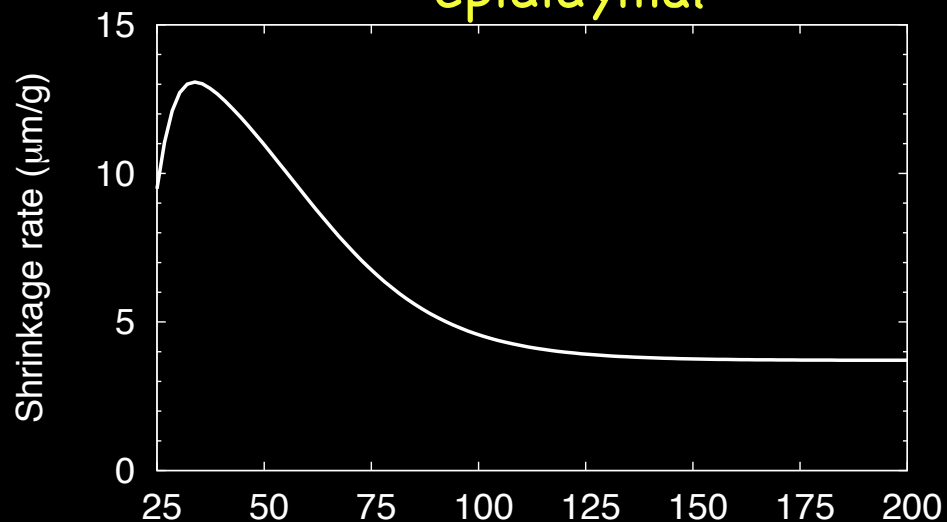


## inguinal



# Size-dependent shrinkage of adipose cells can explain the changes of cell-size distributions under **7wHFD+12wND**

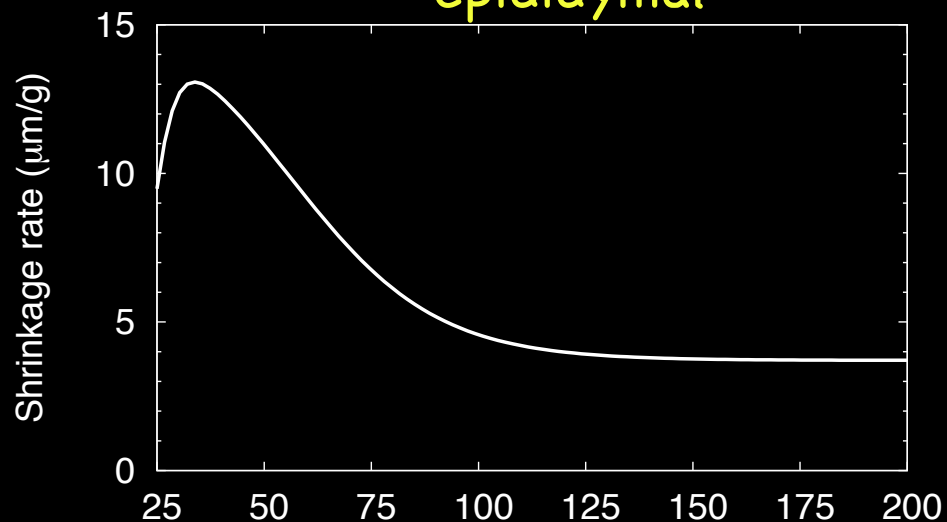
epididymal



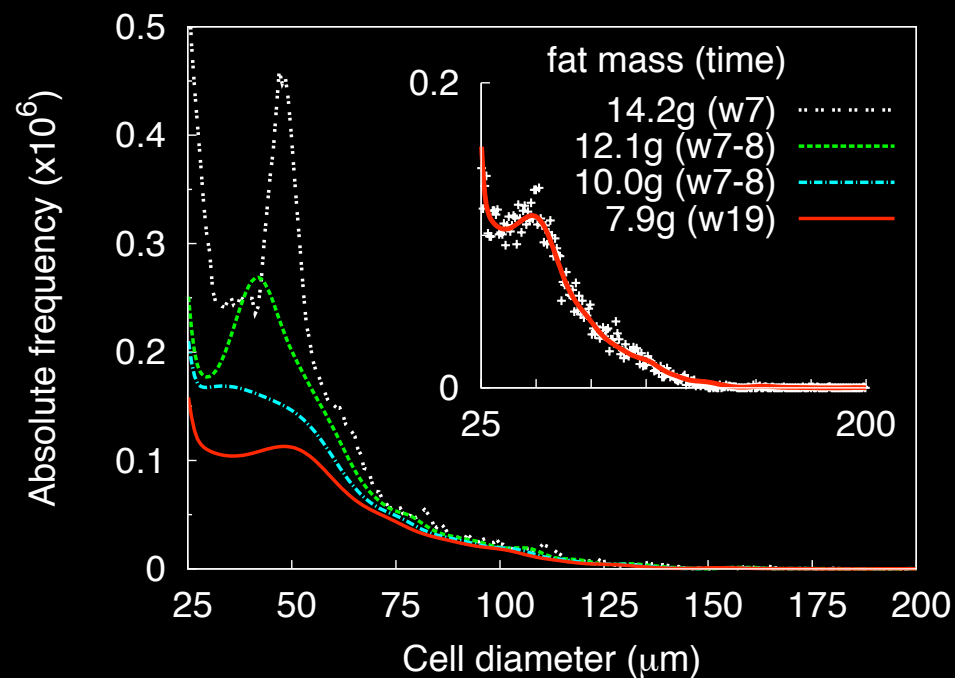
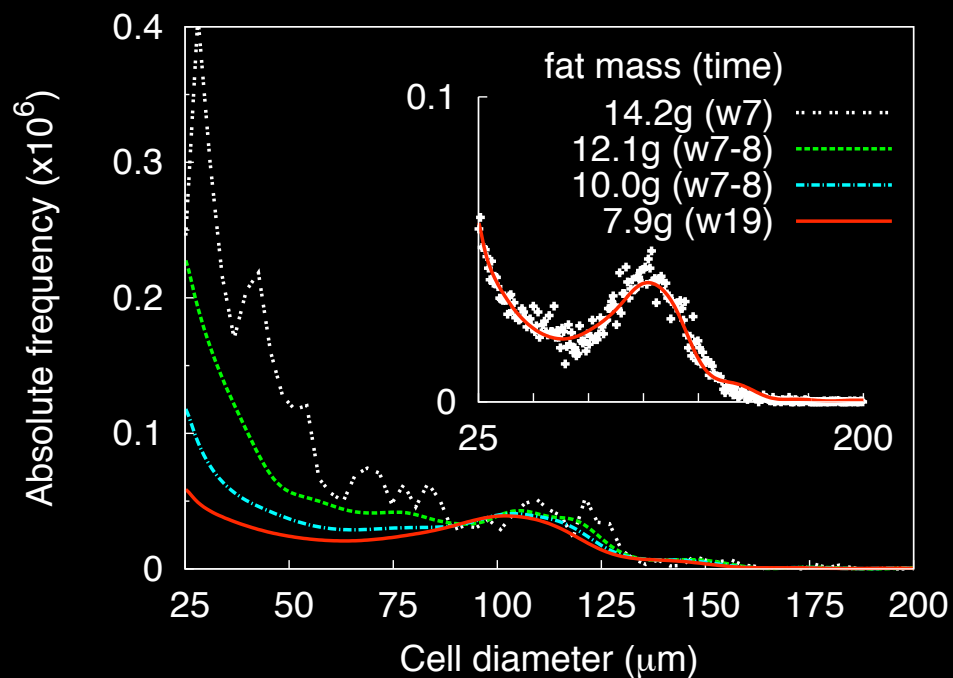
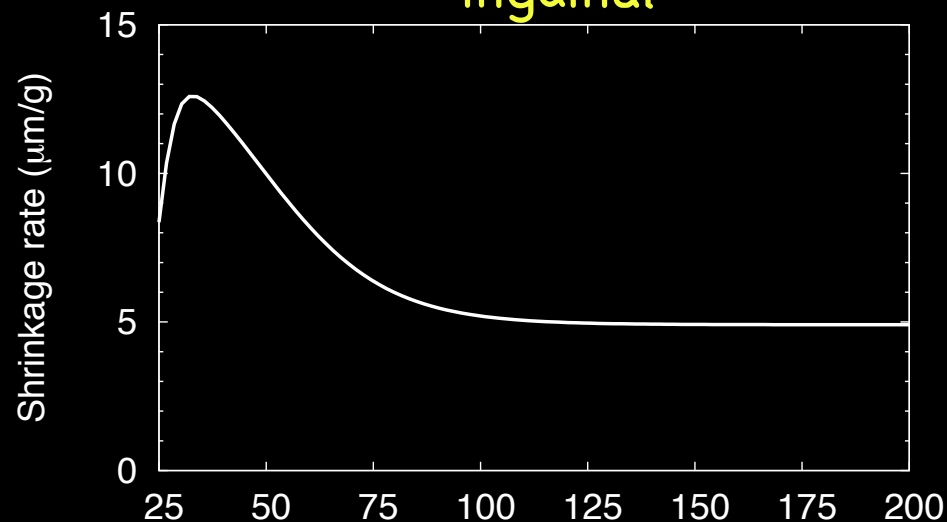


