

# The Data Science Revolution in Biomechanics:

Traditional statistical tests vs modern machine learning methods in the study of lizard locomotion

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**BALL STATE  
UNIVERSITY**



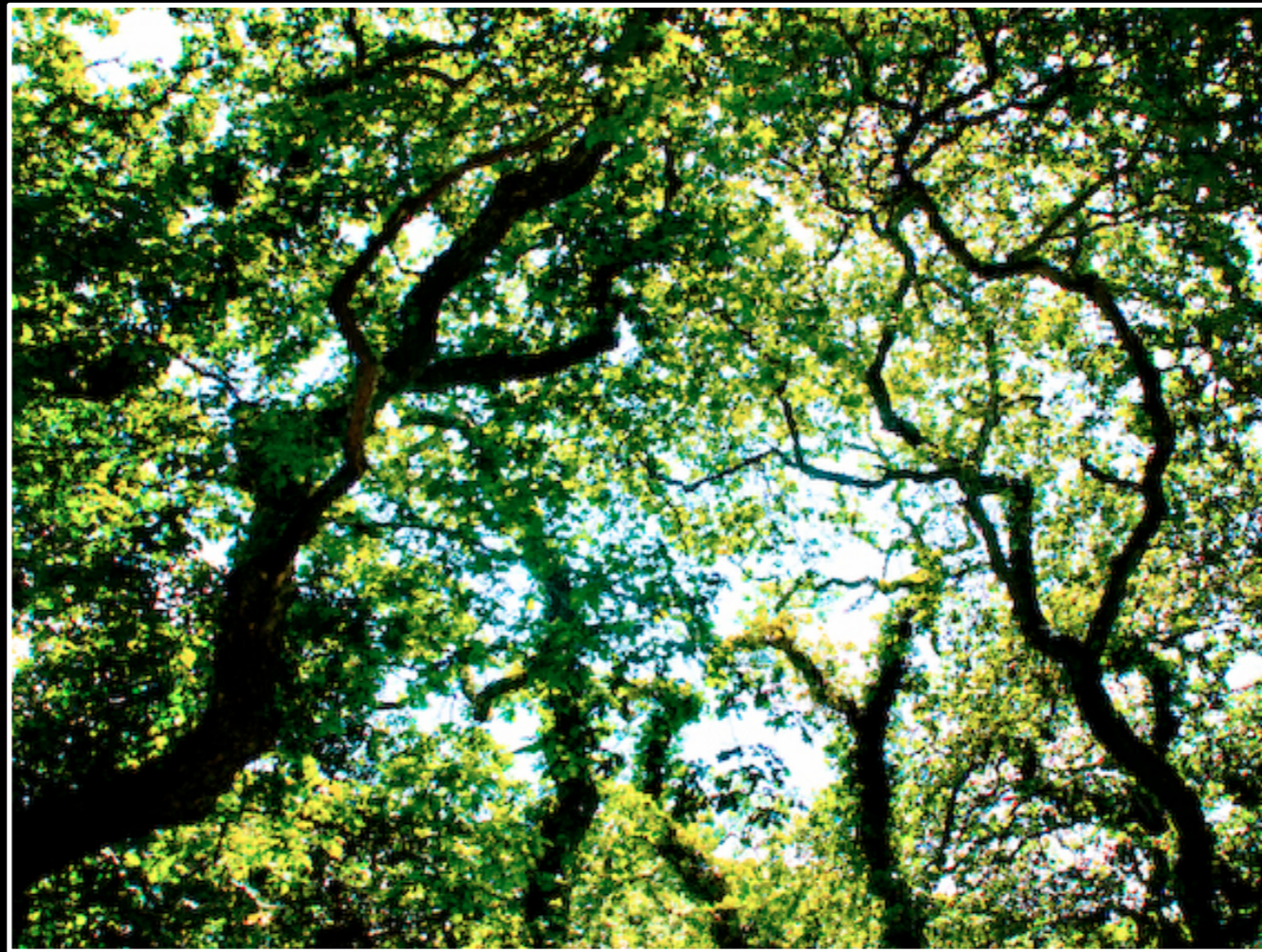
# The environment is complex



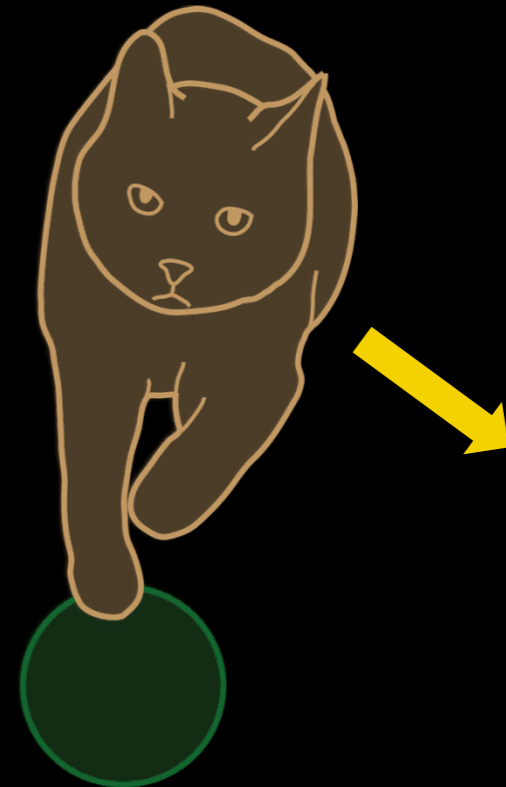
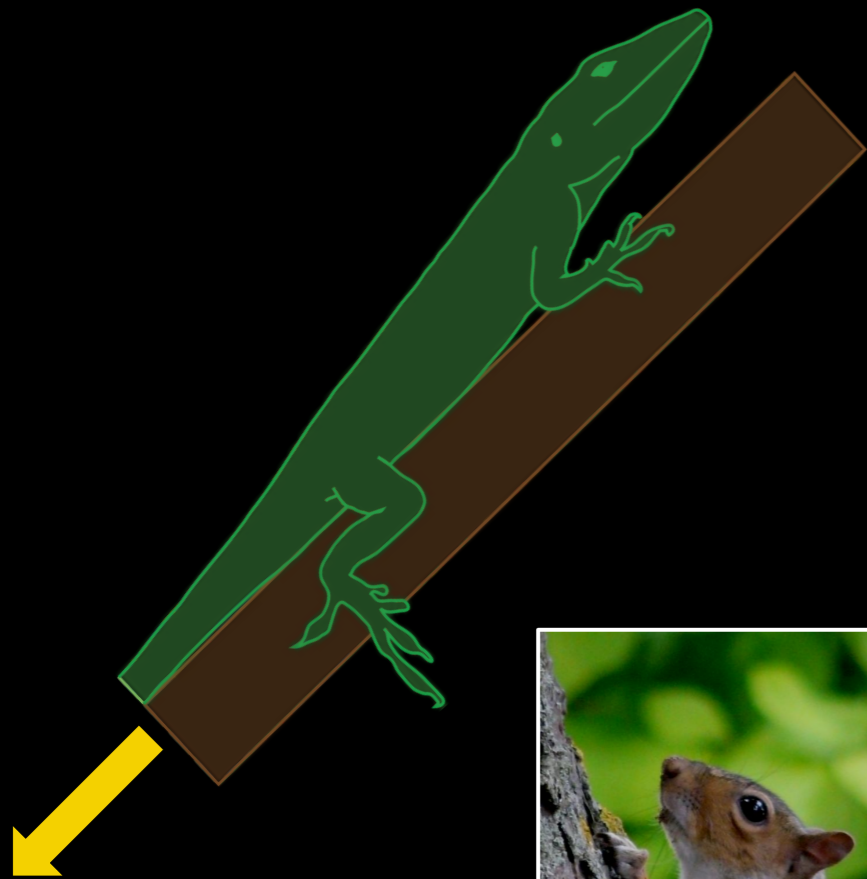
# Animals do all sorts of wild and wacky things



# The challenges posed by arboreality



# The challenges posed by arboreality



# Morphological adaptation meets functional demand



primates



rodents



lizards

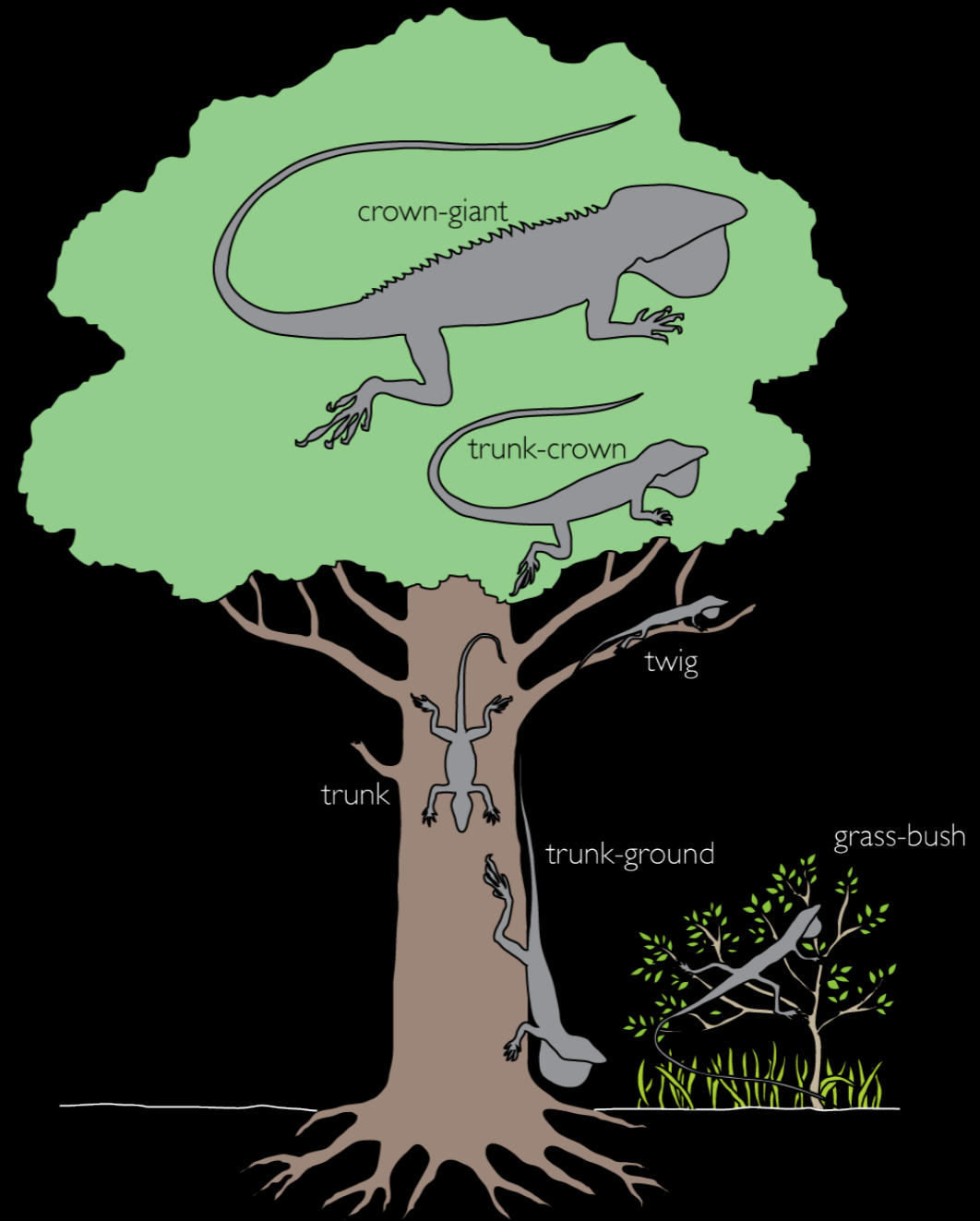


marsupials



kinkajous

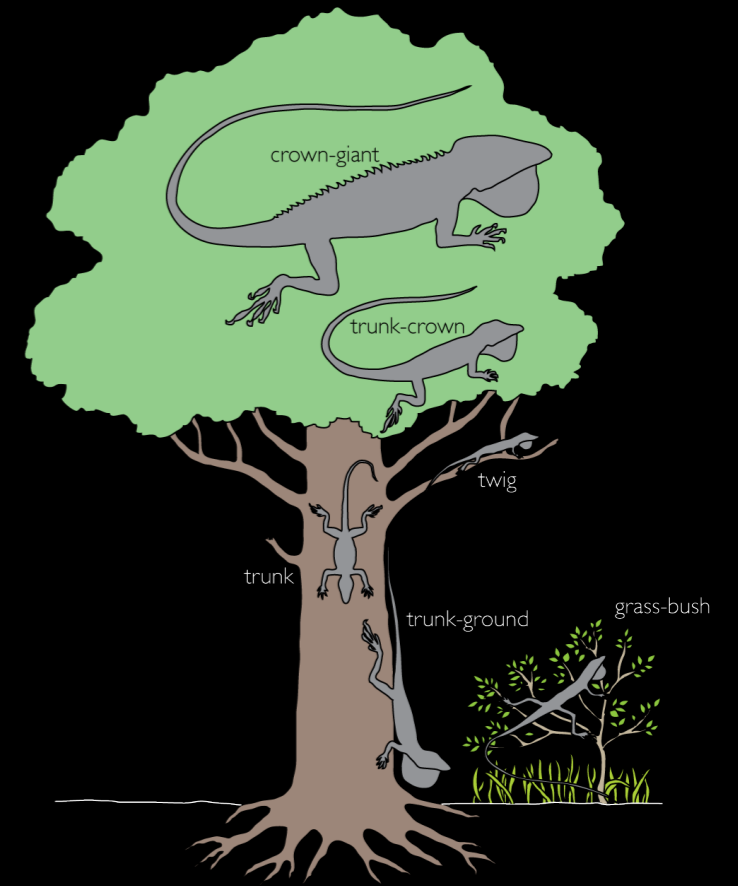
# Morphological adaptation meets functional demand



## *Anolis* ecomorphs

Losos (2009) *Lizards in an evolutionary tree*

# Morphological adaptation meets functional demand

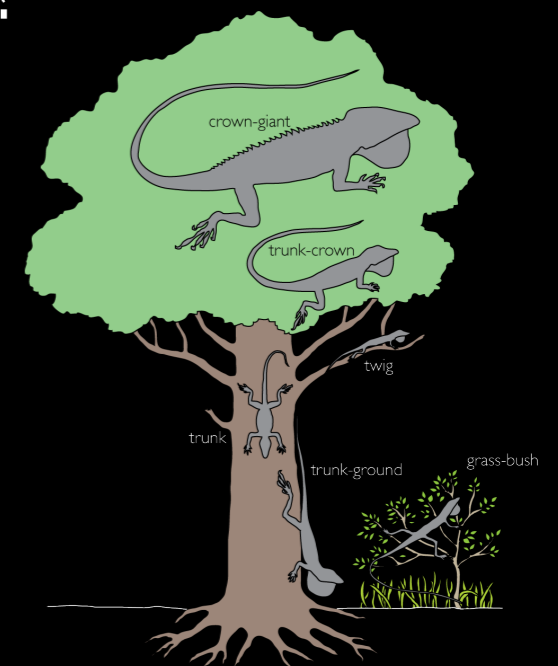


How does the environment impact locomotor behavior of *Anolis* lizards?

