

Fusing disparate measurement data for forecasting the growth of trees via Hidden Markov Models



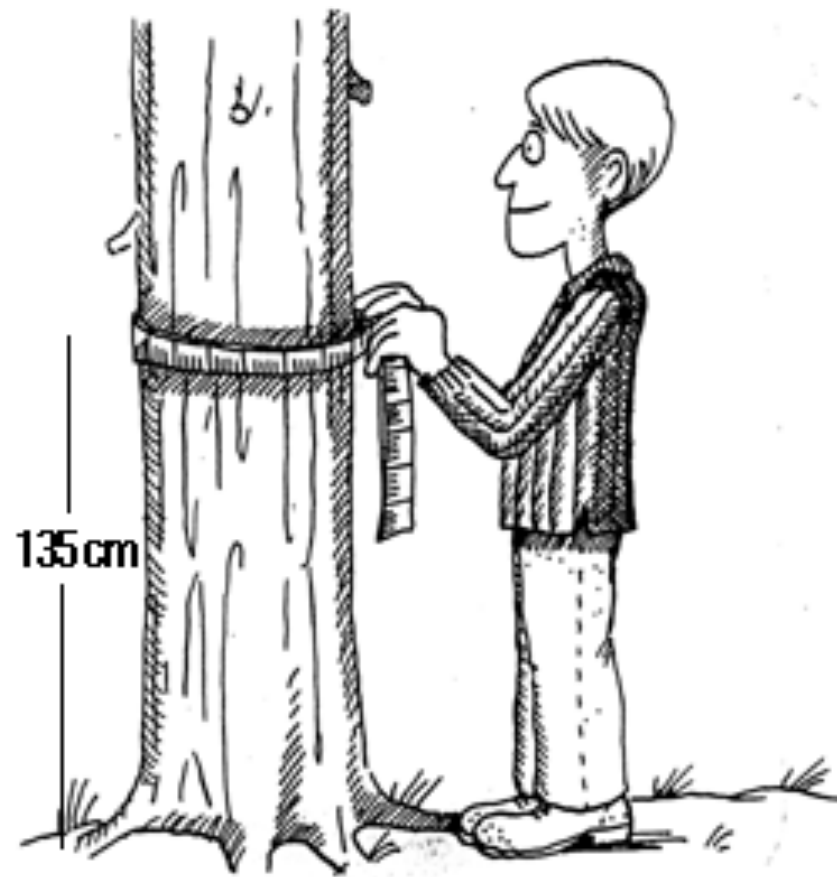
Prof. Albert Y. Kim
UMass Amherst Statistics Seminar Series
Friday, January 22, 2021