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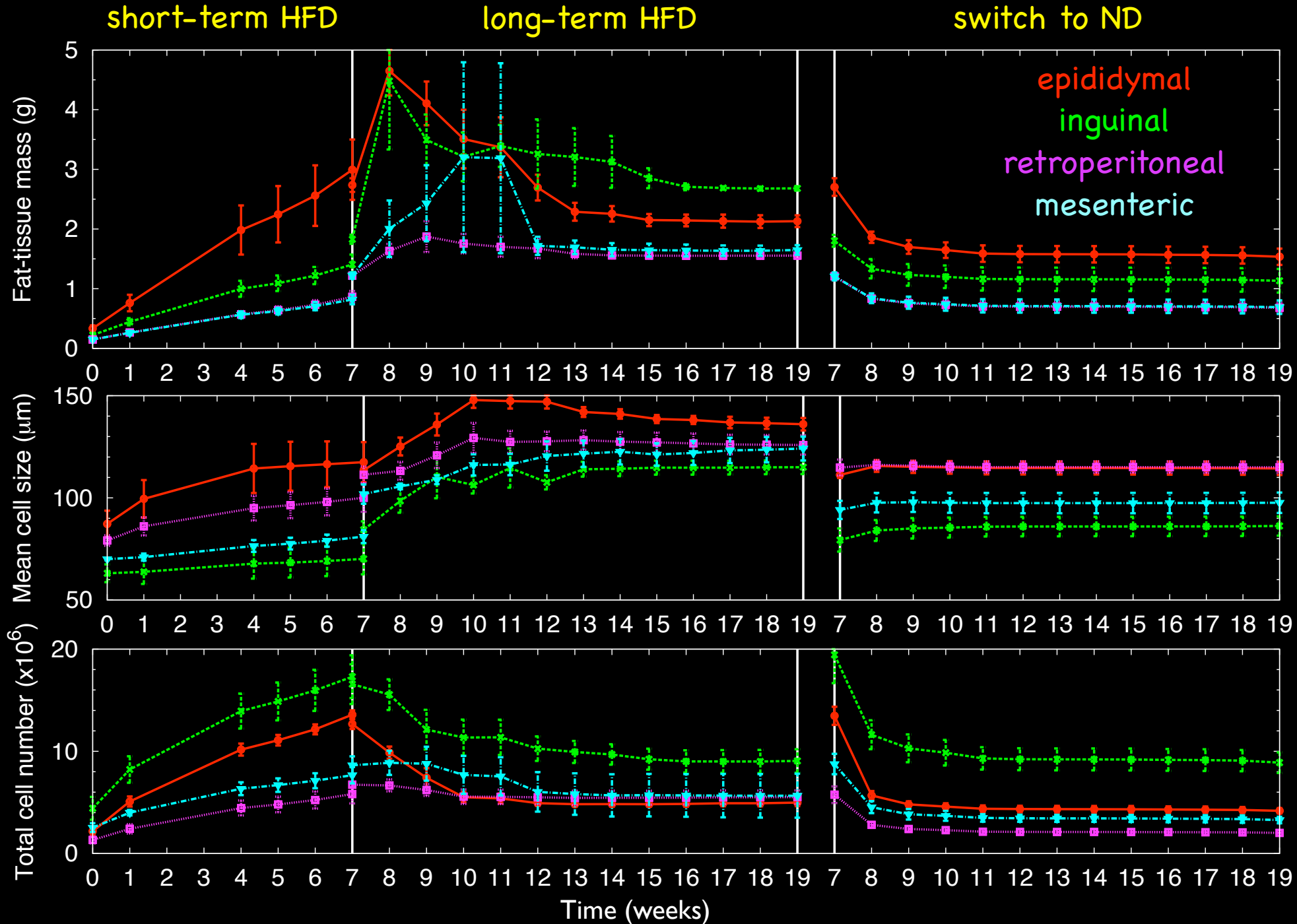
## epididymal