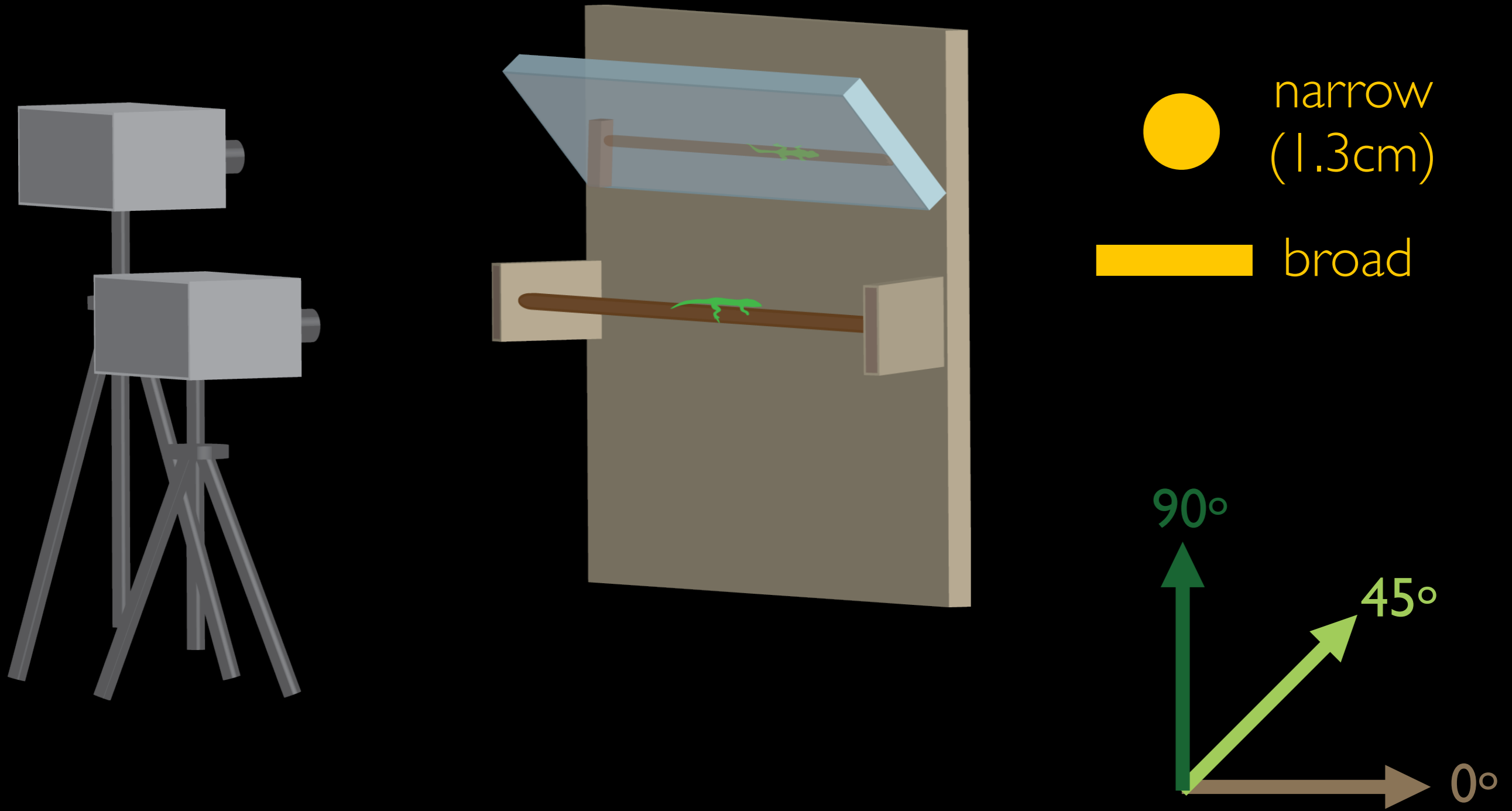


# How does the environment impact locomotor behavior of *Anolis* lizards?

- ♦ How do the **movements** of the forelimbs and hind limbs change on different arboreal surfaces?
- ♦ How does the **coordination** of forelimb and hind limb joints change on different arboreal surfaces?

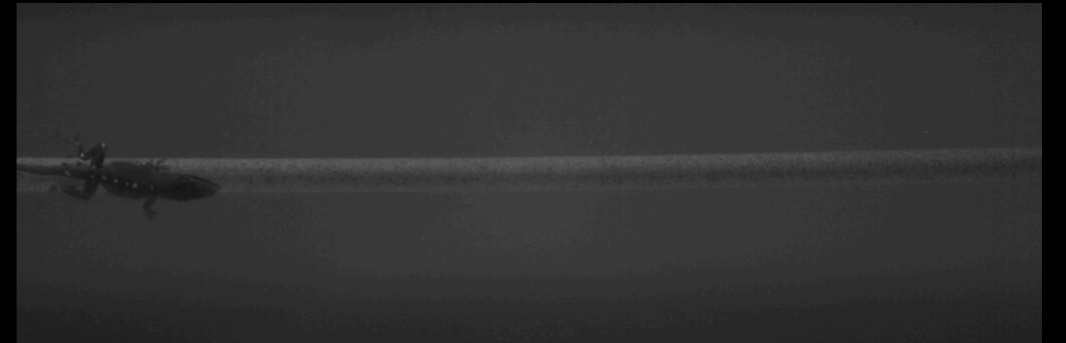


# Experimental setup



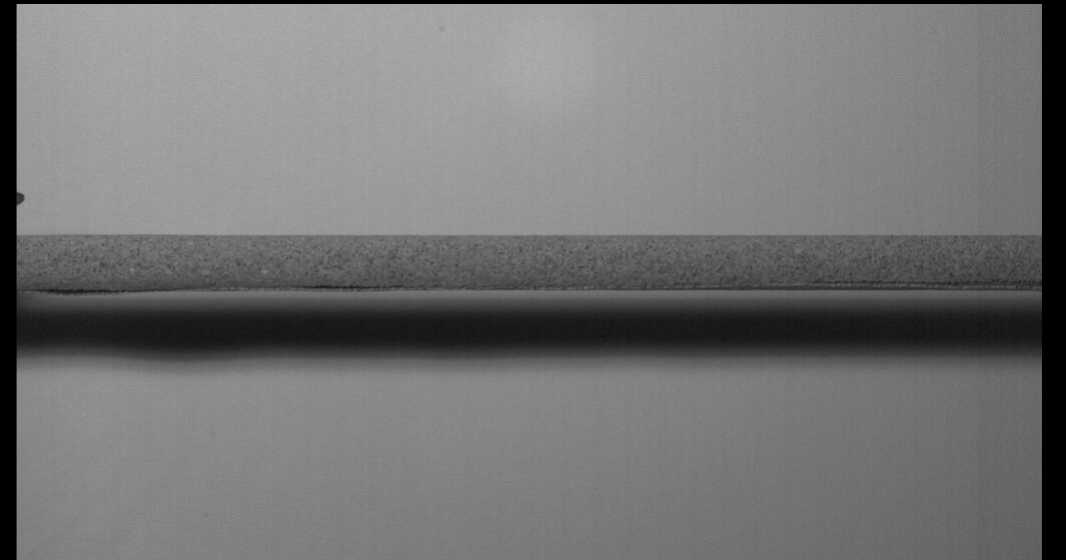
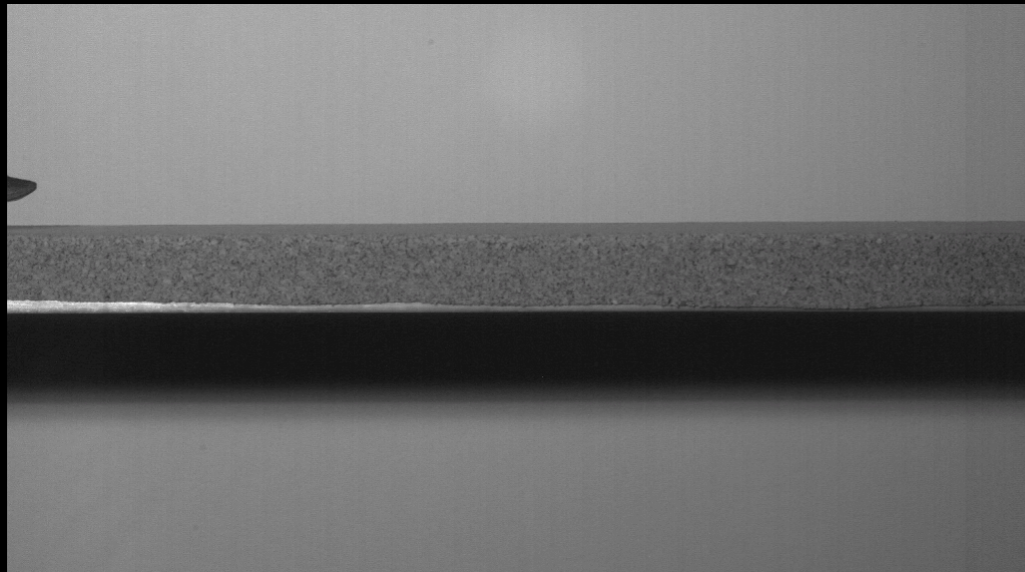
Broad, flat perch