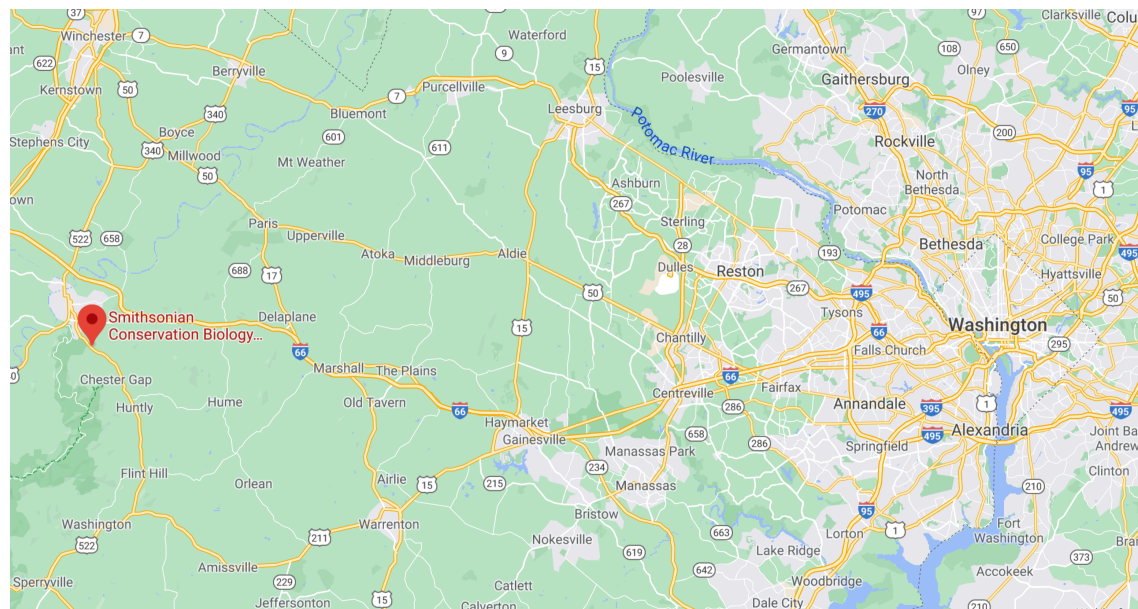
Context

Diameter at Breast Height (dbh)

After species & location, one of the most informative variables about a tree is dbh

