

Big Brown Bat (*Eptesicus fuscus*) Post Hibernation Mass Changes During White-Nose Syndrome Invasion

Presentation by:

Yunjung Jo (1008751138), Victoria Cordova-Morote
(1006907493), Andrew Batmunkh (1007610257)



Photo by: J. Scott Altenbach

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Today's Presentation

- Introduction/Background
- Methods
- Results
- Discussion



Background/Introduction:



Image by Michael Durham:
<https://www.batcon.org/bat/eptesicus-fuscus/>

What is White Nose Syndrome?

- White Nose Syndrome (WNS) is a disease impacting hibernating bat species caused by the fungal pathogen:
 - *Pseudogymnoascus destructans* (Pd)
- Pd grows in temperatures between 12-16°C; Pd thrives in bat winter hibernacula conditions
- as *Heterothermic mammals*: bat species are vulnerable during torpor/hibernation w suppressed immune functioning.



Little Brown Bat with WNS fungus (New York, Oct. 2008.)

WNS and Mass Changes



Little Brown Bat with WNS
fungus (New York, Oct. 2008.)

- *Pd* impacts the epithelial tissues of the bats, leaving an uncomfortable white fuzz and fissures on their muzzles and wings (pictured).
- The arousal of *Pd* infection wakes them up and prematurely burns their brown fat and water, **leading to associated losses in mass.**

Seasonal Cycles:



Little Brown Bat with WNS
fungus (New York, Oct. 2008.)

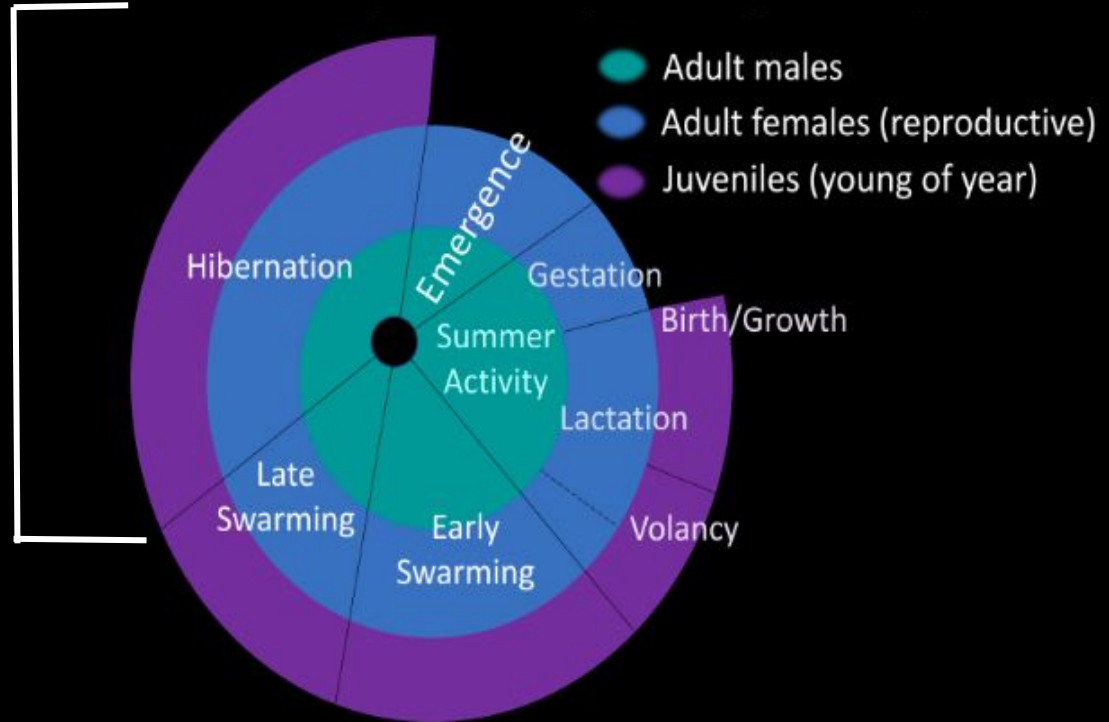


Chart by: Dr. Davy (2020)

Seasonal Cycles:

Pd exposure has carry-over effects in the **mass** of recovered individuals, where:

- Energy expenditure of *Pd* immune responses/stress decreases available energy for emergent seasonal behaviors such as **migration to summer hibernacula** and **initiation of reproduction**; w/ differing energy necessities between adult and juvenile **male** and **female** bats.