Narrow perch

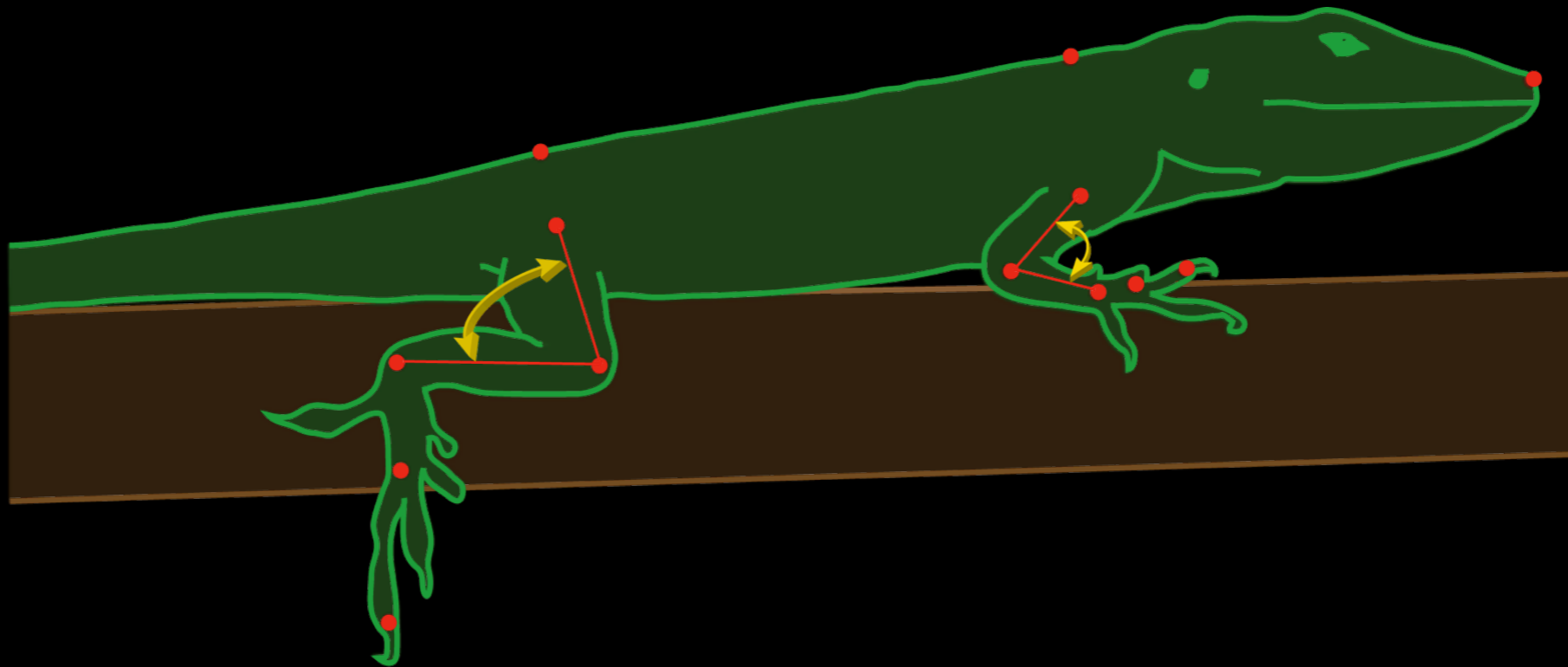


Top view

Side view

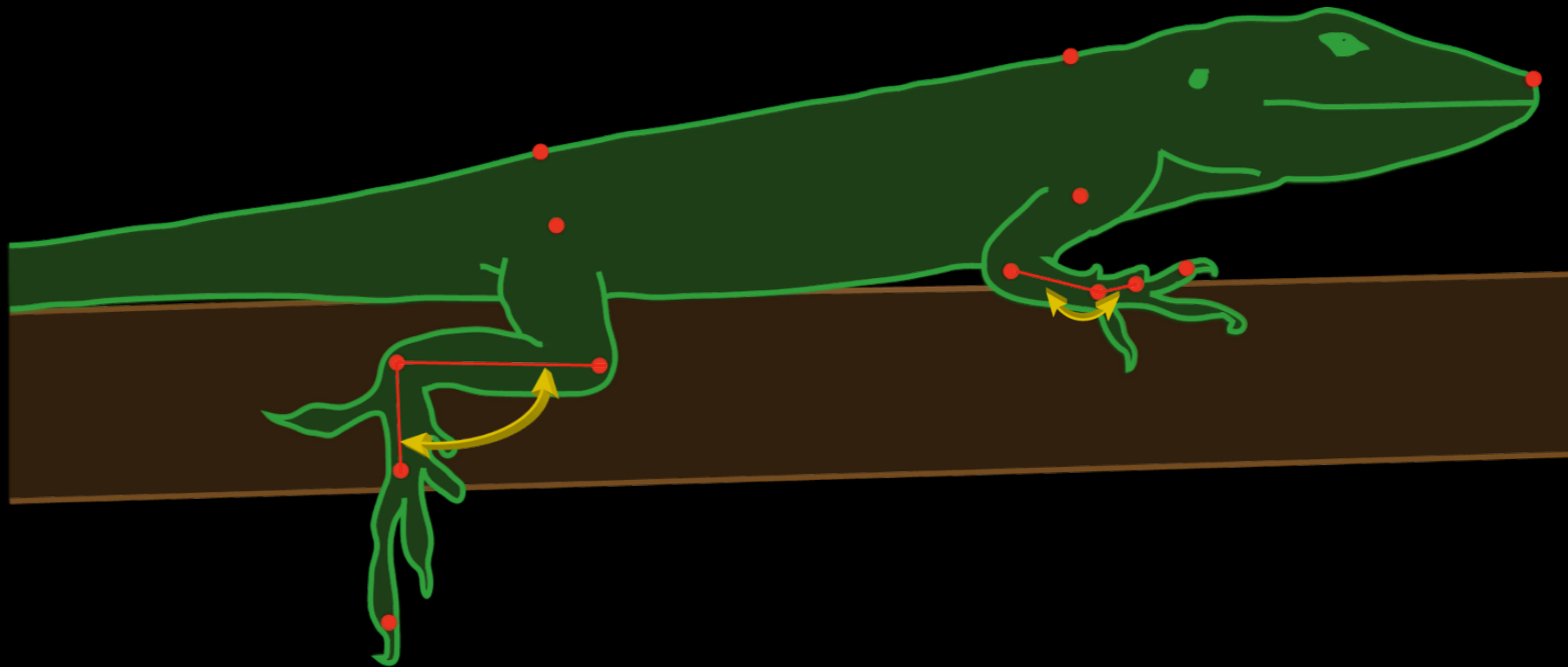


# 3D digitized points and angles



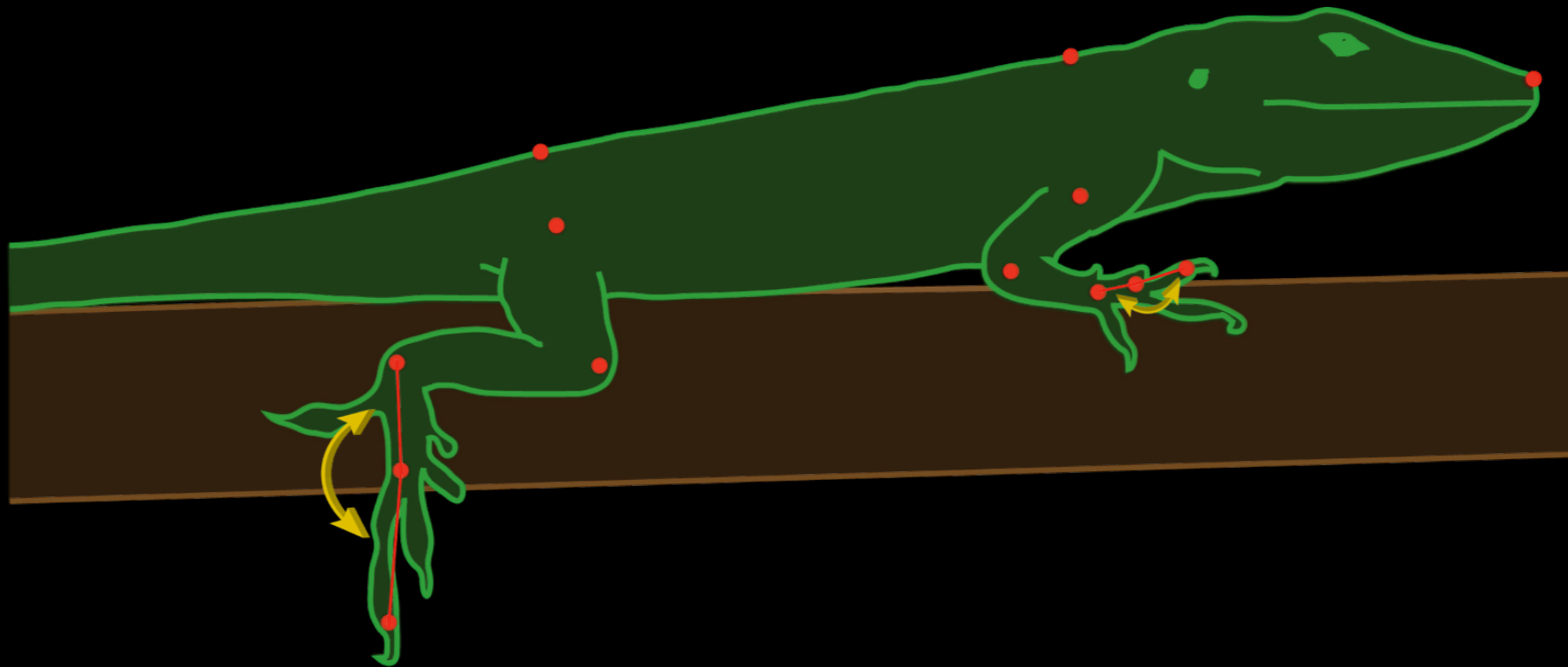
Elbow and knee angle

# 3D digitized points and angles



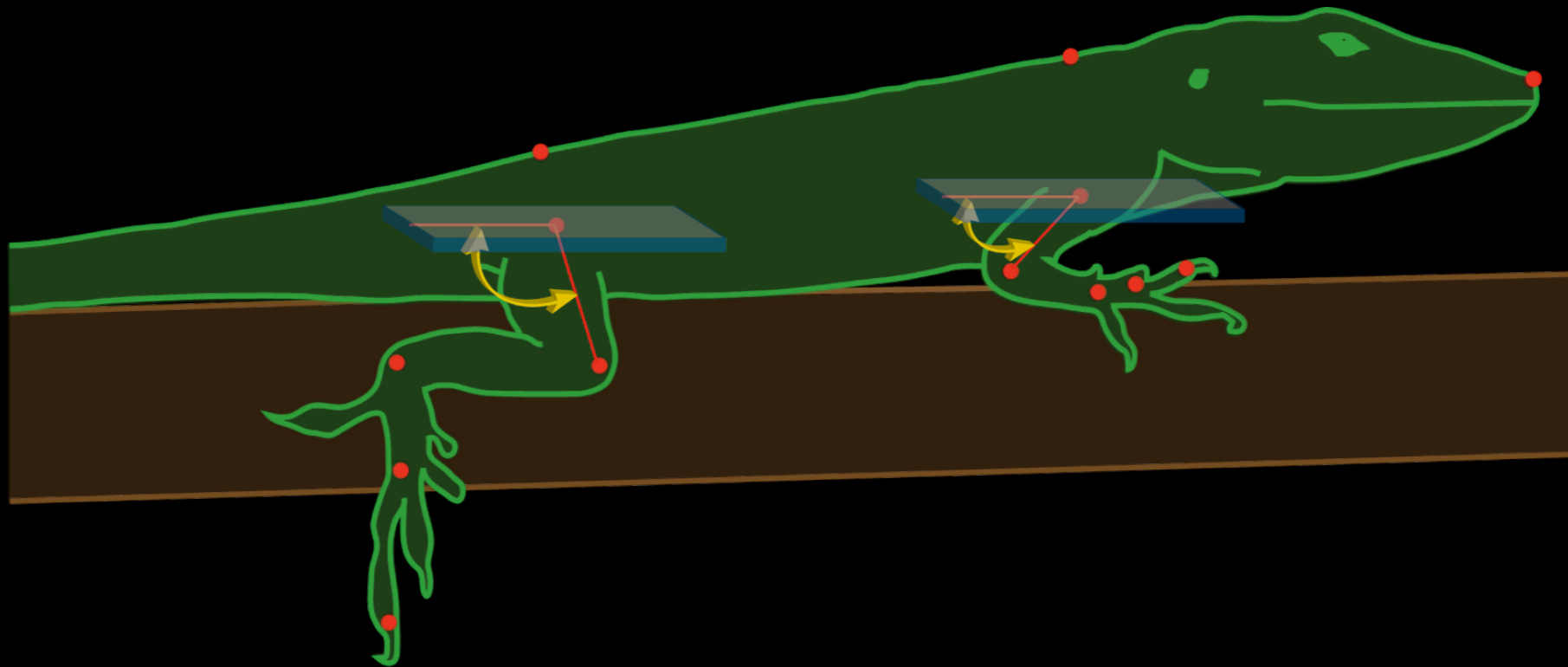
Wrist and ankle angle

# 3D digitized points and angles



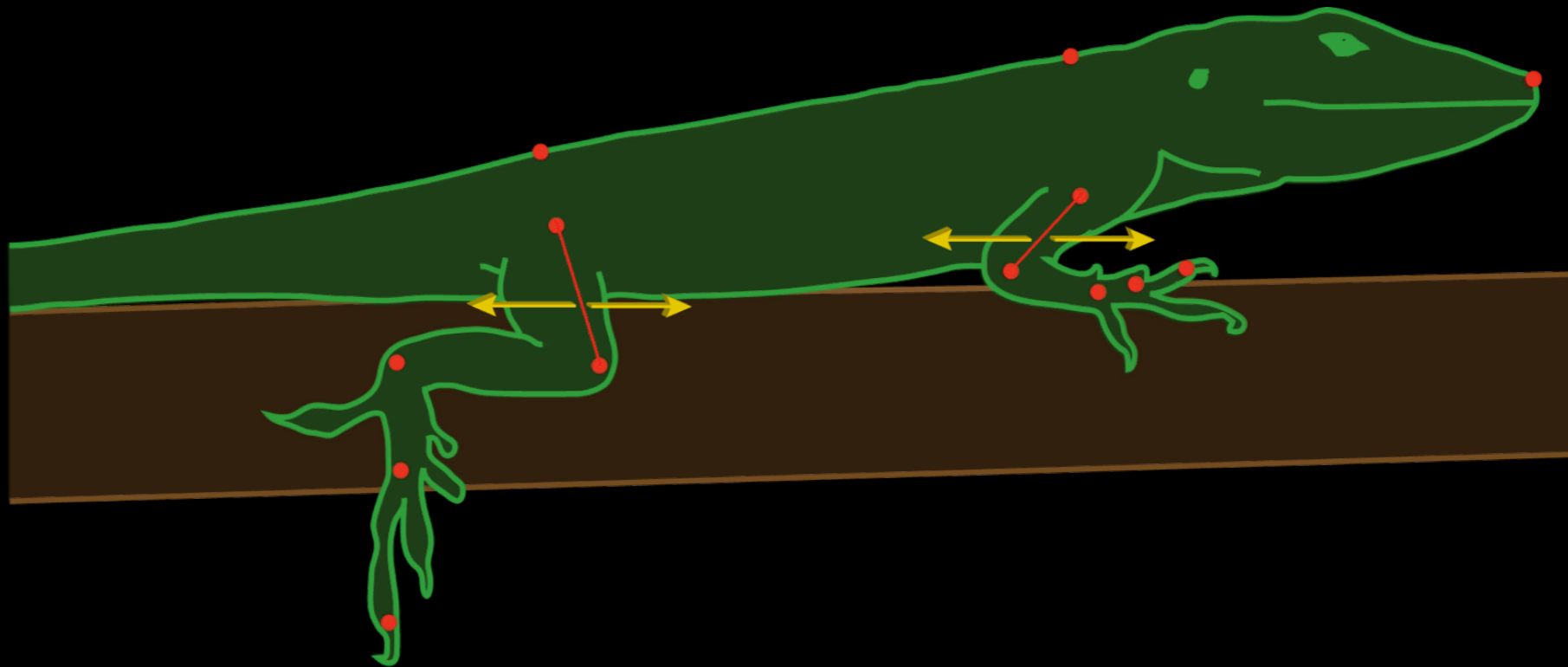
Fore- and hind toe angle

# 3D digitized points and angles



Humerus and femur depression

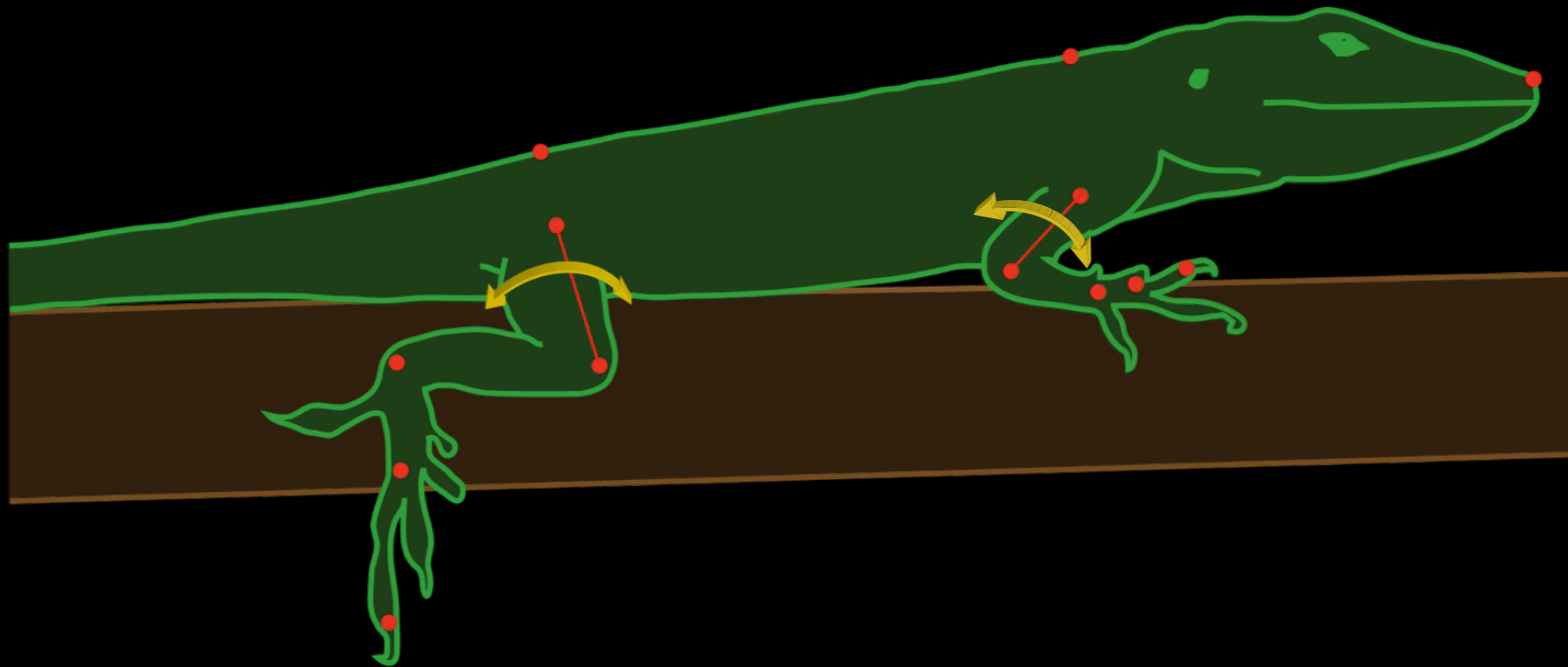
# 3D digitized points and angles



Humerus and femur protraction and retraction



# 3D digitized points and angles

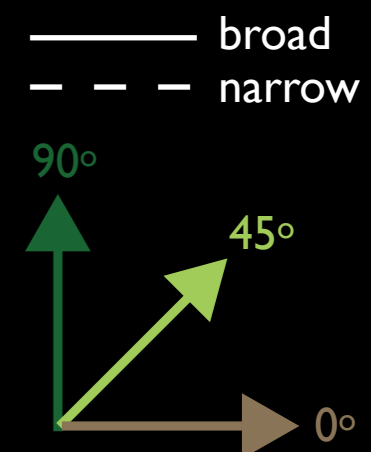
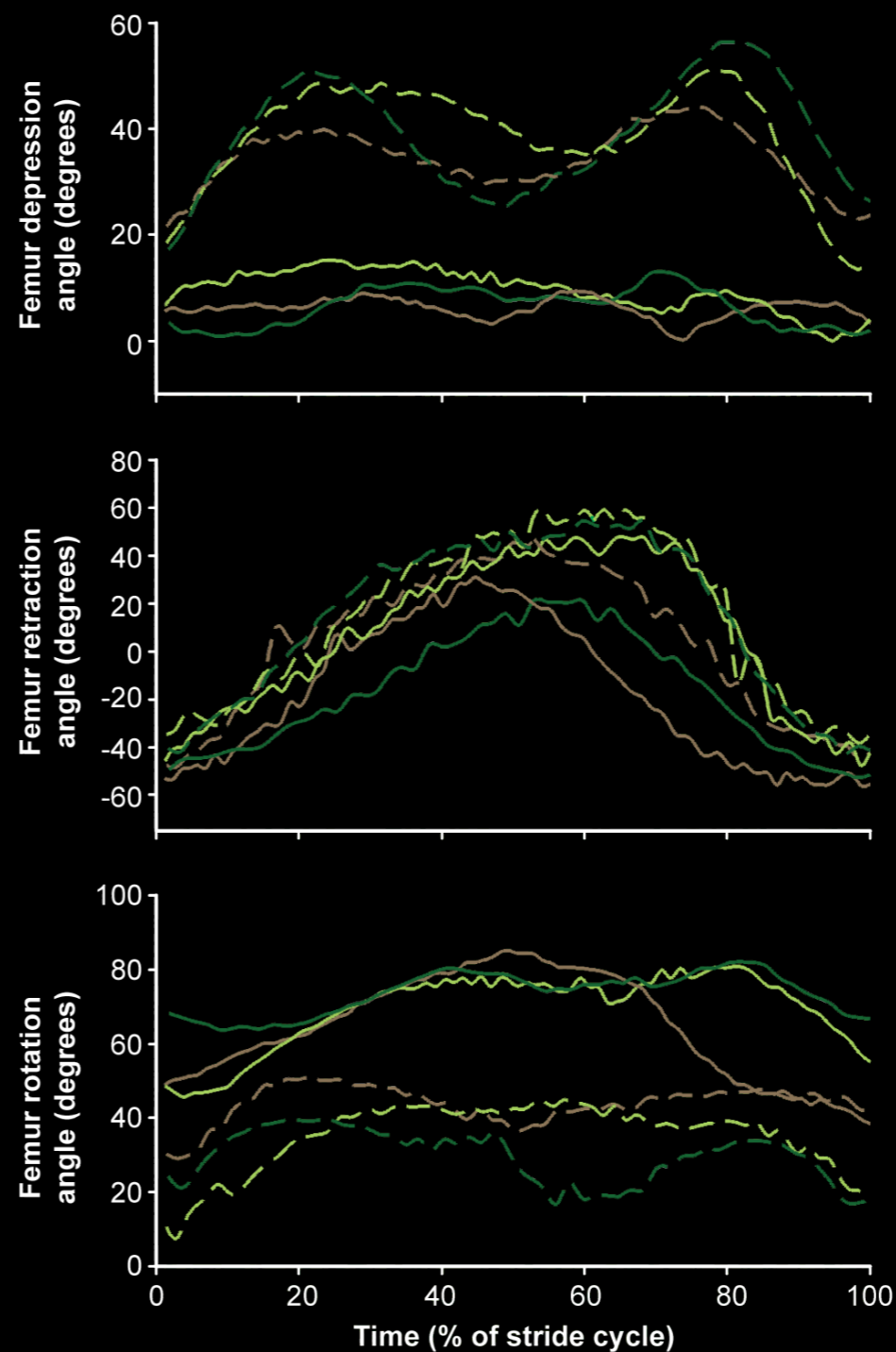
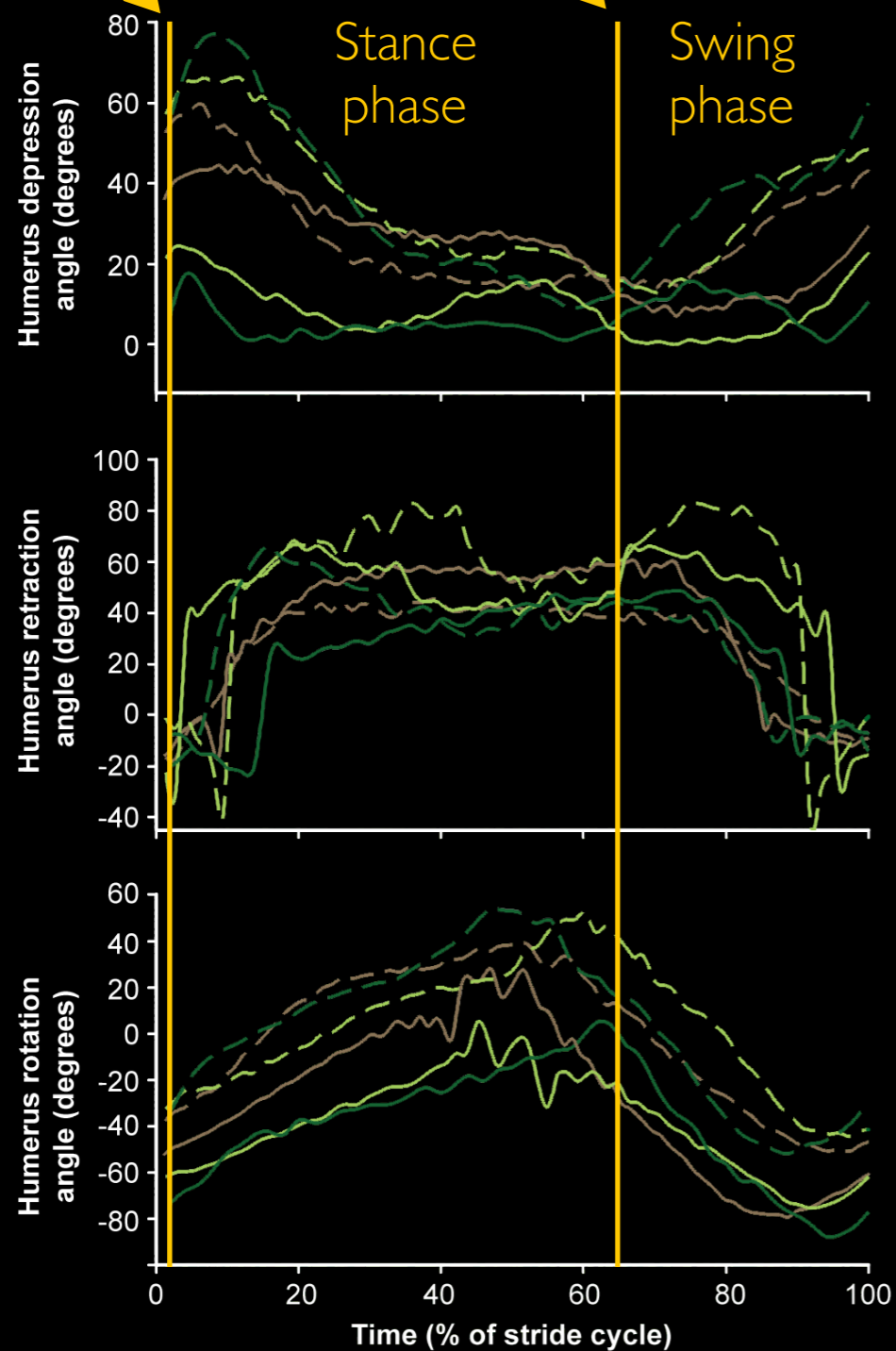