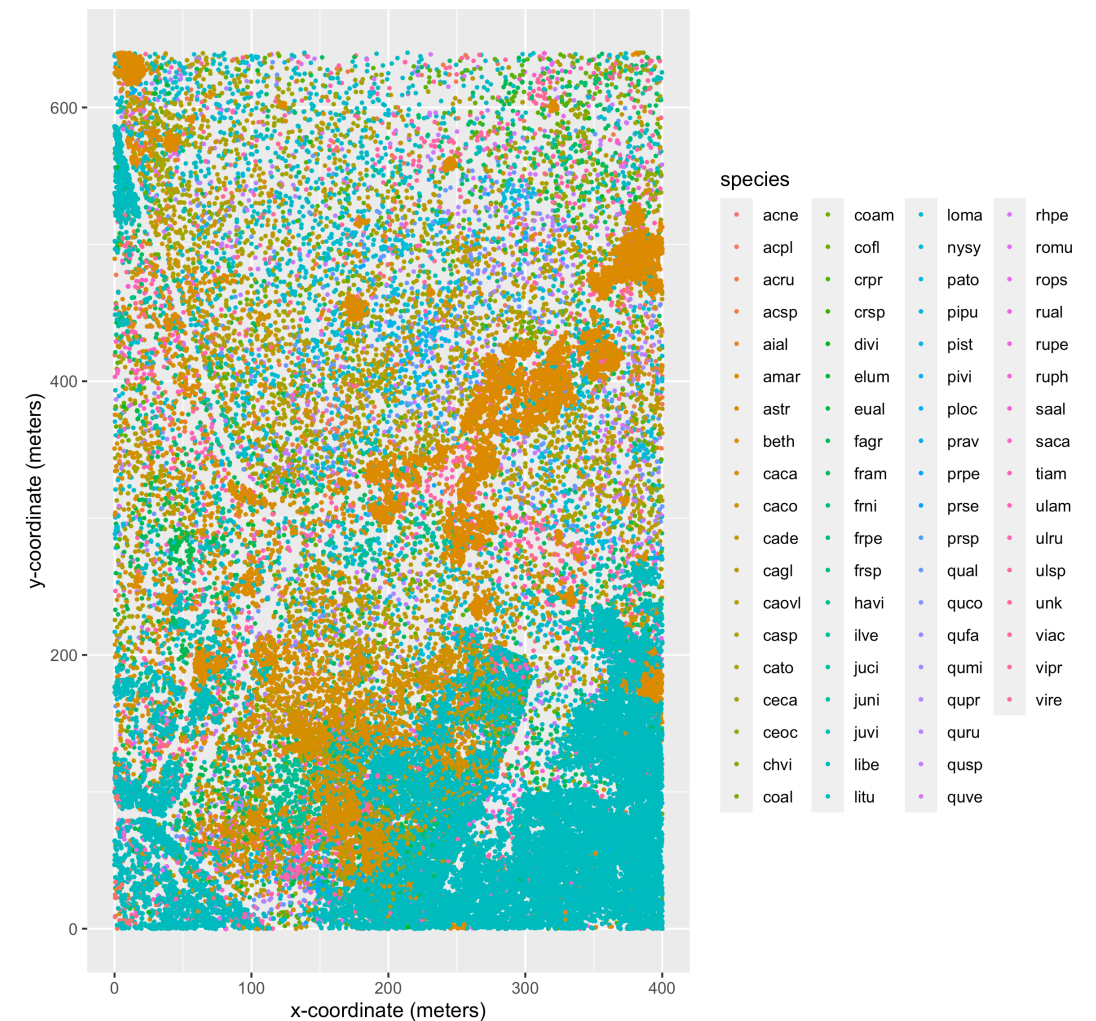


Smithsonian Conservation Biology Institute




25.6 ha = 35.85 soccer fields

Census 2018: 72,555 cataloged trees



dbh > 10mm
are tagged


Data on GitHub





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SCBI-ForestGEO / Dendrobands Unwatch 5 Star 0 Fork 0


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master **Dendrobands / data / scbi.dendroAll_2020.csv** Go to file ...

 **rudeboybert** Replace text month coding with integer month coding for 2019 & 2020. F... ... Latest commit aa41236 on Jul 3 History

4 contributors    

1280 lines (1280 sloc) | 190 KB Raw Blame   

 Search this file...

	tag	stemtag	survey.ID	year	month	day	biannual	intraannual	sp	quadrat	lx	ly	measure	codes	notes
1	10469	1	2020.01	2020	3	11	1	0	litu	109	9.7	1	NA	RE	window too large to measure
2	10587	1	2020.01	2020	3	11	1	0	litu	113	2.6	13	61.41	NA	NA
3	10609	1	2020.01	2020	3	11	1	0	cagl	111	19.5	2.9	81.03	NA	double-checked

Equipment to measure doh



1. Measuring tape. Call this "census" data



2. Tree coring + dendrochronology. Call this "core" data








Equipment to measure doh



3. Dendrobands + Calipers:
Call this "dendro" data



Comparison Chart

	Data source	Measurement	Cost	Sources of Error?
	Census via tape	Diameter	Cheap	Large variation in dbh  technique
	Tree coring	Ring width increment	Expensive	Standardized, cores are dried, no bark effects
	Intraannual dendroband (every 2 weeks)	Increment (from baseline)	High setup, rapid follow-up	Climate induced variation in bark & device (-'ve growth)
	Biannual dendroband (start & end of year)	same	same	+ Less  for: 

Goal 

Can we fuse these disparate data sources into a single model to forecast the growth of trees?

Model

Hidden Markov Models

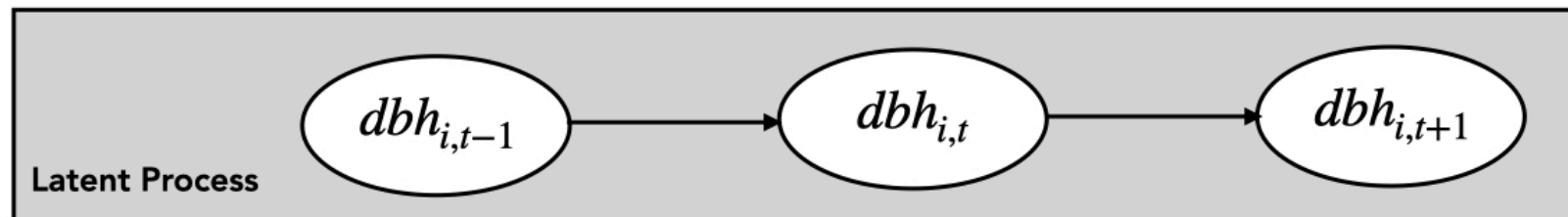
- Hidden: "Data fusion" via latent variables
In our case: "true" dbh
- Markov: y_t depends on y_{t-1}
- Partition sources of error into those that
 - Are not of direct interest
 - Are "one and done" i.e. measurement error
 - Propagate when forecasting