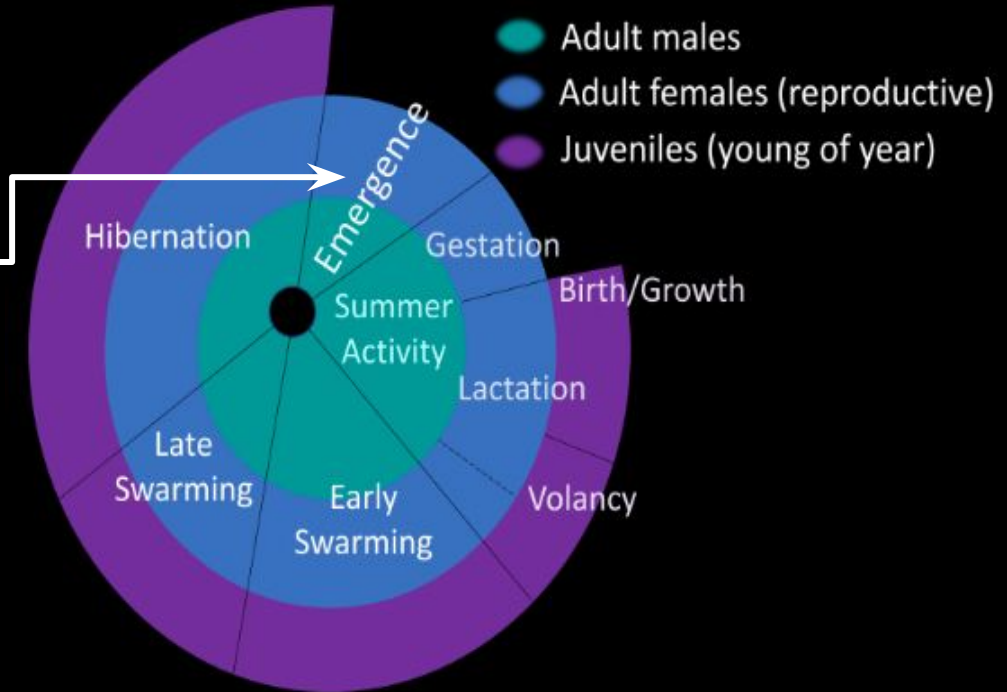


Chart by: Dr. Davy (2020)