Humerus and femur long-axis rotation

# Joint angle - time series & single time points

Begin stance

End stance



# How should we analyze this dataset?

	Univariate	Multivariate	Time series
Visualization	✓	✓	✓
Hypothesis test	✓	✓	
Classification		✓	

# How should we analyze this dataset?

## ◆ Univariate methods

- ◆ ANOVA to test for differences between substrates **in each variable (time point)** - visualized using box-type plots

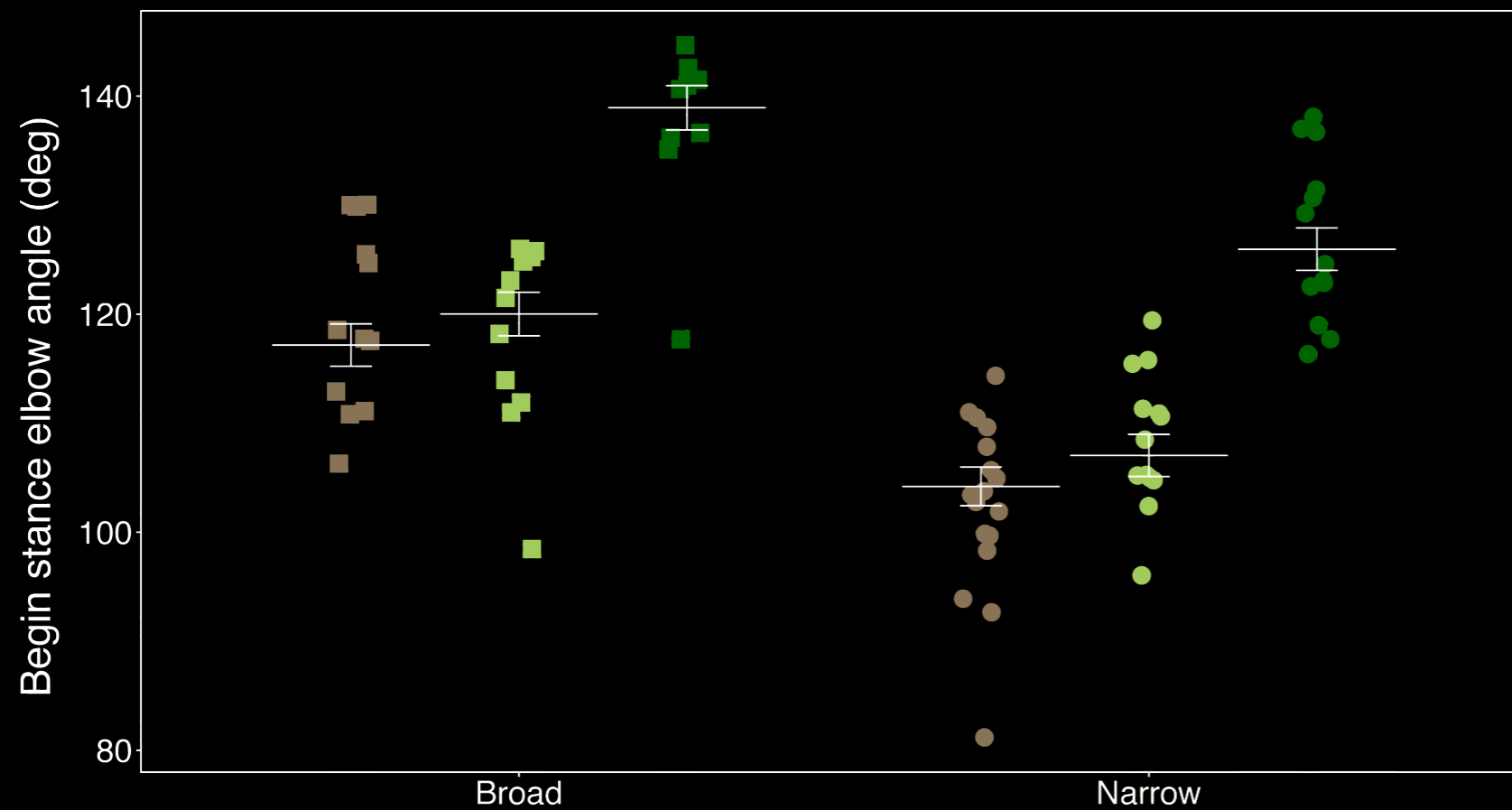
## ◆ Multivariate methods

- ◆ MANOVA to test for differences between substrates **when considering all variables together**
- ◆ Canonical Discriminant Analysis (CDA) for visualization of maximal group separation and to determine how **all variables together** contribute to group separation
- ◆ Linear Discriminant Analysis (LDA) to classify observations (strides) according to group (environment) using the **combined variables**

## ◆ Time series analyses

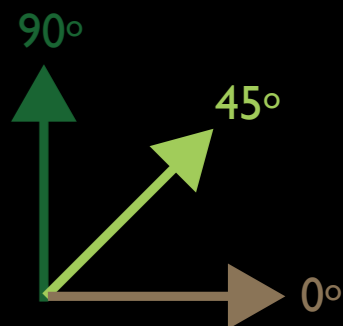
- ◆ Dynamic Time Warping to visualize distances between the **time series of all joint angles**

# Two-way mixed model ANOVA

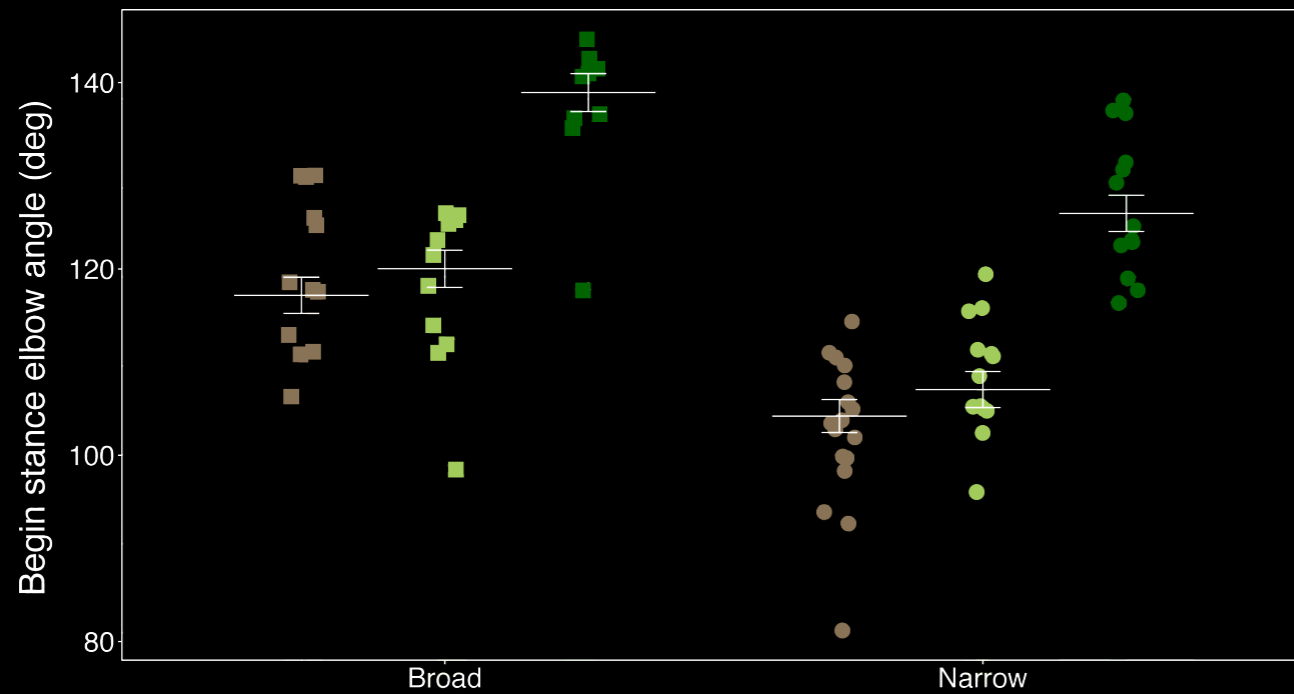


Perch diameter:  $P < 0.0001$   
Incline:  $P < 0.0001$   
Interaction:  $P = \text{N.S.}$

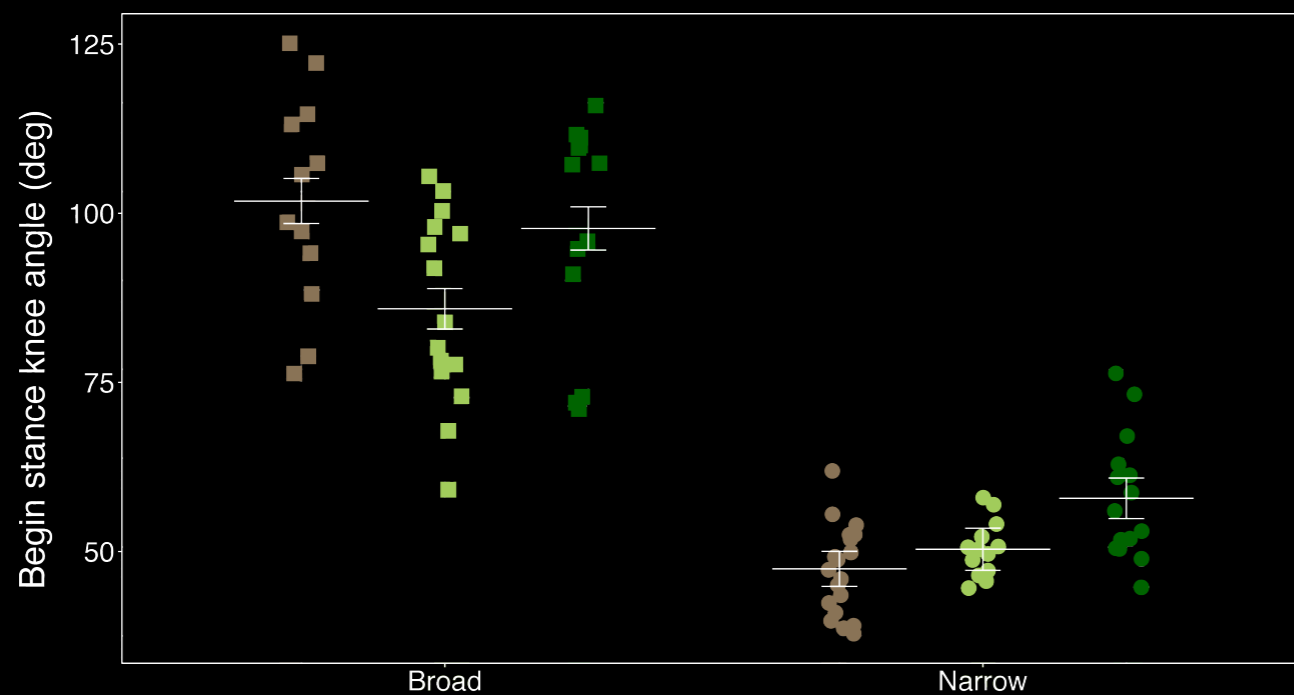
- broad
- narrow



# Two-way mixed model ANOVA

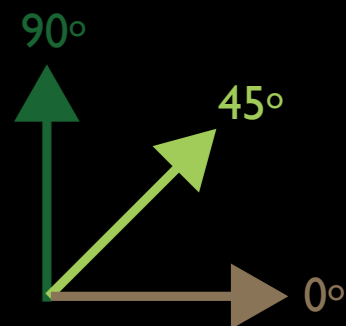


Perch diameter:  $P < 0.0001$   
Incline:  $P < 0.0001$   
Interaction:  $P = \text{N.S.}$



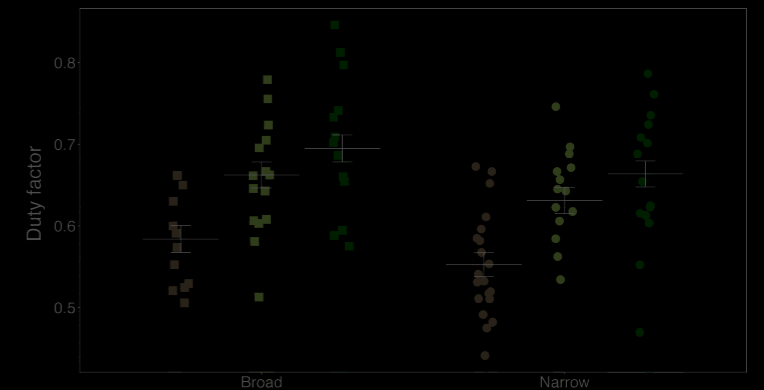
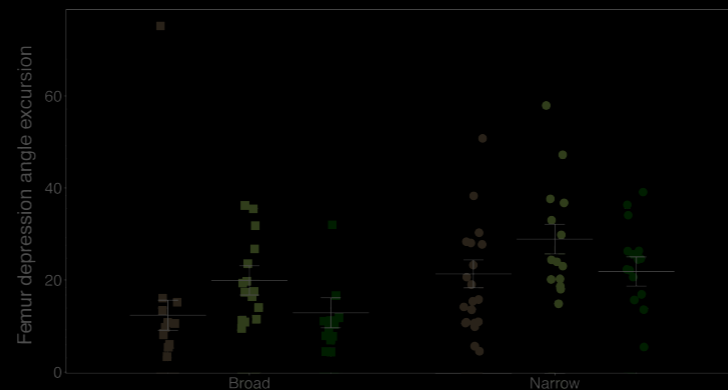
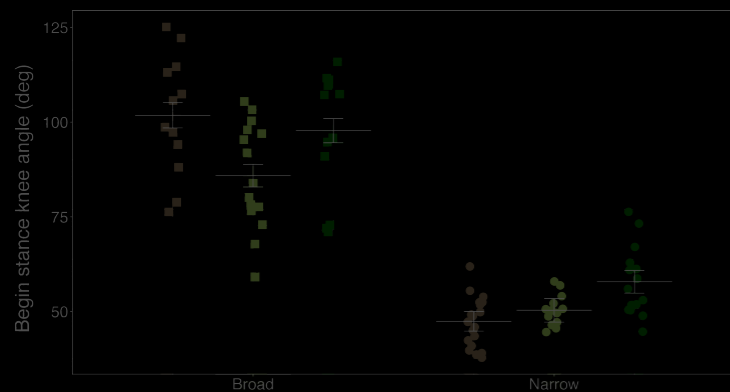
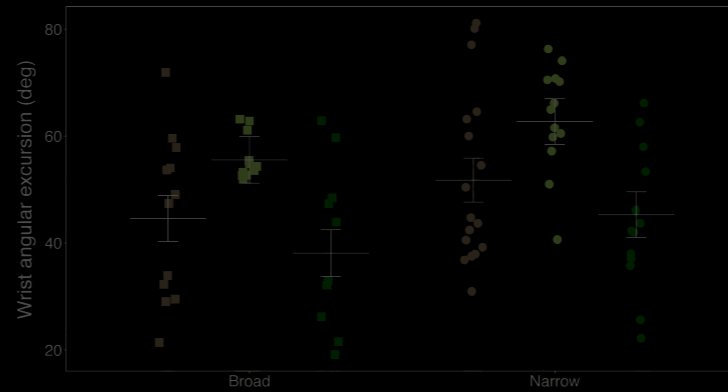
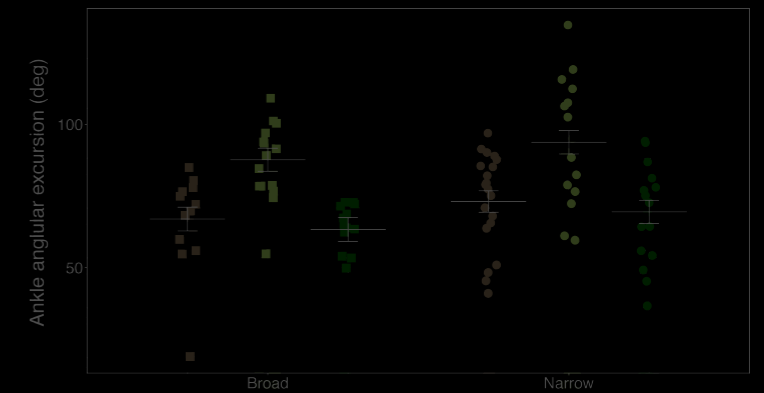
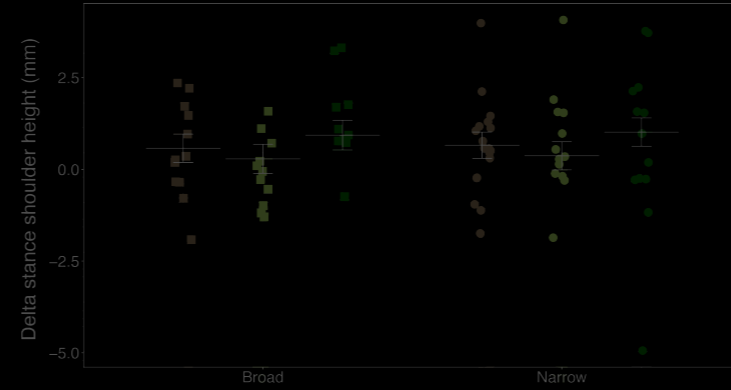
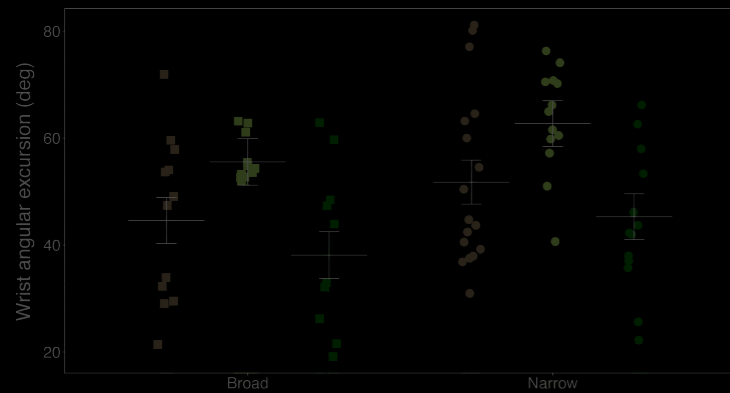
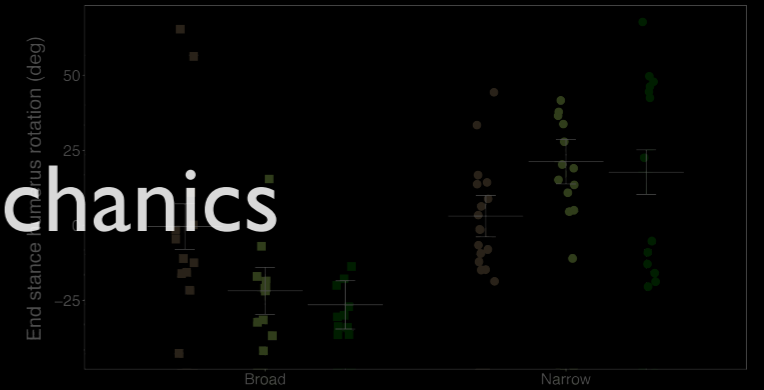
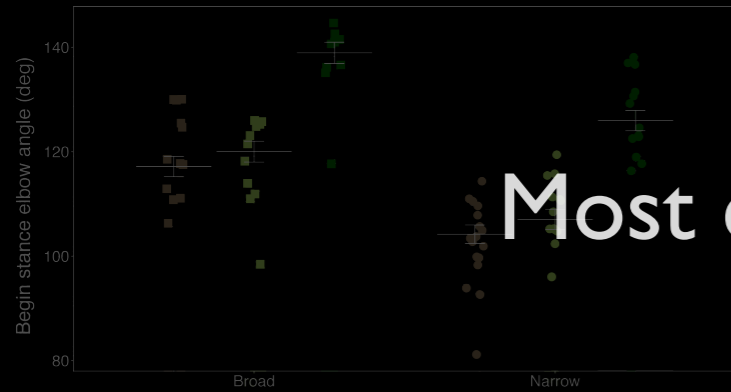
Perch diameter:  $P < 0.0001$   
Incline:  $P = 0.0032$   
Interaction:  $P = 0.0115$