Minimum Viable Product

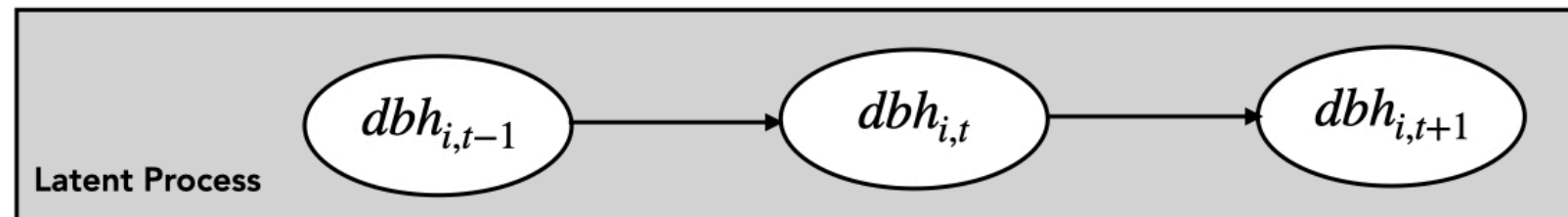
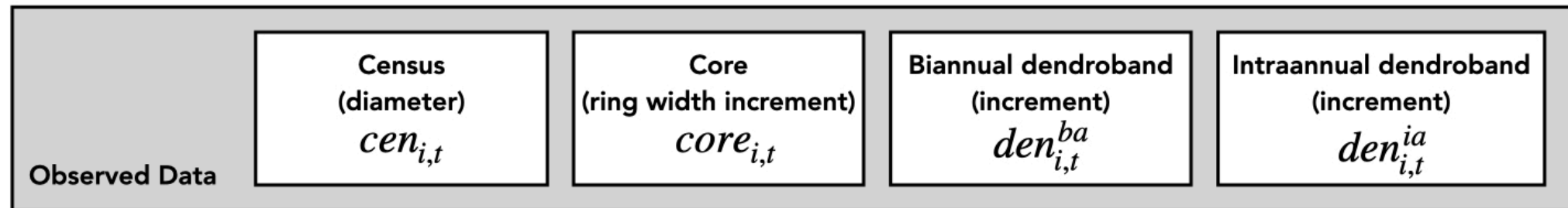
$$dbh_{i,t} = dbh_{i,t-1} + \beta_0 + \beta_i + \beta_t + \epsilon$$

- $dbh_{i,t}$: "True" latent dbh for individual i at time t
- β_0 : Baseline growth
- β_i : Individual tree i random effect
- β_t : Time point t random effect
- $\epsilon \sim \text{Normal}(0, \sigma_\epsilon^2)$