Bats of Ontario – local hibernators



Little Brown Bat



Northern Myotis



**Eastern
Small-footed
Bat**



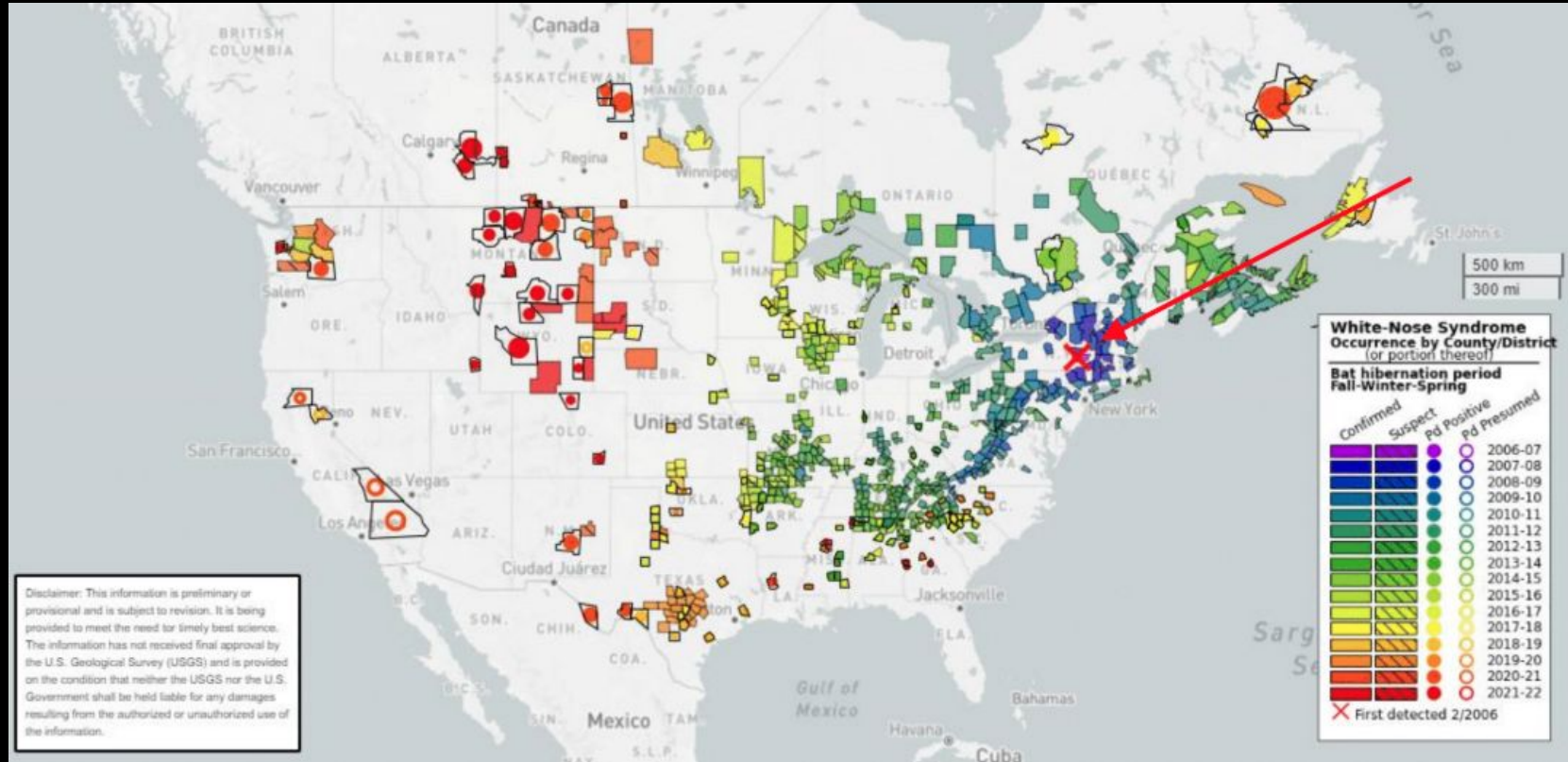
Tricolored Bat



Big Brown Bat
EPFU

Photos: Charles Francis / Brock and Sherri Fentor

Spread of Fungus/Disease Over Time



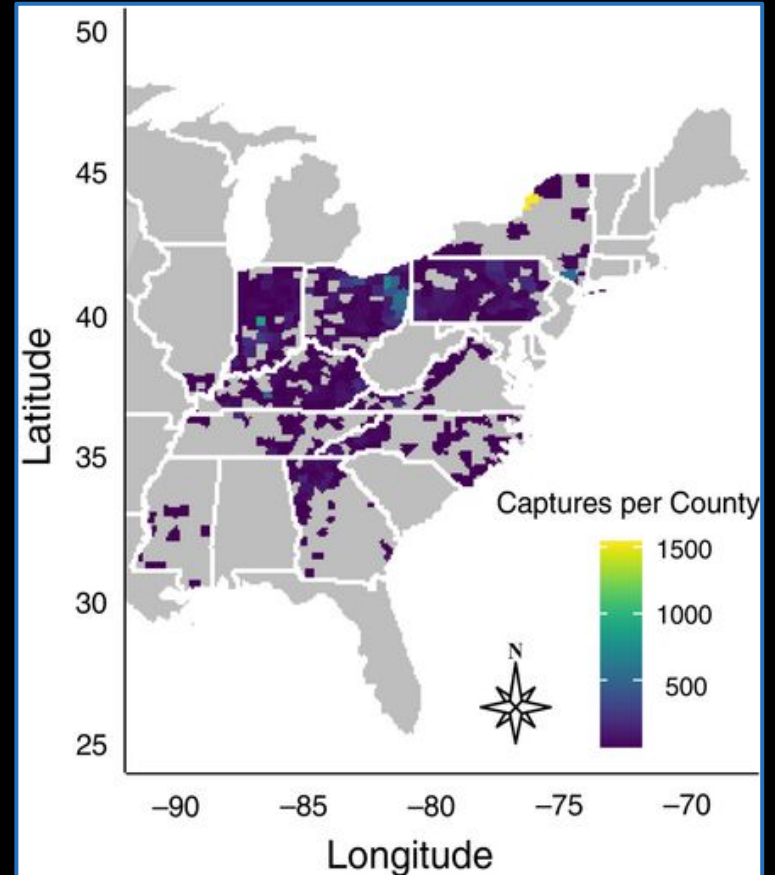
Methodology: Process/Methods



Image by Sherri & Brock Fenton
<https://norfolknaturalist.ca/2023>

About Our Data:

- Big brown bat (EPFU) captured & recorded from 1990-2010 by Simonis et al., (2023)
- 30,497 individual records across 3,797 unique capture sites across eastern United States
- Age, sex, reproductive status, mass, forearm length, etc.
- Focus on mass (more directly impacted by WNS)