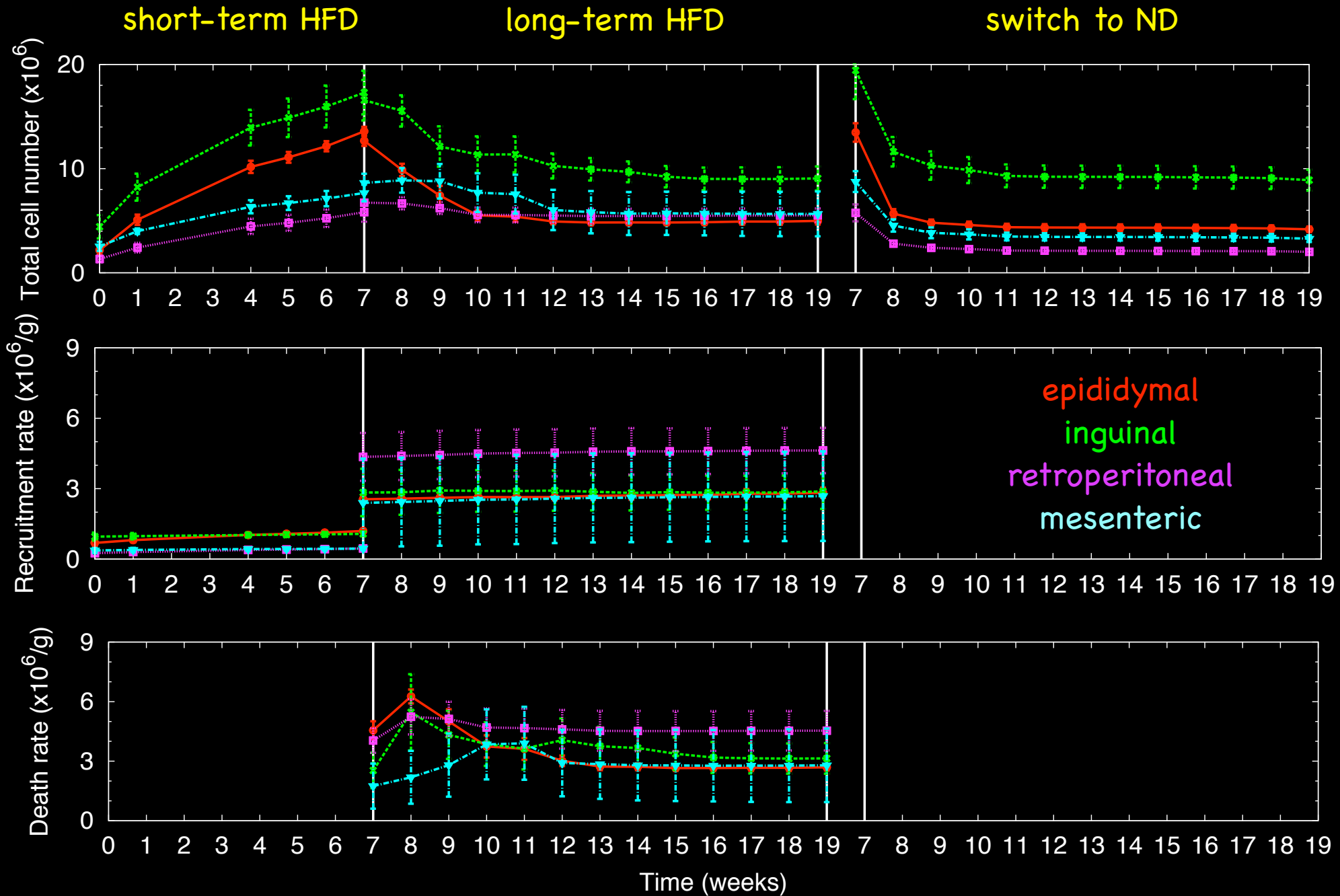
## inguinal



# Mathematical modeling can deduce microscopic changes of adipose tissues



# The model can give the detail of cell number changes in terms of recruitment (influx) and loss (outflux)



# Summary

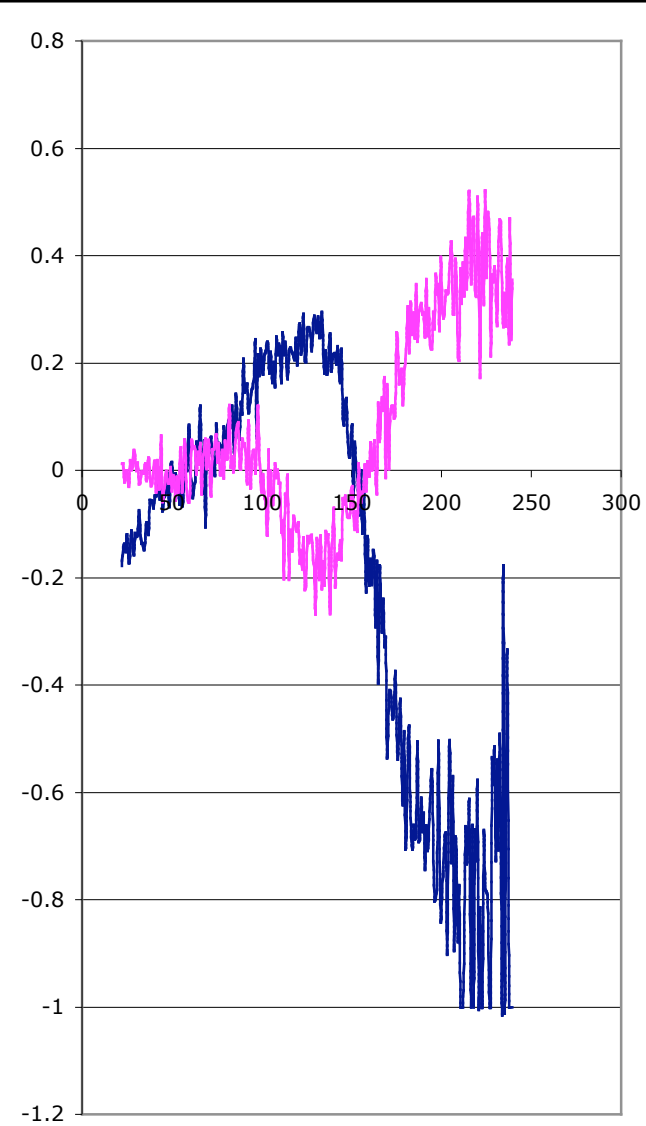
- Fat pads continuously recruited new adipose cells under a high-fat diet.
- Medium-sized adipose cells (30 to 100  $\mu\text{m}$  diameter) showed the largest growth/shrinkage rate under weight gain/loss.
- Cell loss of large adipose cells (above 100  $\mu\text{m}$  diameter) was observed under a prolonged high-fat diet during 19 weeks.
- Adipose tissue responds universally to diet changes to regulate energy storage capacity.

We used an appropriate modulus to find a systematic pattern in cross-sectional mouse data. We expect that there is systematic temporal behavior too. Mice are too small ...