- broad
- narrow



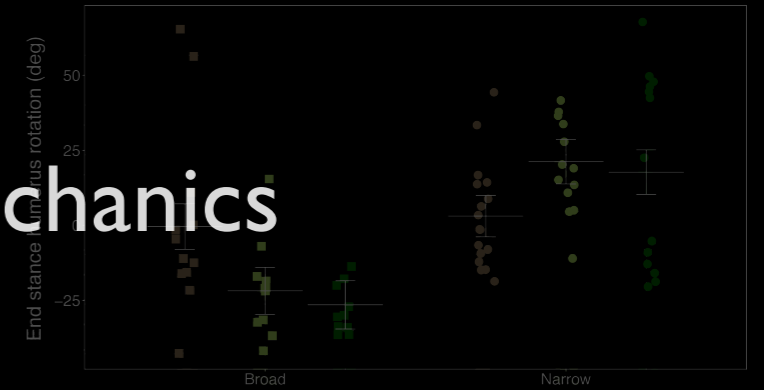
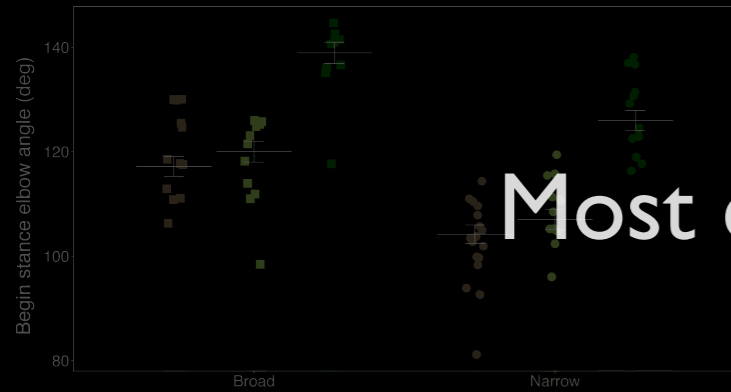
# Two-way mixed model ANOVA

Most common strategy in biomechanics  
AND easy to interpret

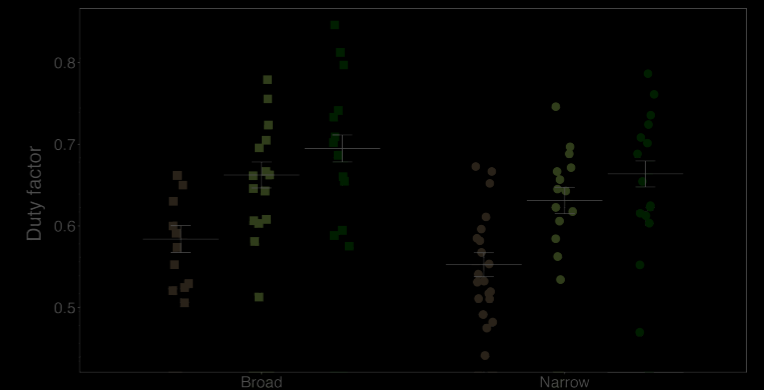
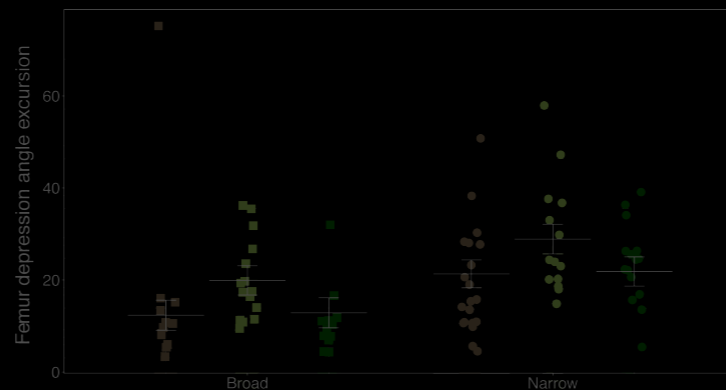
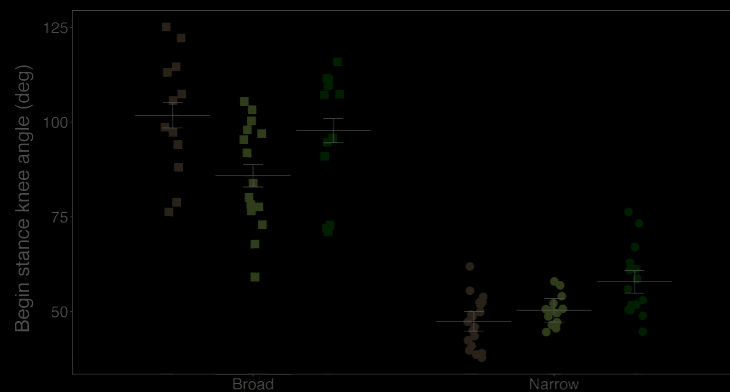
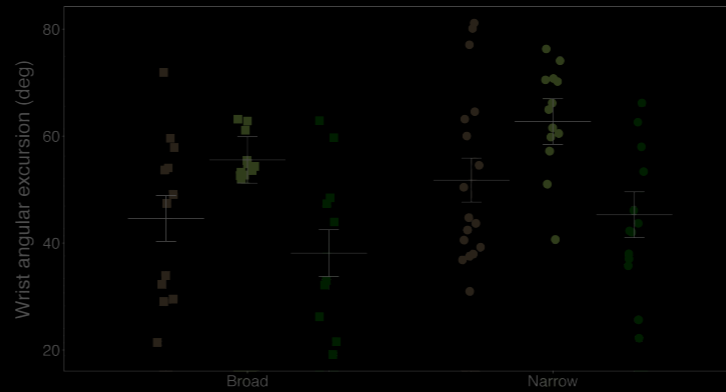
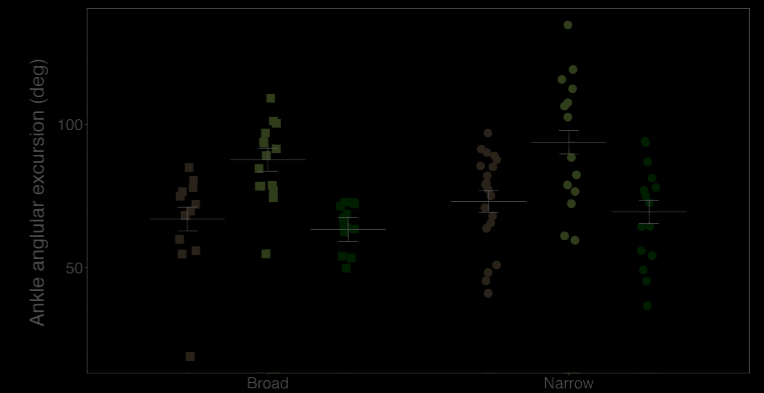
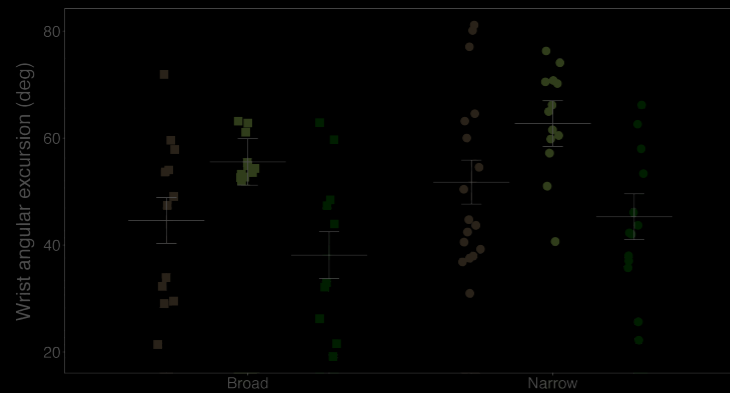
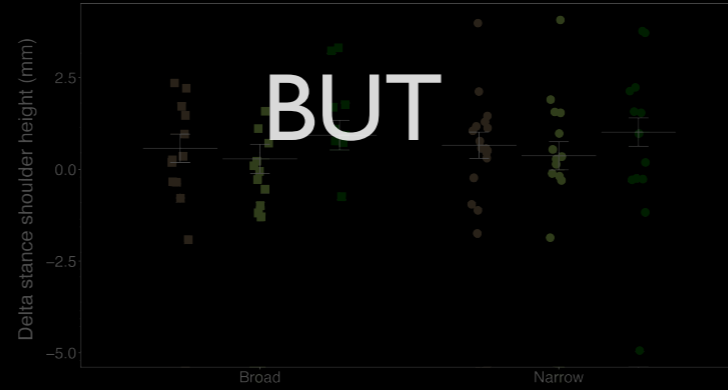


# Two-way mixed model ANOVA

Most common strategy in biomechanics  
AND easy to interpret



BUT



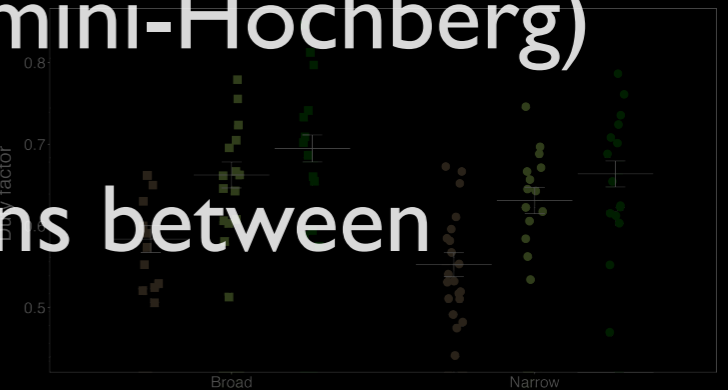
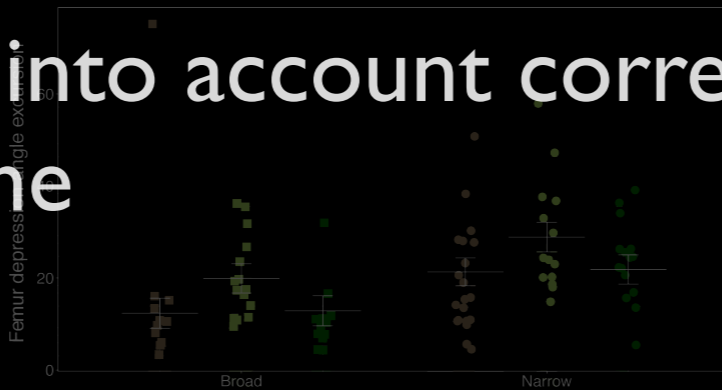
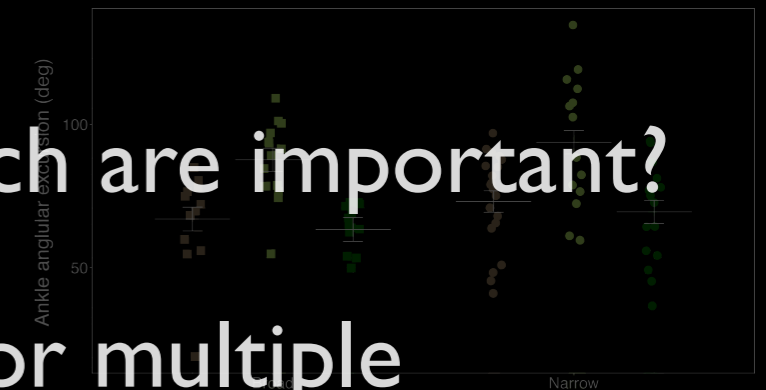
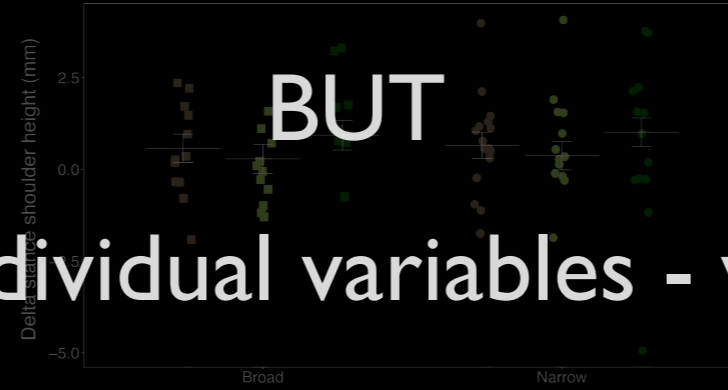
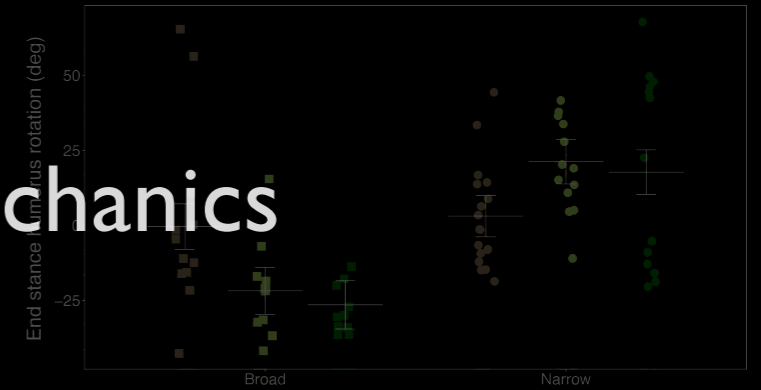
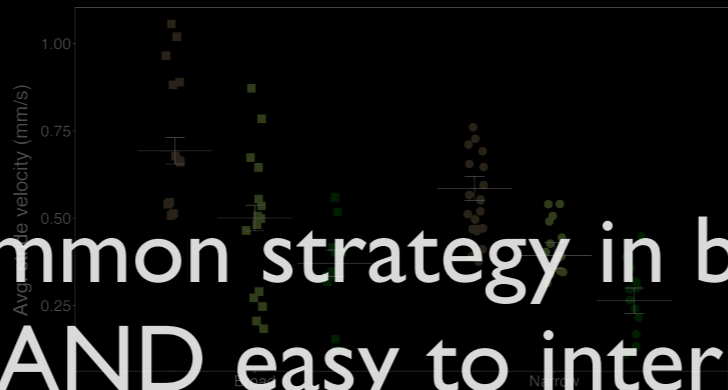
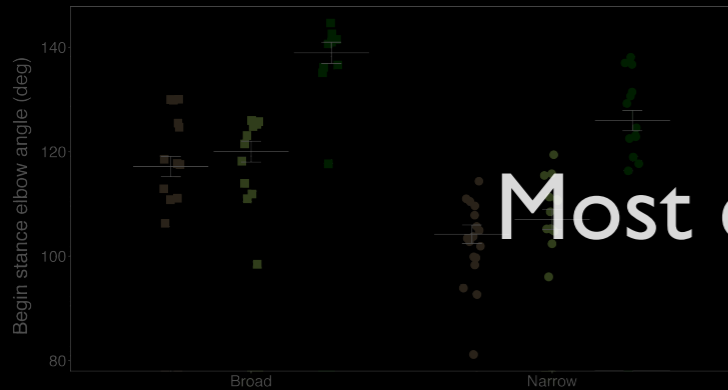