Process:

- **Attempt 1**
 - Gamma distribution. Attempted to find MLE, where parameters are maximized (shape, scale)
 - Mean = shape * scale
 - Variance = shape * scale²
 - CI: lower shape * lower scale – upper shape * upper scale
- **Attempt 2**
 - Try to model using different types of distributions
- **Attempt 3**
 - GLMMs using gamma distribution



Photo by: J. Scott Altenbach

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Hypotheses:

- Null hypothesis
 - Age and sex are not correlated with the effect of Pd on mass of EPFU
- Alternative hypothesis
 - Age and sex are correlated with the effect of Pd on mass of EPFU
- Prediction:
 - Mass and mass variation fitted on a **Gamma distribution** will decrease over Pd exposure time

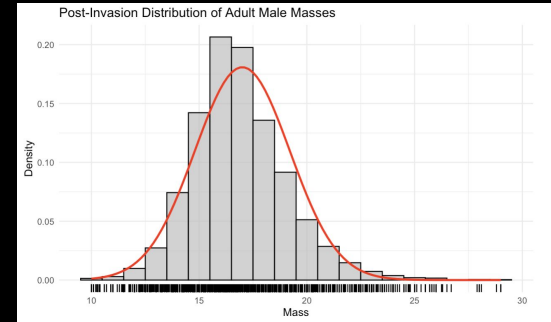


Photo by: J. Scott Altenbach

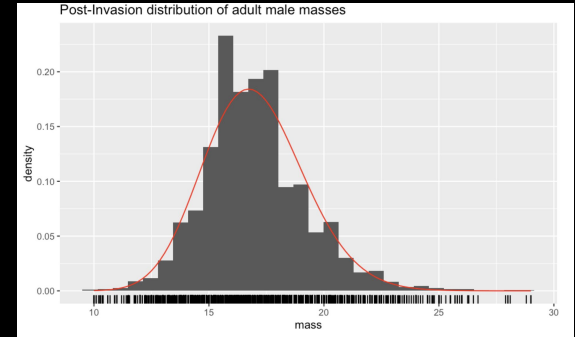
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Model Section: Gamma vs Normal distribution

- Original study fit linear mixed models with normal distributions
- Generalized Linear Mixed Model using Gamma distribution → `glmer()`
- **Gamma distribution** AIC = 155189
- **Normal distribution** AIC = 156321
- Gamma distribution more appropriate than normal (**lower AIC**)



*Figure x: Distribution modeled using a **normal distribution** for Post-Invasion Adult Male Masses.*



*Figure x: Distribution modeled using a **gamma distribution** for Post-Invasion Adult Male Masses.*

Methods of Analysis:

- Link function: “log”
 - multiplicative effects → additive on log scale
 - better than others
- Trying different fixed and random effects, looking at coefficients for effects on mass
- Elevation as a fixed effect → unable to obtain data



Photo by: J. Scott Altenbach

https://georgiabiodiversity.org/portal/profile?es_id=21048&group=bats

A thermal image showing a large colony of bats in flight against a dark background. The bats appear as bright, glowing shapes, with a large, prominent bat in the upper left and many smaller ones scattered throughout the lower half of the frame. The text "Results/Discussion" is overlaid on the right side of the image.

Results/Discussion

<https://www.animalcapturewildlifecontrol.com/blog/big-brown-bats-101-understanding-our-nocturnal-neighbors/>

Comparison of Random Effects

Random Effects Tested:

1. **Model 1:** Random intercept for year
2. **Model 2:** Random intercept for site_mask.
3. **Model 3:** Random intercept for state

Model	Random Effect	AIC
Model 1	year	138844.5
Model 2	site_mask	132567.2
Model 3	state	137431.1

Figure x: *Model Comparison: Site_mask model fits bat mass data best.*

Model 2 (Random Effect: site_mask) has the lowest AIC, indicating it provides the **best fit for the data.**