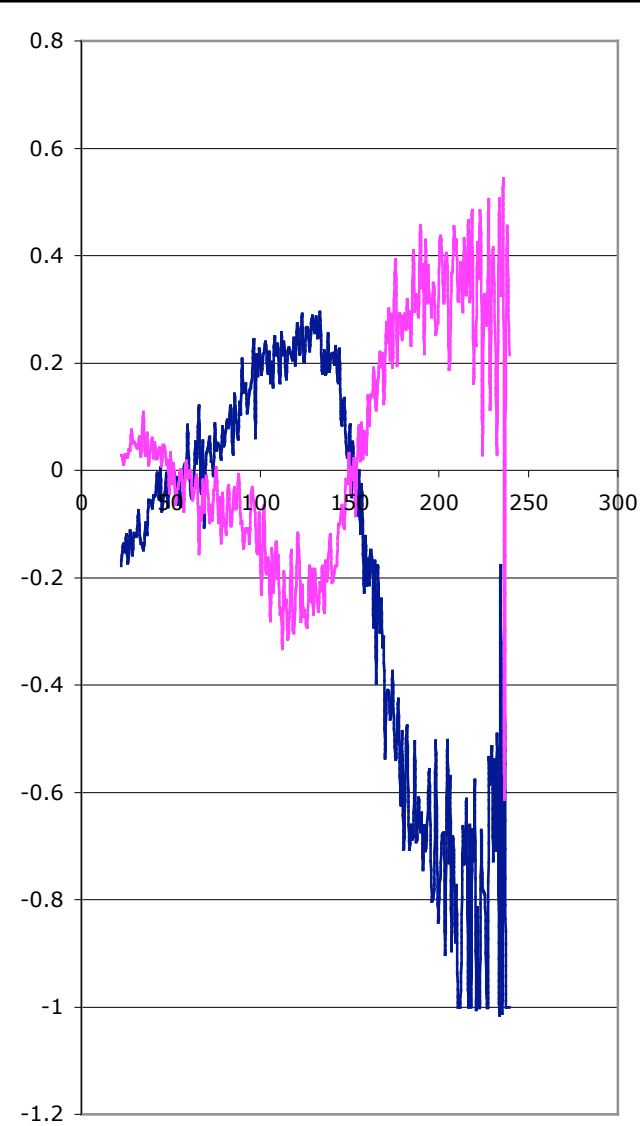
# A longitudinal data set

- Zucker fatty (fa/fa) genetically obese rats
- Microbiopsies at multiple time points over 151 days (rat 1) and 163 days (rat 2)
- Cell-size distributions for each time point

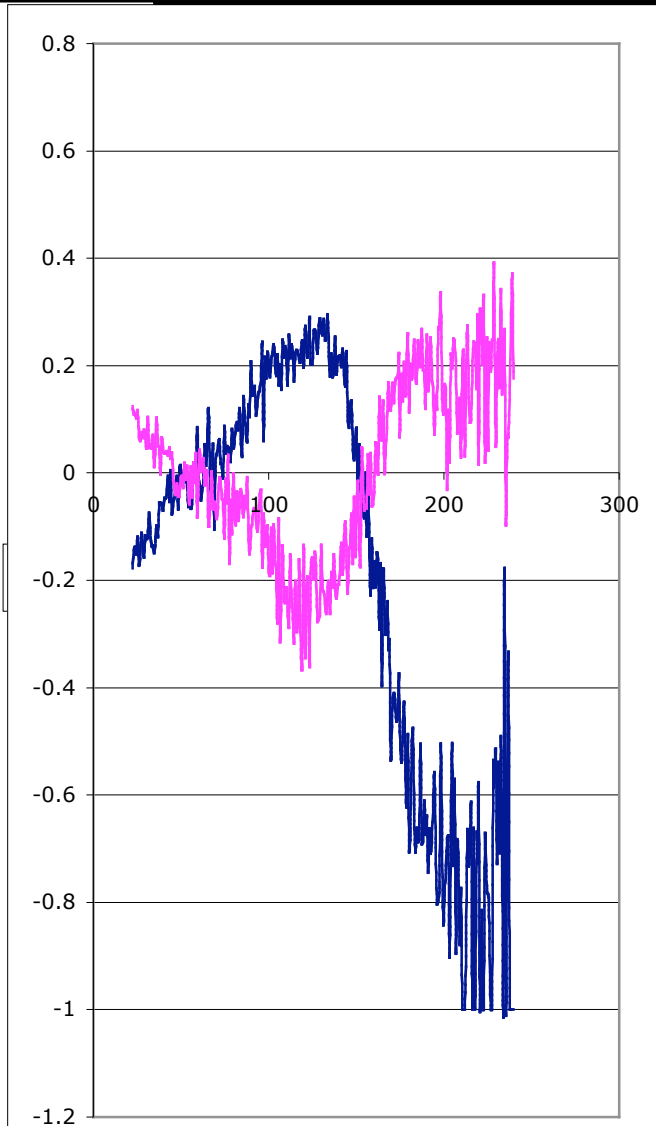
# Data suggests periodicity



day 33



day 86



day 134

day0  
day1

Problem:  
Prove periodicity and  
determine the period



# Bayesian analysis

- Sparse irregularly distributed time points
- Model period = period-bin-size  $\times$  number-of-bins
- Additional variable defining a model: phase of first time-point
- Consider models with periods between 30 and 100 days
- 579 models for rat 1 and 683 models for rat 2

# Cost function

- A model assigns period-bin numbers to data time-points.
- For each cell-size-bin and each period-bin, what is the probability of the cell number counts at different time points assigned to this period-bin belonging to the **same** log-normal distribution?
- For each cell-size-bin and each period-bin, two parameters for the lognormal distribution (median and width).

# Data limitations

- Limited number of time points implies requirements on testable periodic models.
- $(0\ 0\ 0\ 1\ 1\ 1)$  is not a model that tests periodicity because there are no non-contiguous time points that are assigned to the same period-bin. However,  $(0\ 1\ 0\ 0\ 1\ 1)$  and  $(0\ 1\ 0\ 1\ 0\ 1)$  are bin assignments that test periodicity.