# Two-way mixed model ANOVA

Most common strategy in biomechanics  
AND easy to interpret

BUT

- ◆ There are 72 individual variables - which are important?
- ◆ Need to do some sort of correction for multiple tests since some results must be significant by random chance (e.g. Bonferroni, Benjamini-Hochberg)
- ◆ Does not take into account correlations between variables or time



# Two-way mixed model ANOVA

	% variables significantly affected	
	Perch diameter	Incline
Forelimb	42	64
Hind limb	69	69

- ◆ Significant changes to at least one variable at nearly every joint in response to changes in incline or perch diameter (88 significant results in total)
- ◆ Generally, decreased height above surface, greater limb flexion, slower running speed, and greater duty factor on steeper/narrow surfaces → crouching for stability

# Multivariate Analysis

- ◆ MANOVA

- ◆ to **test** for differences between substrates **when considering all variables together**

- ◆ Canonical Discriminant Analysis (Classical perspective)

- ◆ to visualize maximal group separation and to determine **how all variables together** contribute to group separation

- ◆ Linear Discriminant Analysis (Modern perspective)

- ◆ to classify strides according to environment based on the **combined variables**

# MANOVA

- ◆ Multivariate ANOVA to test the hypothesis that the **vectors** of means for two or more groups are different
- ◆ Tested by calculating the ratio of the multivariate between-group sum of squares to within-group sum of squares → F-distribution



Canonical Discriminant Analysis

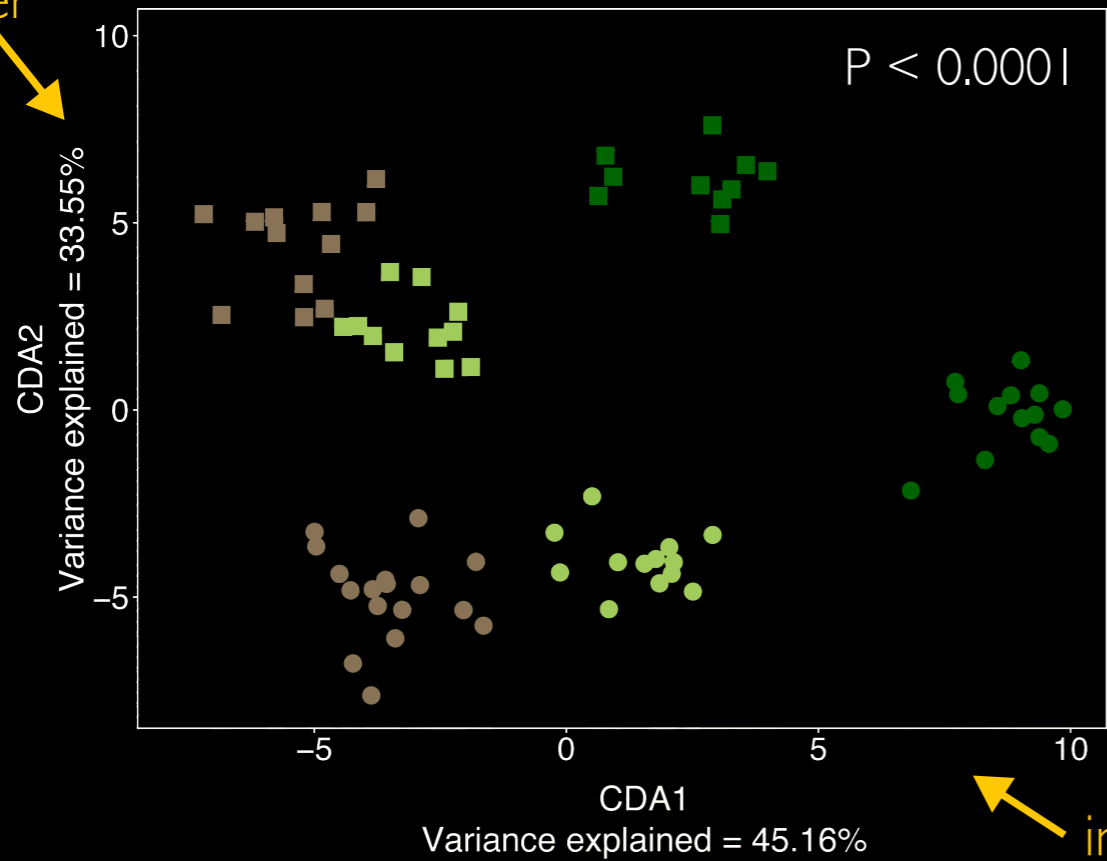
# Canonical Discriminant Analysis

- ◆ Goal: visualization of group separation & determine **how** the groups are separated
- ◆ Eigenanalysis based on variance/covariance matrices between and within user-defined groups
- ◆ Defines “canonical axes” as linear combinations of variables that maximize the between-group variation relative to within-group variation
- ◆ Correlation of variables against CDA axes can be used to interpret the source of the variation contributing to group separation

# Canonical Discriminant Analysis

perch diameter

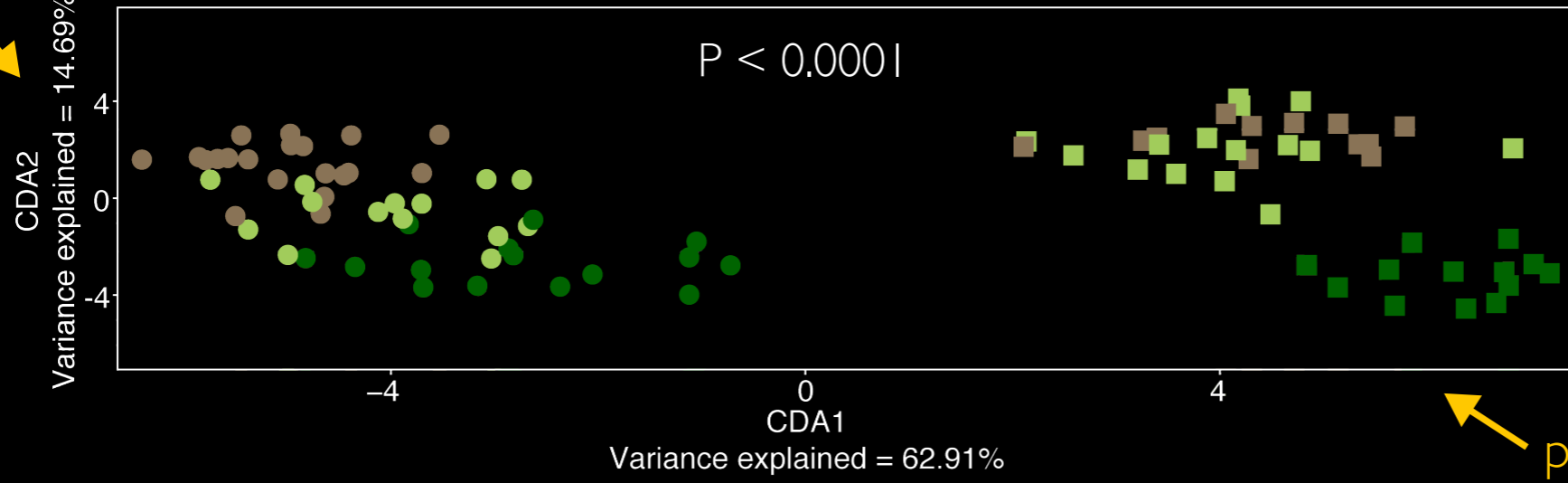
Forelimb



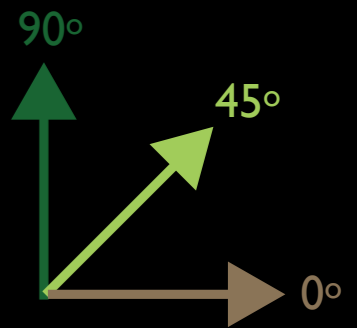
incline

incline

Hind limb



- broad
- narrow



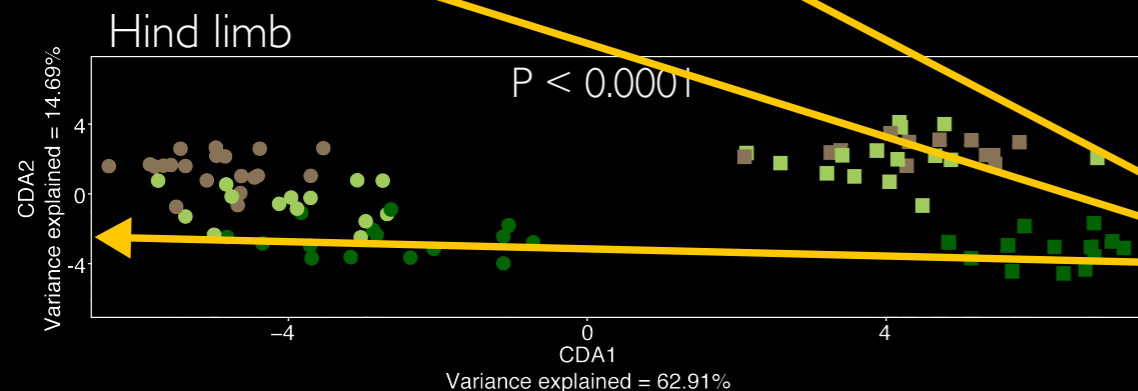
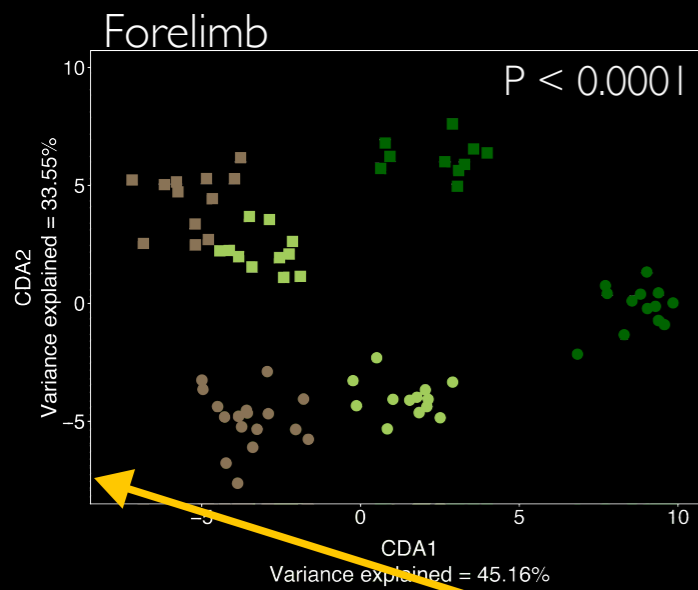
perch diameter

# Canonical Discriminant Analysis

Important variables for group separation:

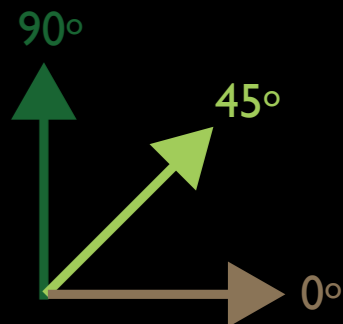
- ♦ Speed, duty factor, stride frequency
- ♦ Shoulder height, elbow angle/angular velocities, humerus depression/rotation
- ♦ Hip height, angles/angular velocities of femur depression/retraction & knee/ankle

**Lizards run more slowly, are more crouched, and swing limbs faster to increase proportion of time feet are in contact with the ground**



Linear combination of variables

■ broad  
● narrow



# Canonical Discriminant Analysis



## Linear Discriminant Analysis

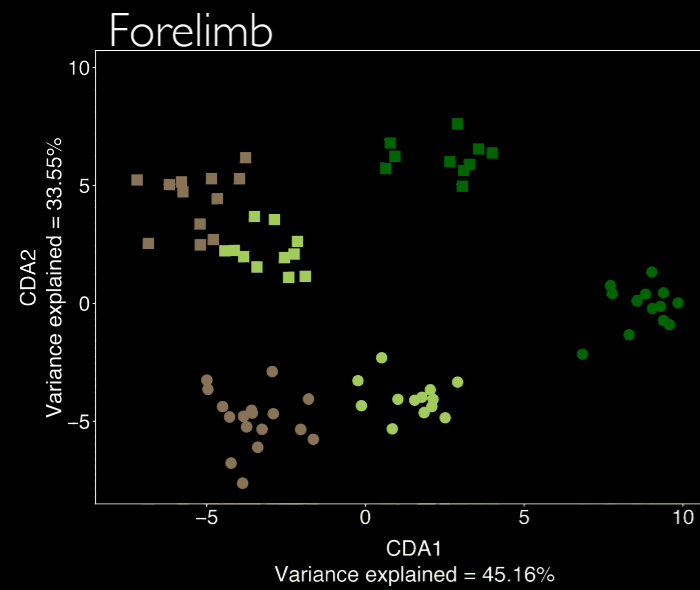
- ◆ CDA maximized the ratio of the between to within-group variation
- ◆ This gives a eigenvector corresponding to the direction of maximal separation between groups and the corresponding separation hyperplane
- ◆ This hyperplane can be used as a decision boundary for classification purposes
- ◆ Under assumptions of normality and equal variance between groups, this hyperplane corresponds to the classification boundary of LDA



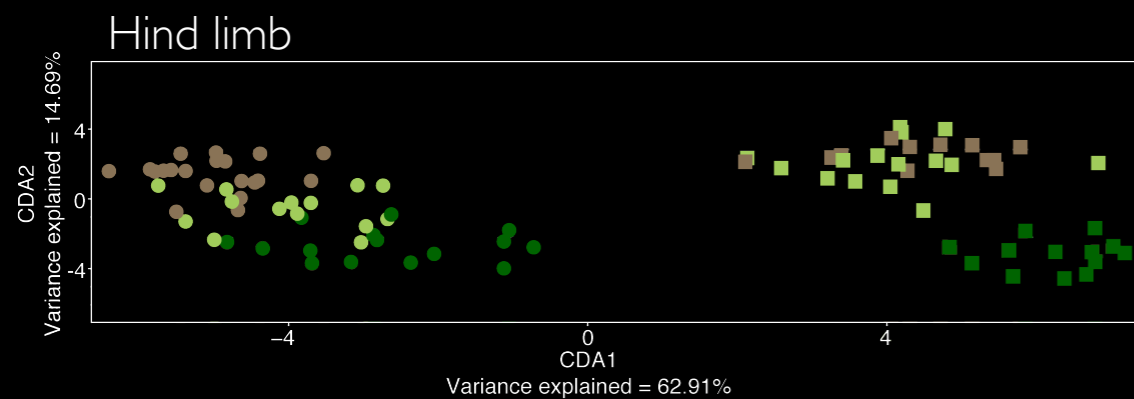
# Linear Discriminant Analysis

- ♦ Goal: Classify the observations (strides) into the correct groups (environments)
- ♦ Classification model assessed via Leave-One-Out Cross-validation
- ♦ Classification accuracy + Confusion Matrix to assess the quality of the classification model both overall and by group