Results: Gamma GLMM with Random Effects

```
Family: Gamma ( log )
Formula: mass ~ disease_group * sex * pregnancy_status * age + (1 | site_mask)
Data: data_new

      AIC      BIC   logLik deviance df.resid
132567.2 132667.1 -66271.6 132543.2    30484

Scaled residuals:
    Min       1Q   Median       3Q      Max
-4.7955 -0.6163 -0.0102  0.5994  6.4132

Random effects:
 Groups   Name      Variance Std.Dev.
site_mask (Intercept) 0.003297 0.05742
Residual              0.013525 0.11630
Number of obs: 30496, groups: site_mask, 3797

Fixed effects:
              Estimate Std. Error t value Pr(>|z|)
(Intercept)    2.967439   0.002390 1241.774 < 2e-16 ***
disease_groupPre-Invasion 0.015950   0.004951   3.222 0.00128 **
sexmale       -0.138996   0.001839  -75.570 < 2e-16 ***
pregnancy_statusPregnant 0.151456   0.003565  42.486 < 2e-16 ***
agejuvenile   -0.194312   0.002675  -72.630 < 2e-16 ***
disease_groupPre-Invasion:sexmale -0.001418   0.005010  -0.283 0.77723
disease_groupPre-Invasion:pregnancy_statusPregnant -0.028060   0.009989  -2.809 0.00497 **
disease_groupPre-Invasion:agejuvenile 0.021801   0.007604   2.867 0.00415 **
sexmale:agejuvenile 0.072712   0.003635  20.005 < 2e-16 ***
disease_groupPre-Invasion:sexmale:agejuvenile 0.003845   0.010337   0.372 0.70988
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Data Overview

- **Focus:** Pre- and Post-Invasion Bat Mass Differences with **different fixed effects**
- **Predictors:** disease_group, sex, pregnancy_status, age
- **Random Effects:** site_mask
- **Key Fixed Effects:**
 - disease_group Pre-Invasion:
 - pregnancy_status → Pregnant:
 - age → juvenile:

GLMM captures the skewed distribution of bat mass better than LMM.

Figure x: GLMM Summary: Key predictors and variances for bat mass.

Results: Interaction Plot

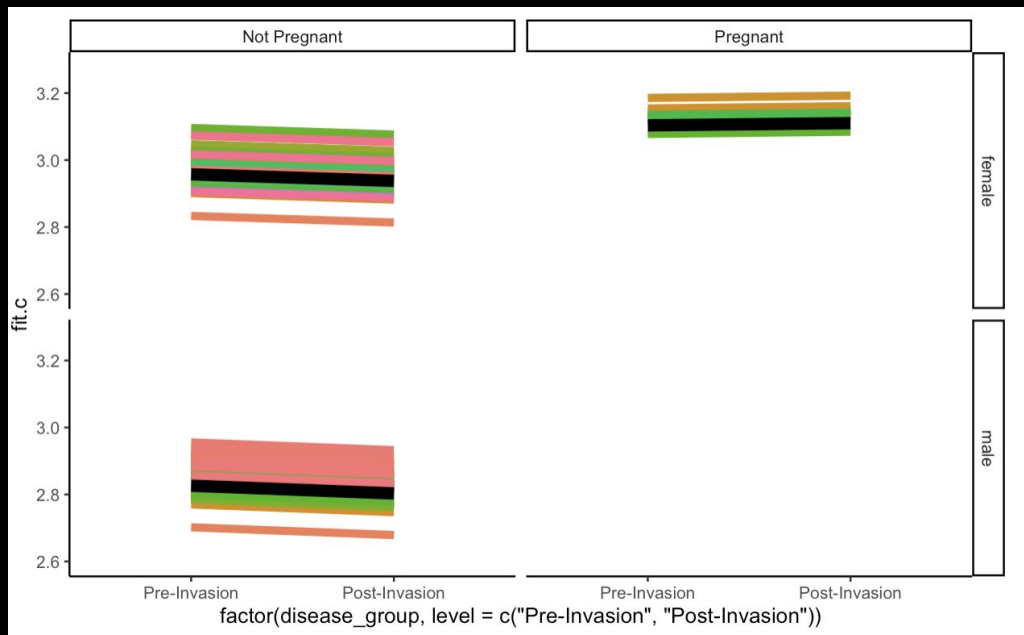


Figure x: Predicted log-mass by disease group, sex, and pregnancy status.

Key Takeaways

- **Gamma GLMM** captures positively skewed mass data.
- Significant predictors include **disease_group**, **pregnancy_status**, and **age**.
- Random effect for **site_mask** accounts for spatial variability.

Conclusion:

- **Alt hypothesis/prediction accepted:**
 - Age and sex are correlated with the effect of Pd on mass of EPFU, with the Gamma as the best fit to account for the skewed distribution of mass compared to the Normal Distribution (i.e. for pregnant status)
- Gamma GLMM accounts for the ecological implications of WNS, where disease introduction (by year), sex, and reproductive status, highlight the understanding of bat mass dynamics across geographic regions during invasion periods.



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Thank you! Questions?