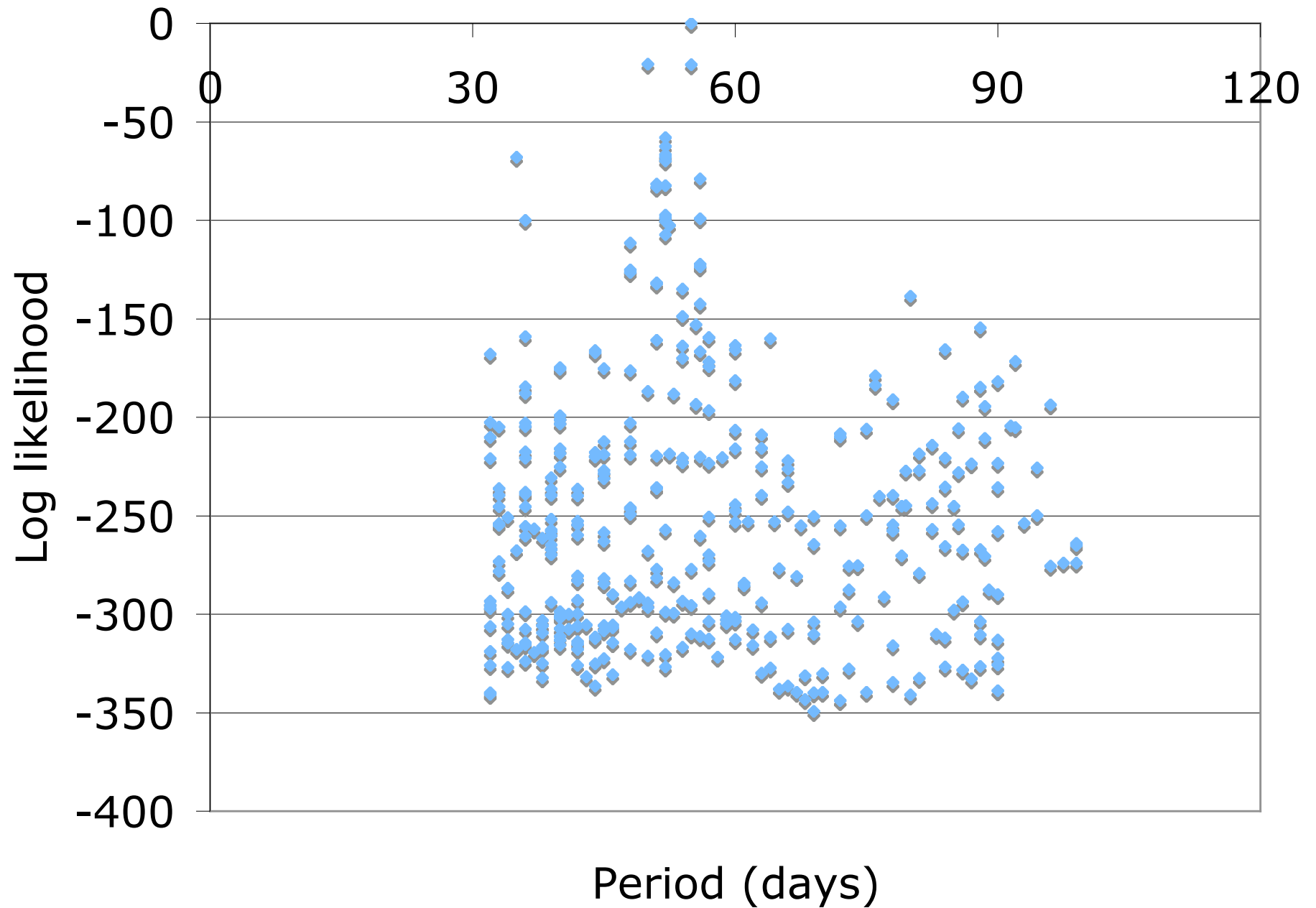
# Parallel-tempering MC

- Parallel tempering Monte Carlo to compute the model likelihood for each model
- Ten equally spaced inverse temperatures
- For each cell-size-bin and each temperature, run a Monte Carlo ( $10^4$  steps for equilibration,  $10^3$  steps for integration)
- The mean likelihood for all the temperatures is the model likelihood, balancing model complexity against goodness of fit.