# Linear Discriminant Analysis

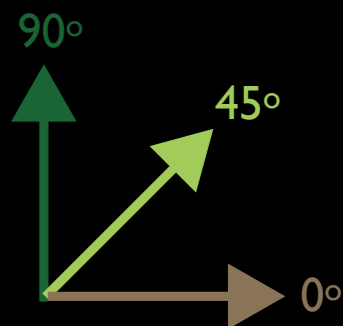


% observations  
correctly classified = 93.4%

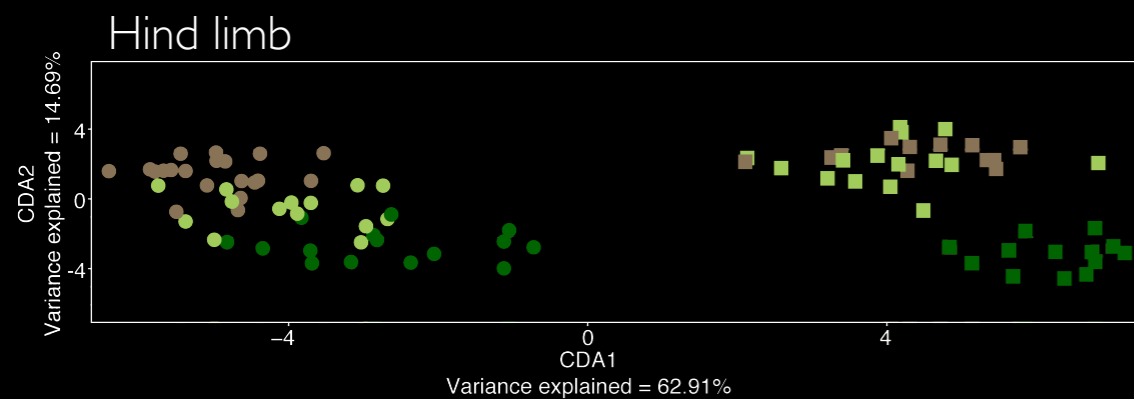
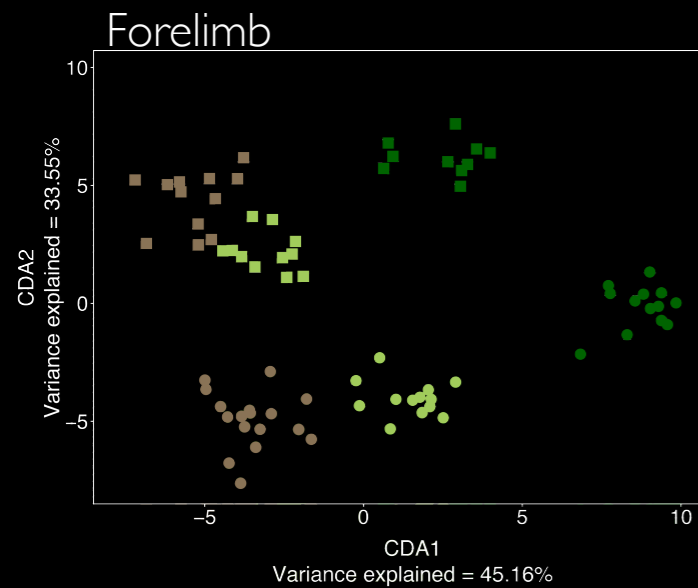


% observations  
correctly classified = 84.3%

- broad
- narrow



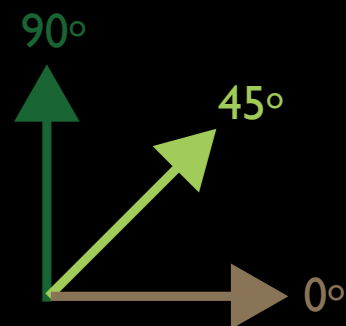
# Linear Discriminant Analysis



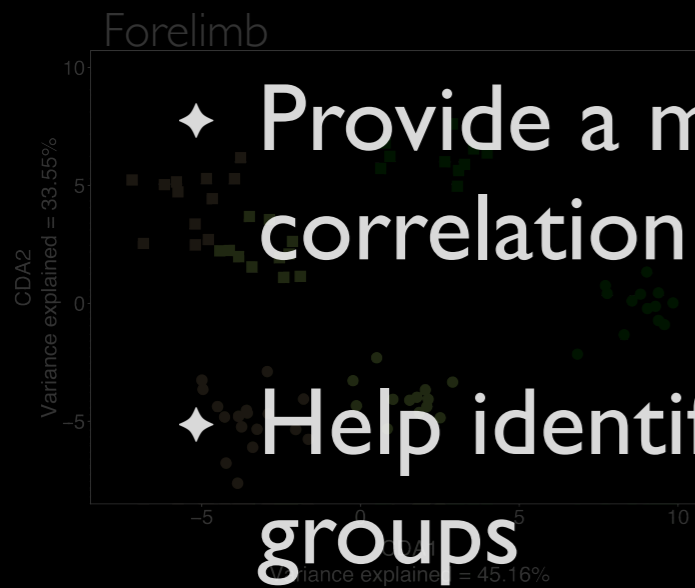
% observations correctly classified

	Forelimb	Hind limb
Angles only	82.9%	84.3%
Angular velocities only	68.4%	58.4%
All variables	93.4%	84.3%

■ broad  
● narrow

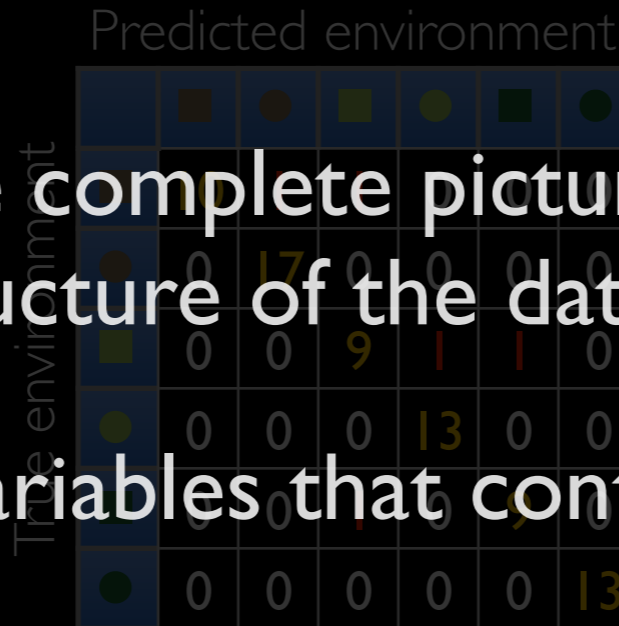


# Linear Discriminant Analysis Multivariate Analyses



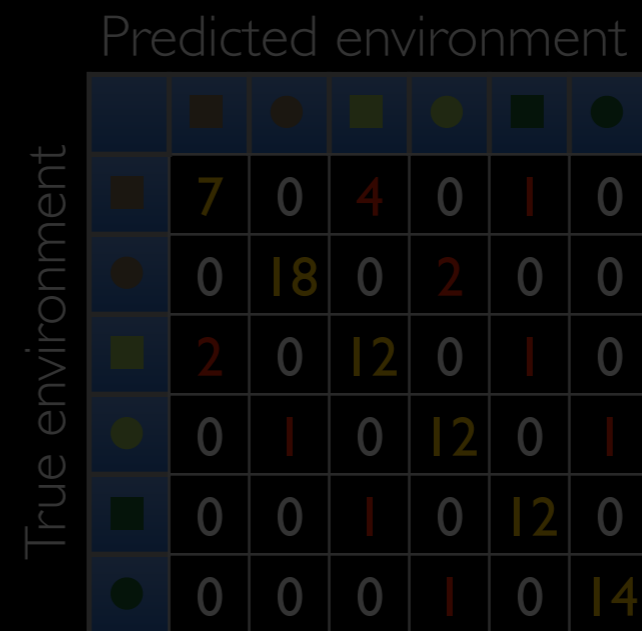
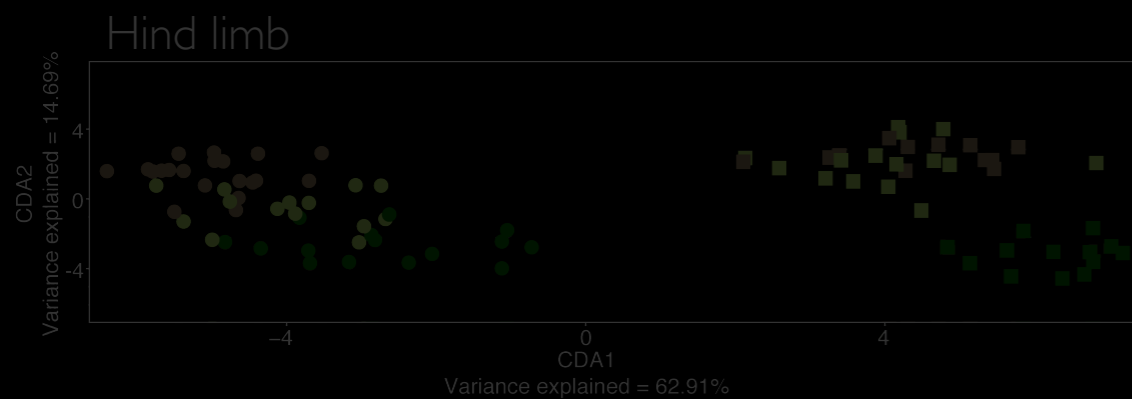
◆ Provide a more complete picture that incorporates correlation structure of the dataset

◆ Help identify variables that contribute to definitions of

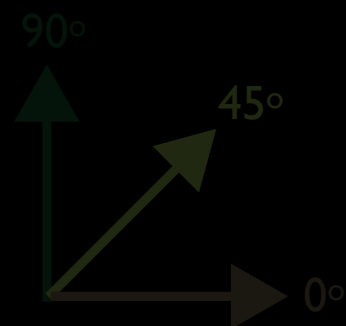


classifies all observations well

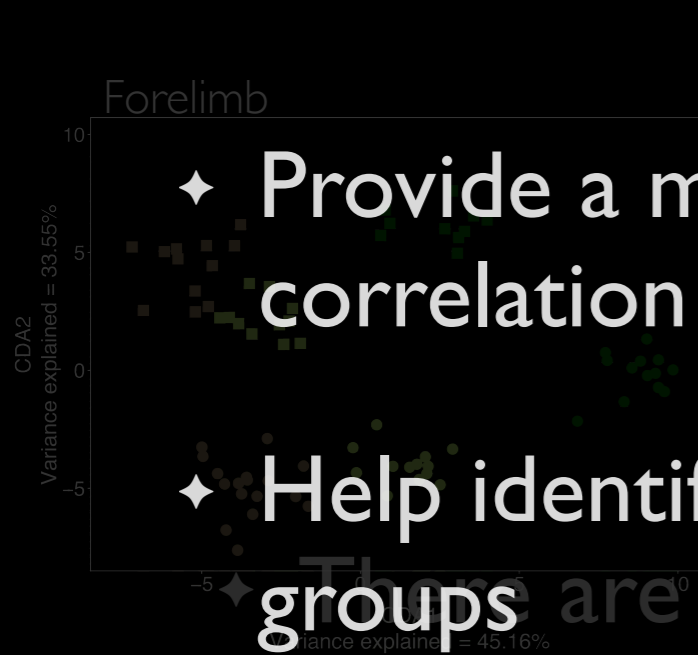
classifies diameter perfectly  
& only makes mistakes  
between inclines



■ broad  
● narrow



# Linear Discriminant Analysis Multivariate Analyses



◆ Provide a more complete picture that incorporates correlation structure of the dataset

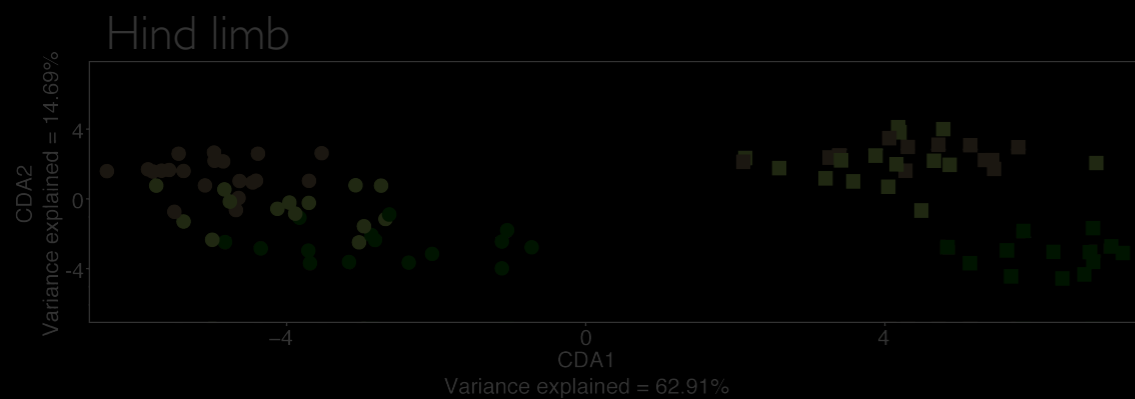
◆ Help identify variables that contribute to definitions of groups

Predicted environment

True environment	■	●	■	●	■	●
■	17	0	0	0	0	0
●	0	9	1	1	0	0
■	0	0	13	0	0	0
●	0	0	0	13	0	0
■	0	0	0	0	13	0
●	0	0	0	0	0	13

classifies all observations well

classifies diameter perfectly & only makes mistakes between inclines

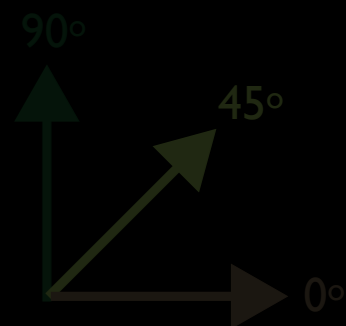


BUT

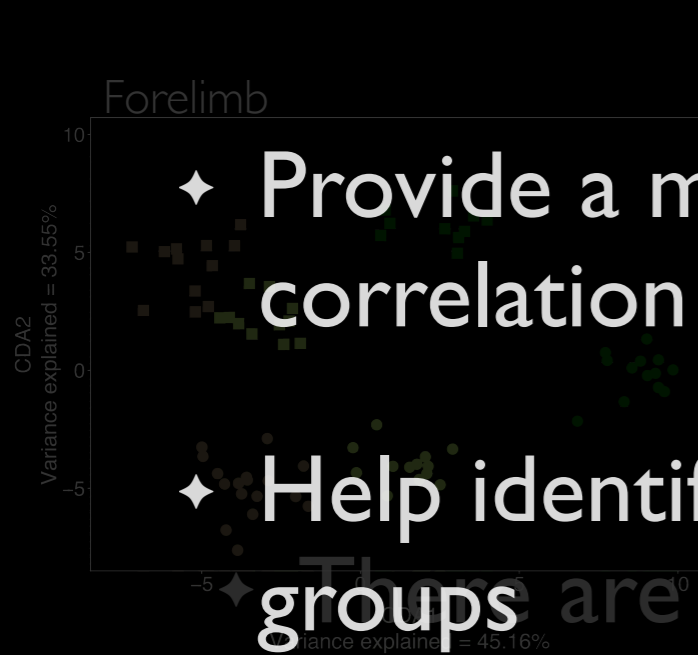
Predicted environment

True environment	■	●	■	●	■	●
■	7	0	4	0	1	0
●	0	18	0	2	0	0
■	2	0	12	0	1	0
●	0	1	0	12	0	1
■	0	0	1	0	12	0
●	0	0	0	1	0	14

■ broad  
● narrow



# Linear Discriminant Analysis Multivariate Analyses



◆ Provide a more complete picture that incorporates correlation structure of the dataset

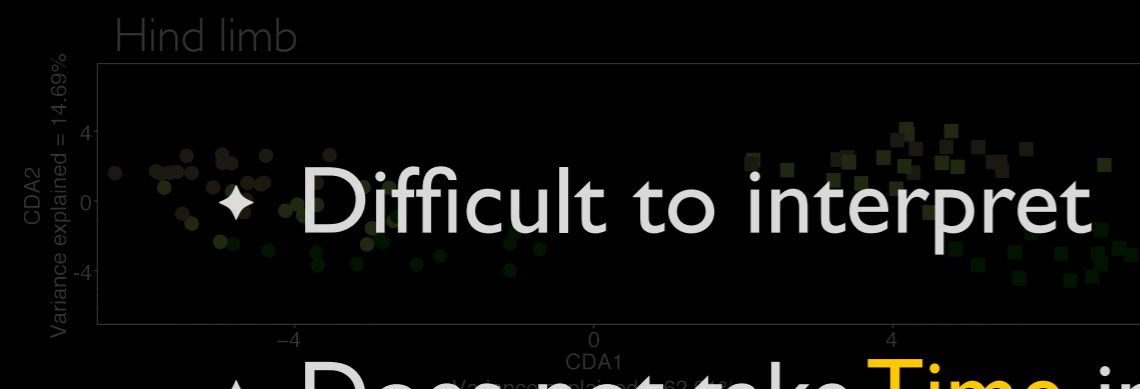
◆ Help identify variables that contribute to definitions of groups