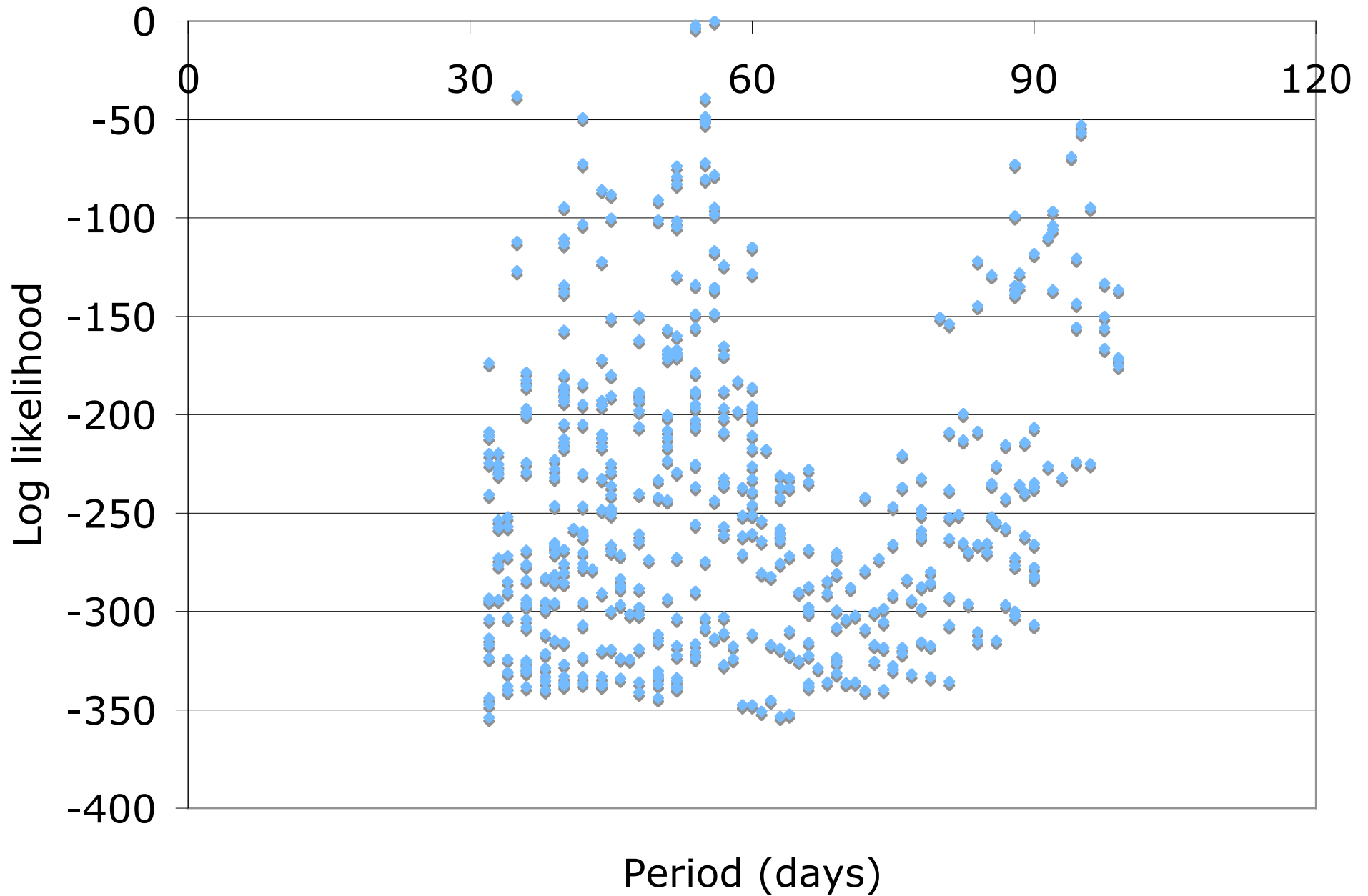
# Results

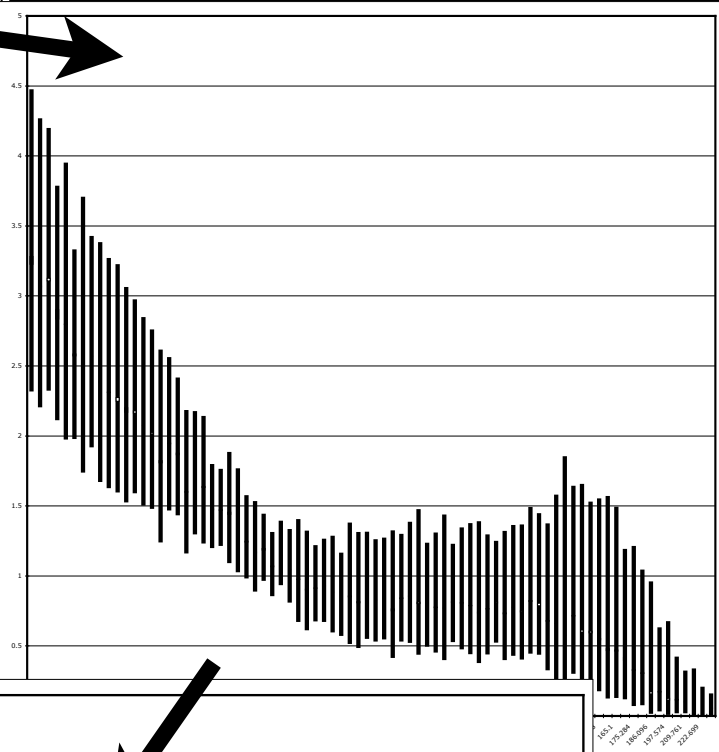
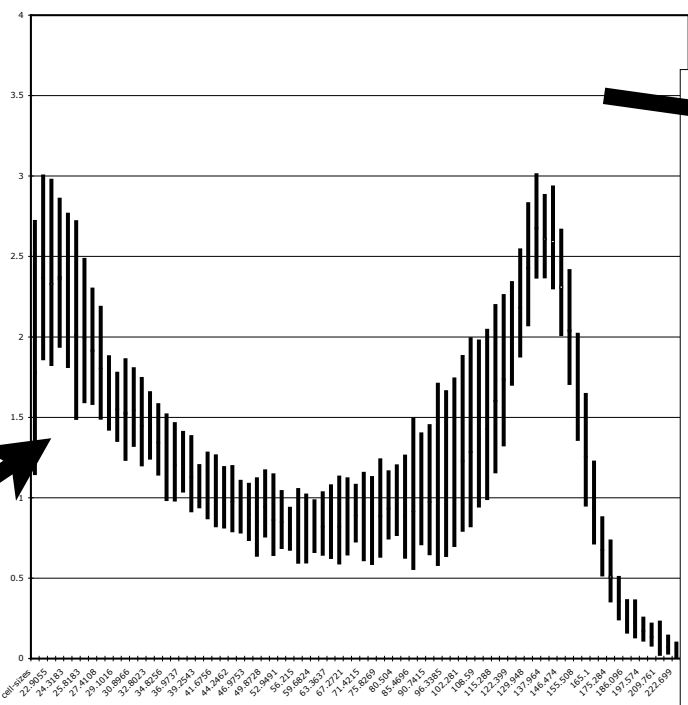
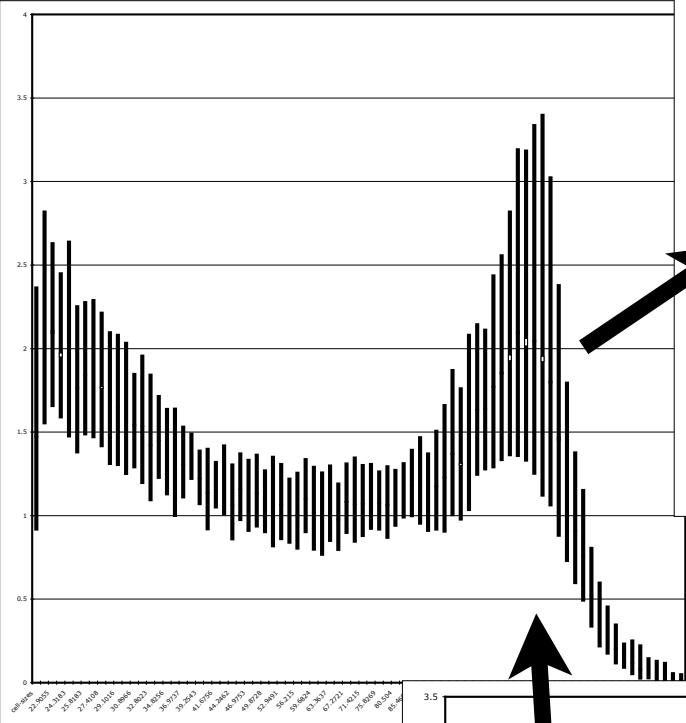
- Periodic with period approximately  $53 \pm 3$  days, independently for both rats
- Different bin numbers selected for different animals for most likely model: Rat 1  $\rightarrow$  5 bins of 10 days each, Rat 2  $\rightarrow$  7 bins of 8 days each

# Log Likelihood of Period (Rat 1)

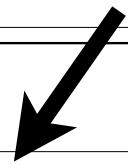
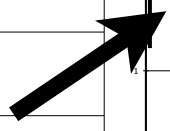
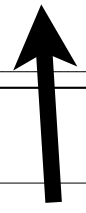
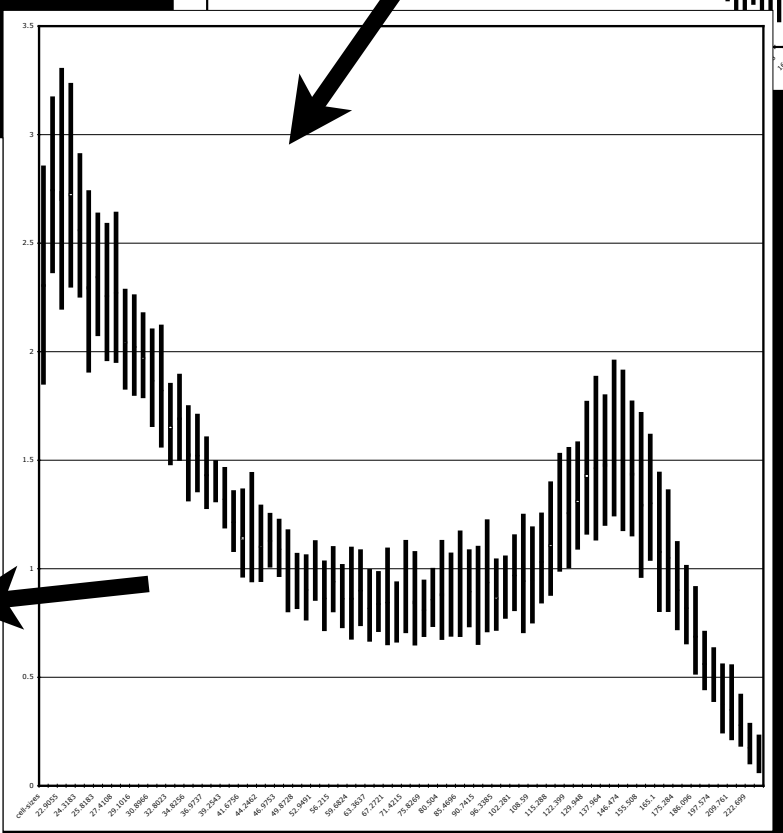
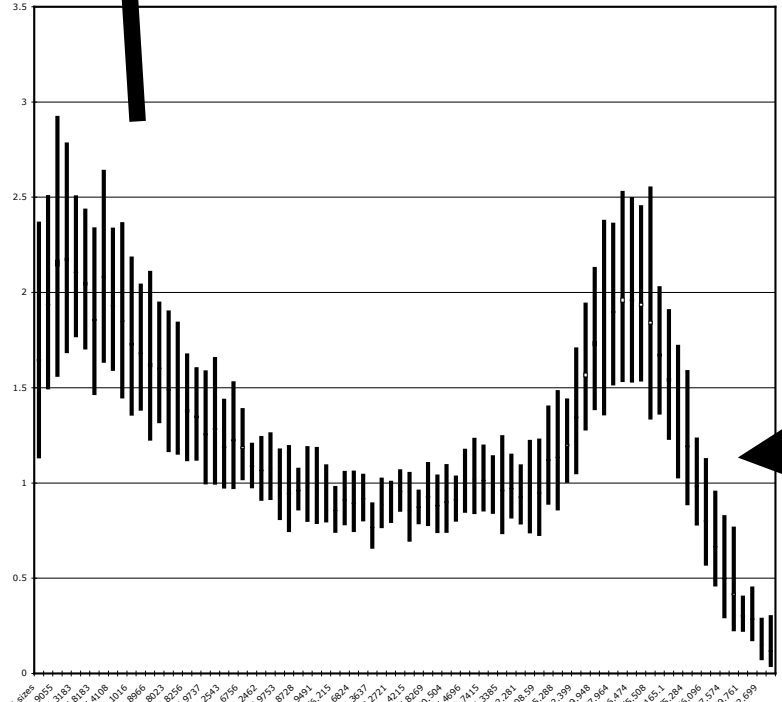


Log Likelihood of Period (Rat 2)



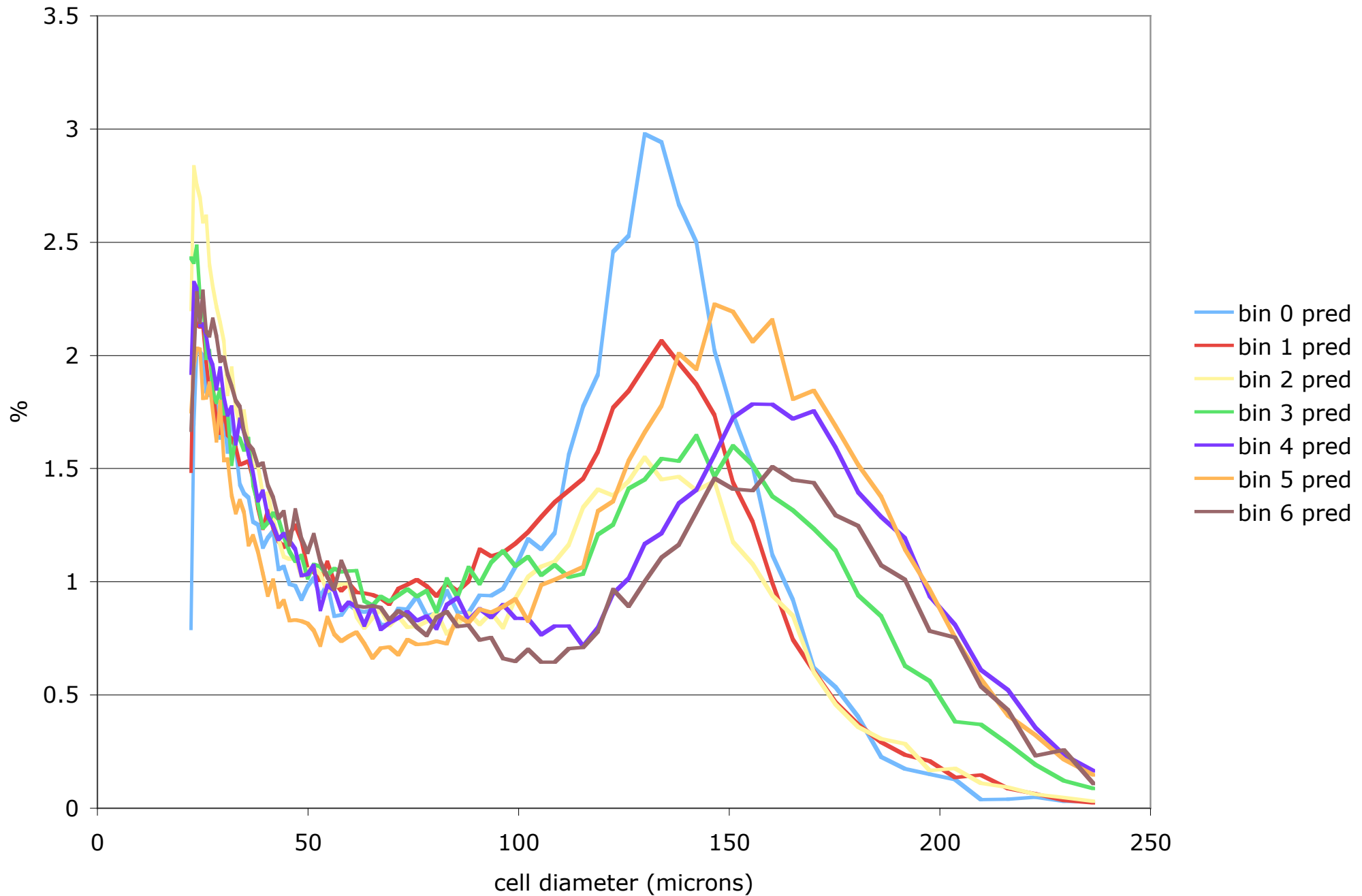


Rat I cycle





# Rat 2 cell-size distributions



# Towards dynamics

- Model as a compression of data
- Model as a limited predictor
- Compressing the model - connecting hypertrophy and hyperplasia