Predicted environment

True environment	■	●	■	●	■	●
■	17	0	0	0	0	0
●	0	9	1	1	0	0
■	0	0	13	0	0	0
●	0	0	0	13	0	0
■	0	0	0	0	13	0
●	0	0	0	0	0	13

classifies all observations well

classifies diameter perfectly & only makes mistakes between inclines



◆ Difficult to interpret

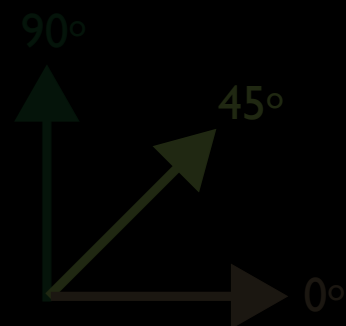
◆ Does not take **Time** into account

BUT

Predicted environment

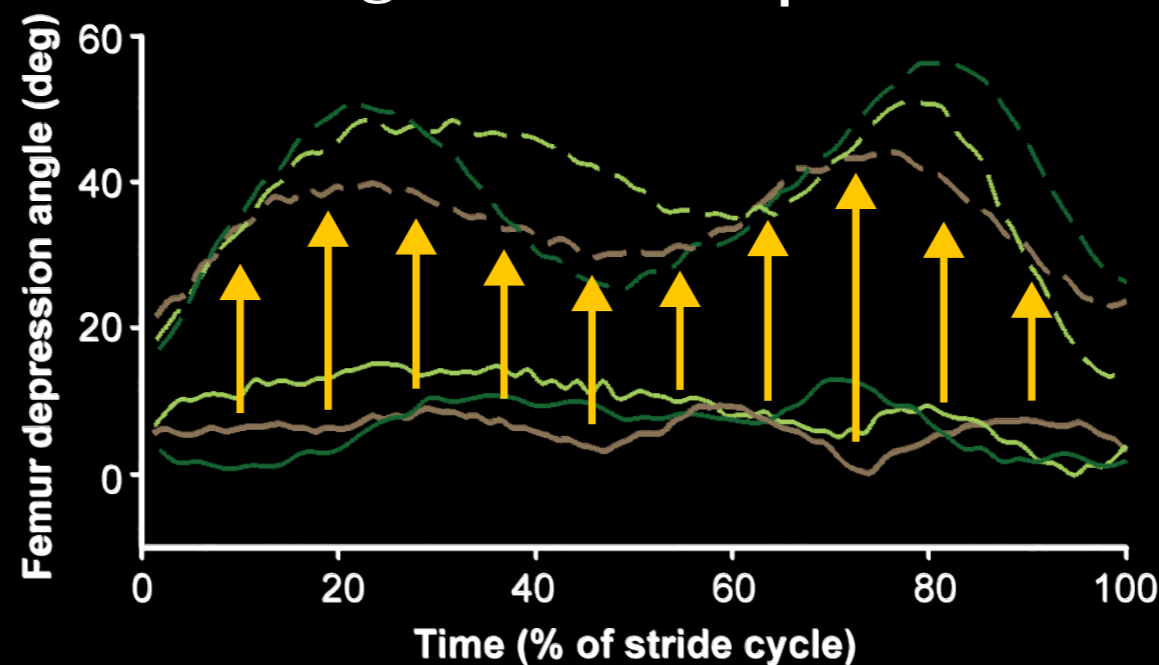
True environment	■	●	■	●	■	●
■	7	0	4	0	1	0
●	0	18	0	2	0	0
■	2	0	12	0	1	0
●	0	1	0	12	0	1
■	0	0	1	0	12	0
●	0	0	0	1	0	14

■ broad  
● narrow

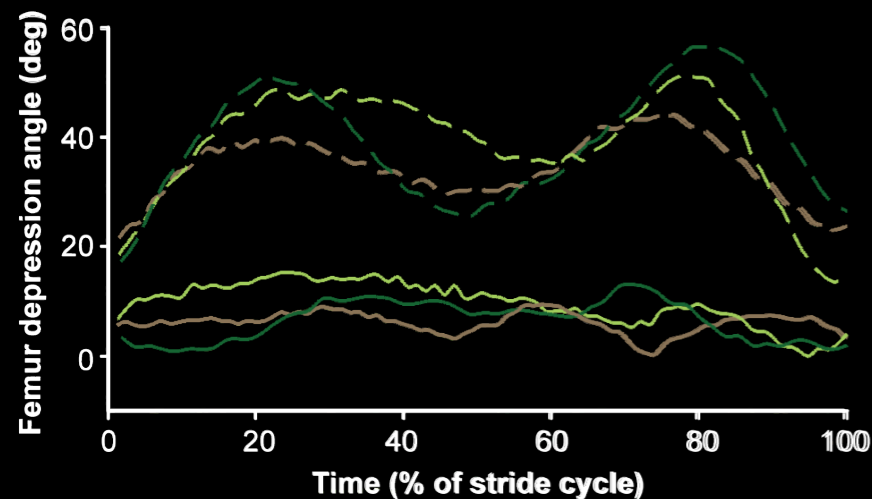
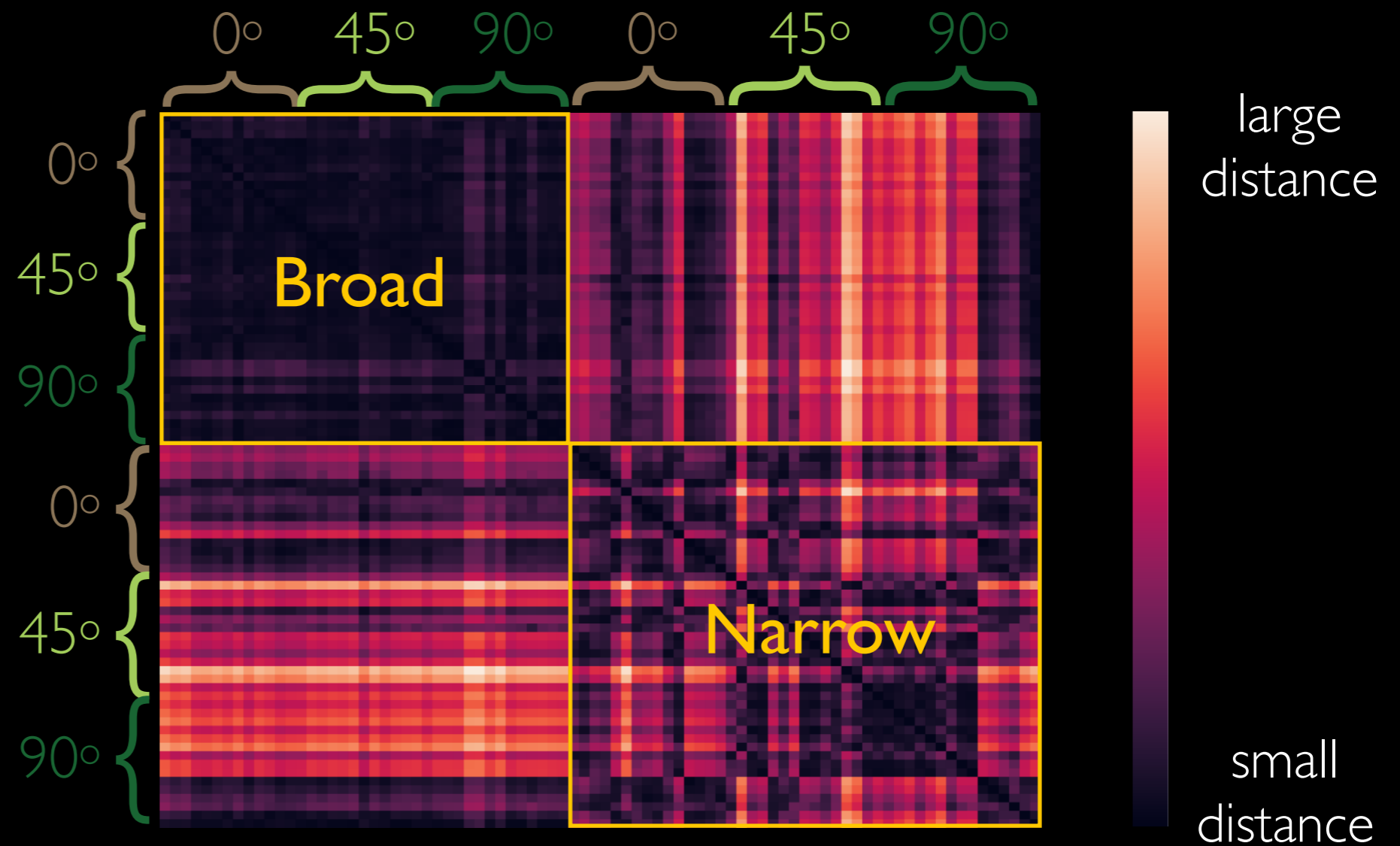


# Dynamic Time Warping

- ◆ Goal: to measure & visualize the distance between time series of joint angles of each stride
- ◆ For each joint angle, calculate the distance between each instant in time during one stride and the corresponding time point in every other stride (allows non-linear alignment for time series of different length)
- ◆ The cumulative distance between time series of each stride can be visualized using a **Heat Map**



# Dynamic Time Warping - Heat Map



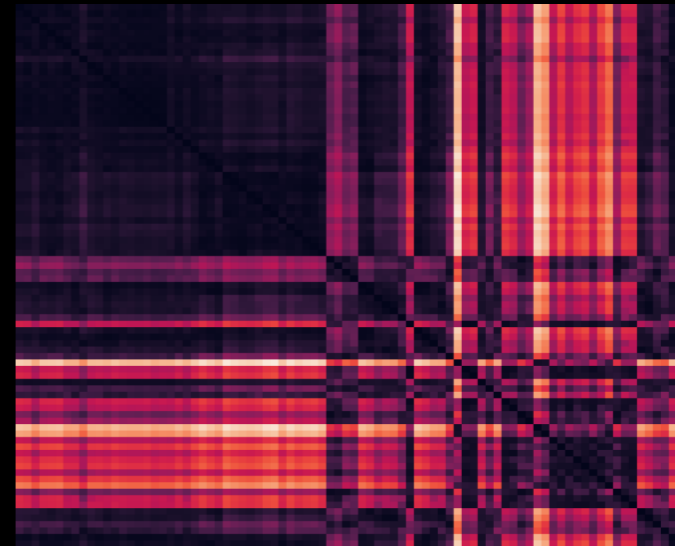


# Dynamic Time Warping - Heat Map

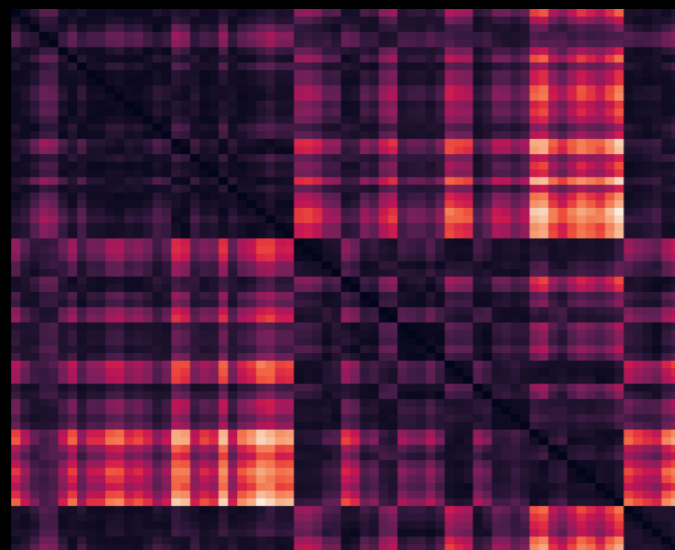
Femur depression



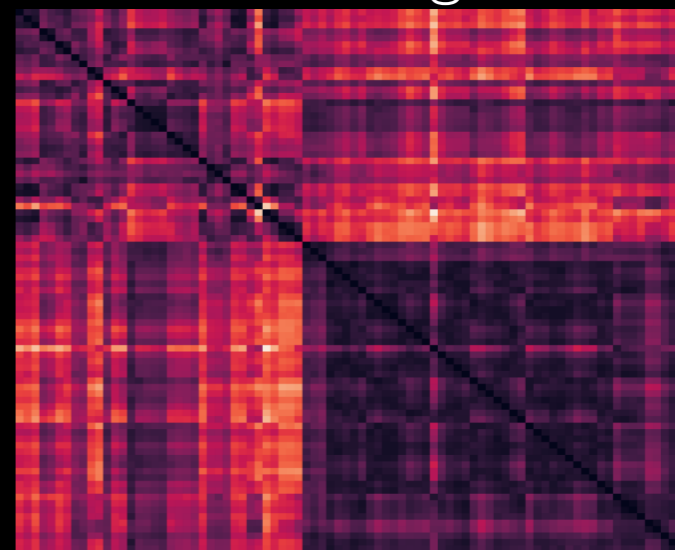
Femur rotation



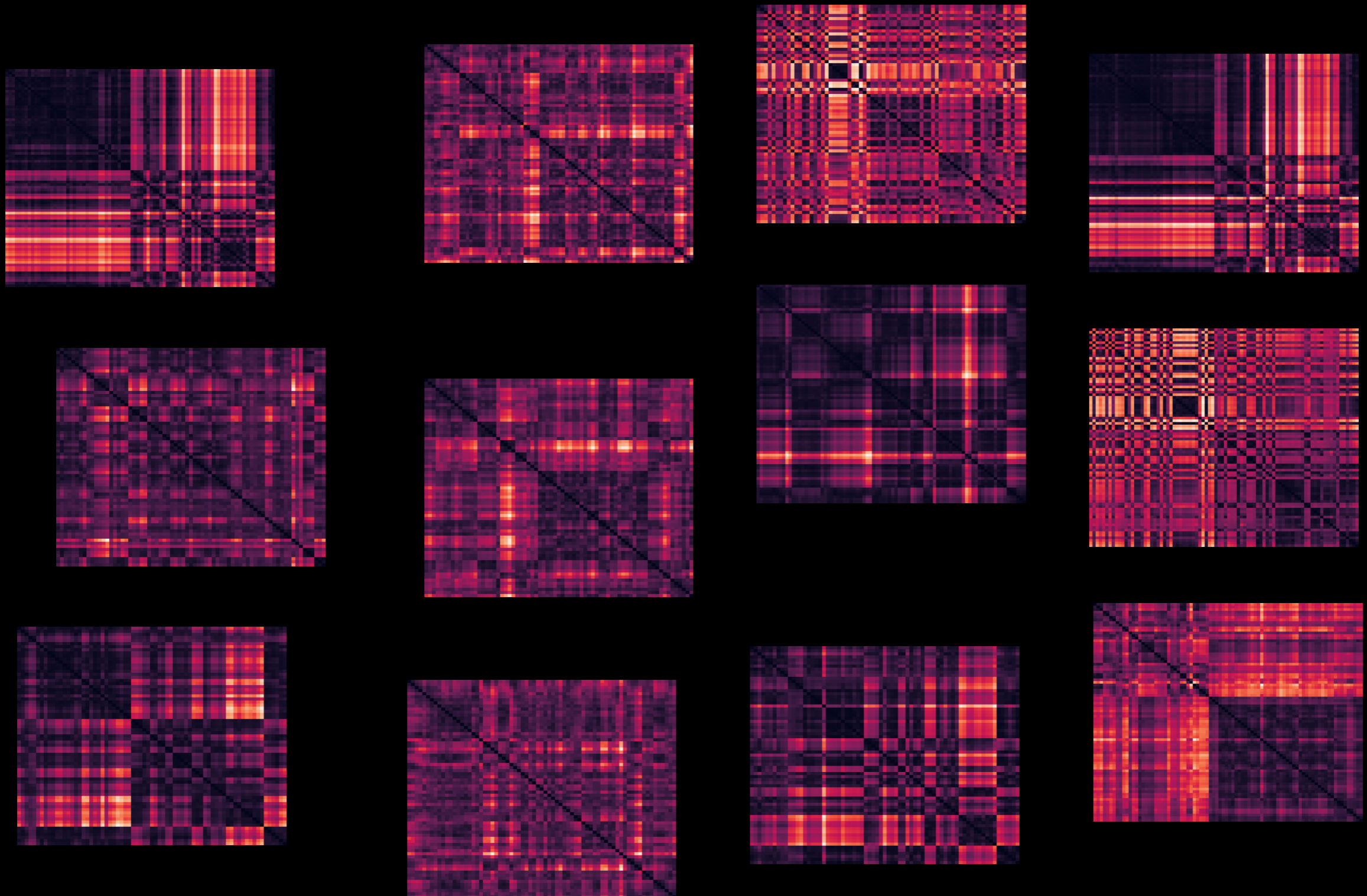
Humerus rotation



Knee angle



# Dynamic Time Warping - Heat Map



# Take home messages

- ◆ Environment clearly impacts the locomotor behavior of *Anolis* lizards
- ◆ The choice of statistical method has significant effect on the information we can extract from this dataset
- ◆ Although univariate methods can provide answers that are easily interpretable, more sophisticated and modern multivariate, classification, and time series methods can provide new, valuable insights that could change how biomechanists approach their data analysis



# Questions?



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