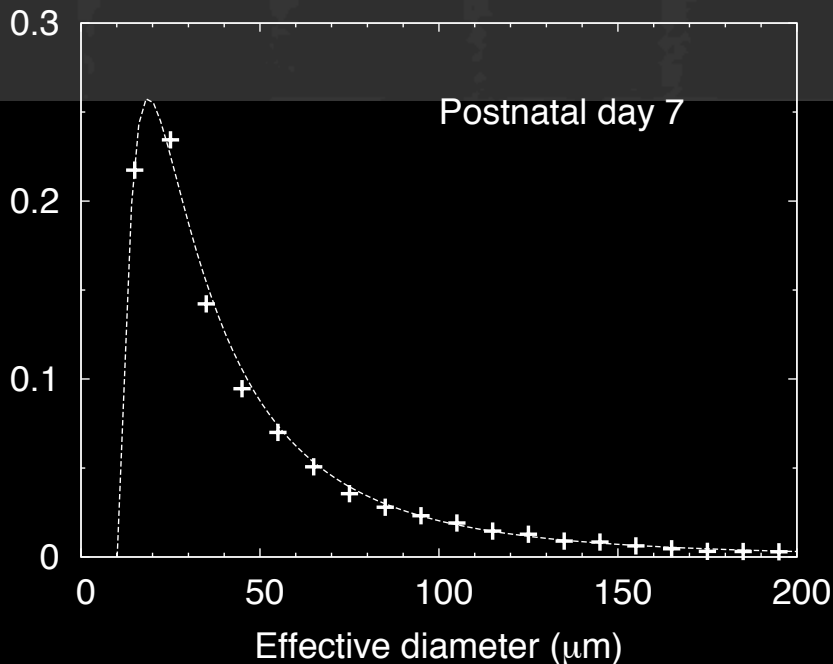
# What could lead to periodicity?

- Maximal lipid uptake occurs when there are a large number of cells in the convective size range.
- If lipid is available but there is limited uptake capacity available, then new recruitment is needed.

# Size distribution of pancreatic islets

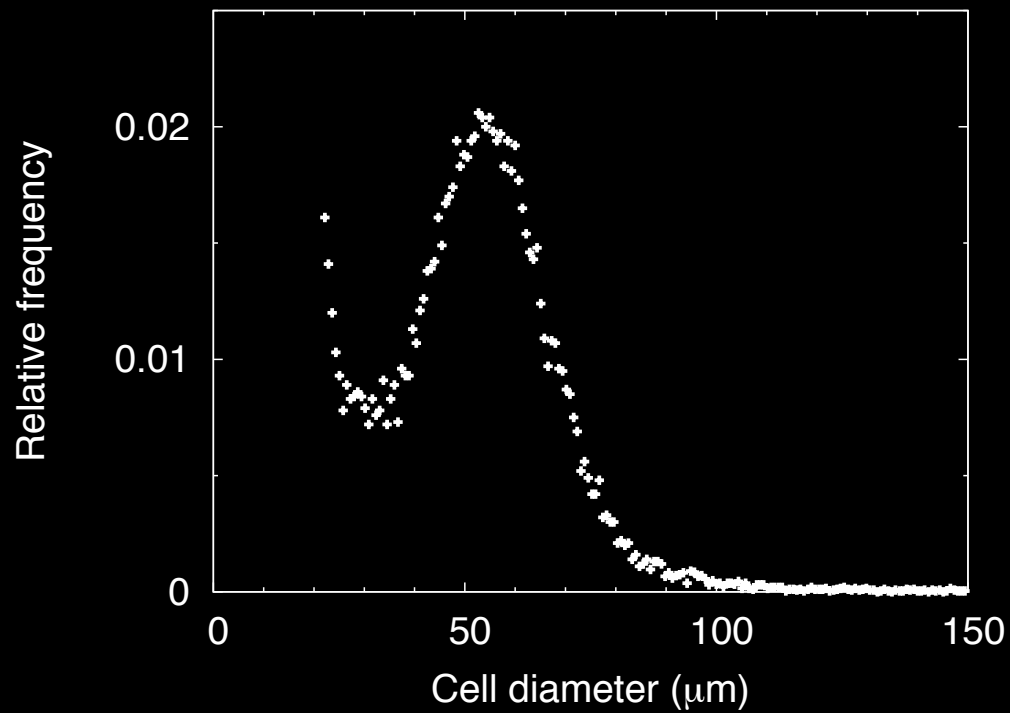
MIP-GFP mouse

5 mm

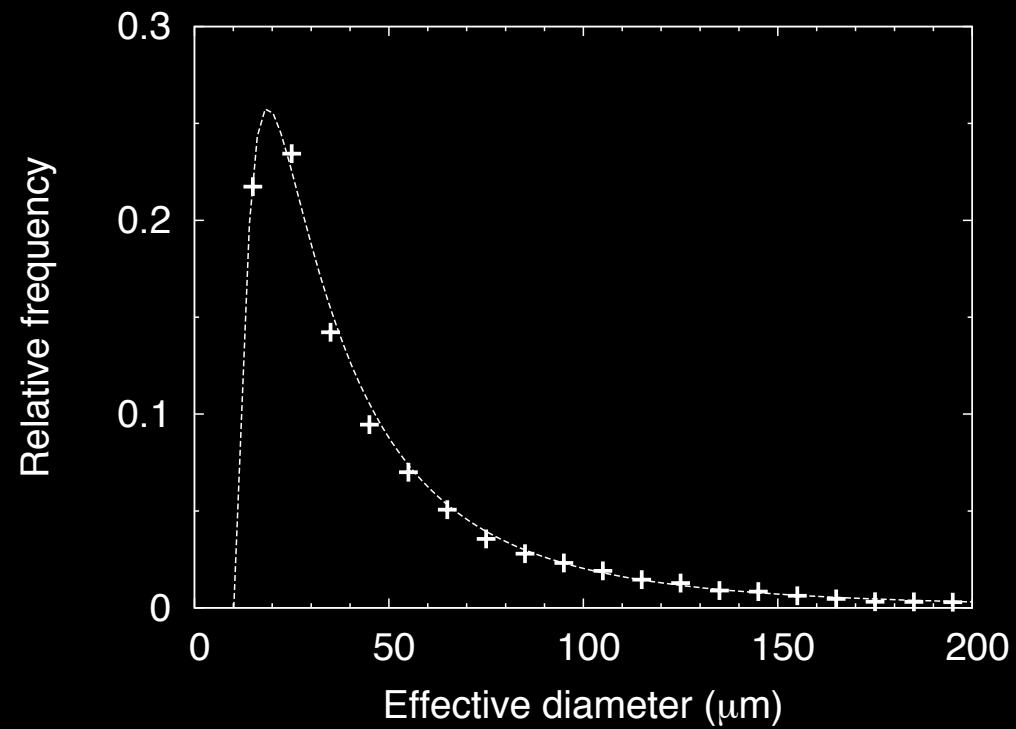


- **basic development** (islet neogenesis, proliferation potential, and islet fission)
- **physiological and pathological conditions** (aging, pregnancy, diabetes, obesity, and insulinoma)

## size distribution of adipose cells



## size distribution of pancreatic islets



# Acknowledgment

Fellow: Junghyo Jo

Collaborators:

- Samuel W. Cushman (Diabetes Branch)  
Shawn Mullen  
Teresa Liu
- Kevin D. Hall (Lab. of Biological Modeling)  
Juen Guo
- Oksana Gavrilova (Mouse Metabolism Core Lab.)  
Stephanie Pack  
William Jou
- Anne E. Sumner (Clinical Endocrinology Branch)
  
- Manami Hara (The University of Chicago)  
German Kilimnik  
Abraham Kim