Model

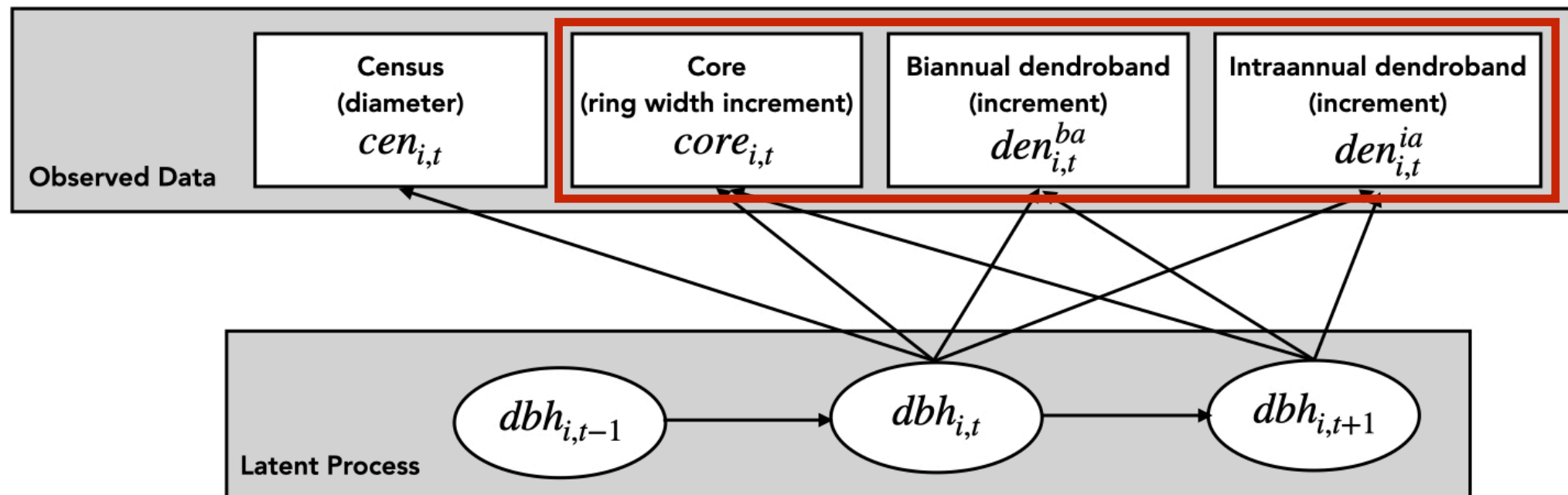


Model

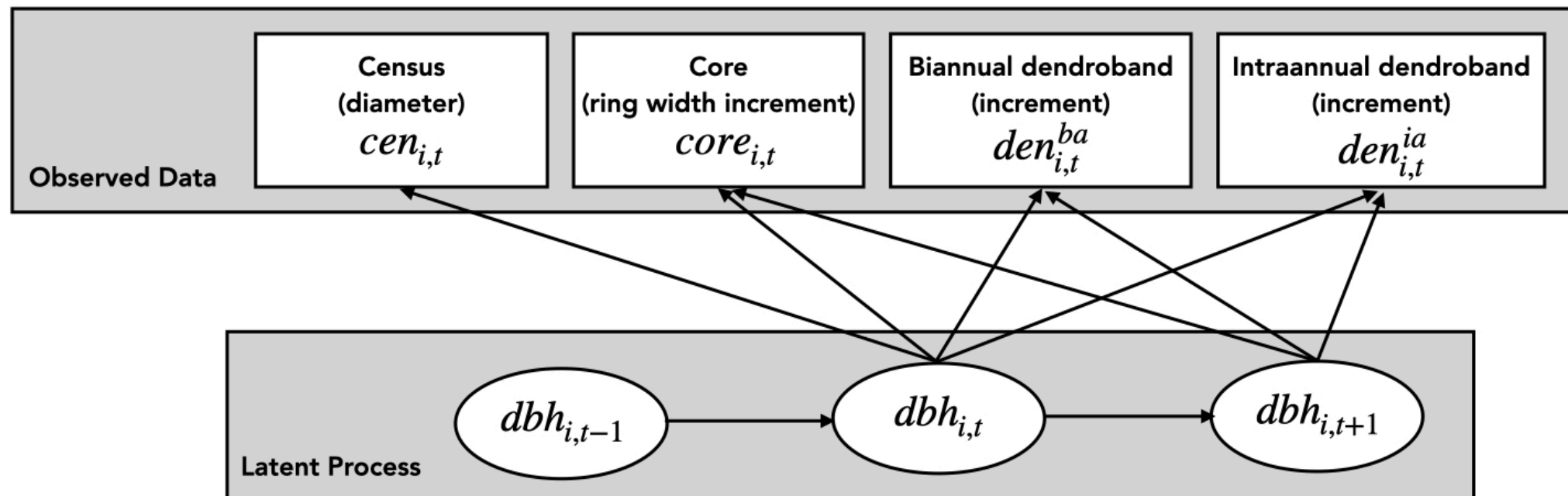


Model

$$\text{Increments} = dbh_{i,t-1} - dbh_{i,t}$$

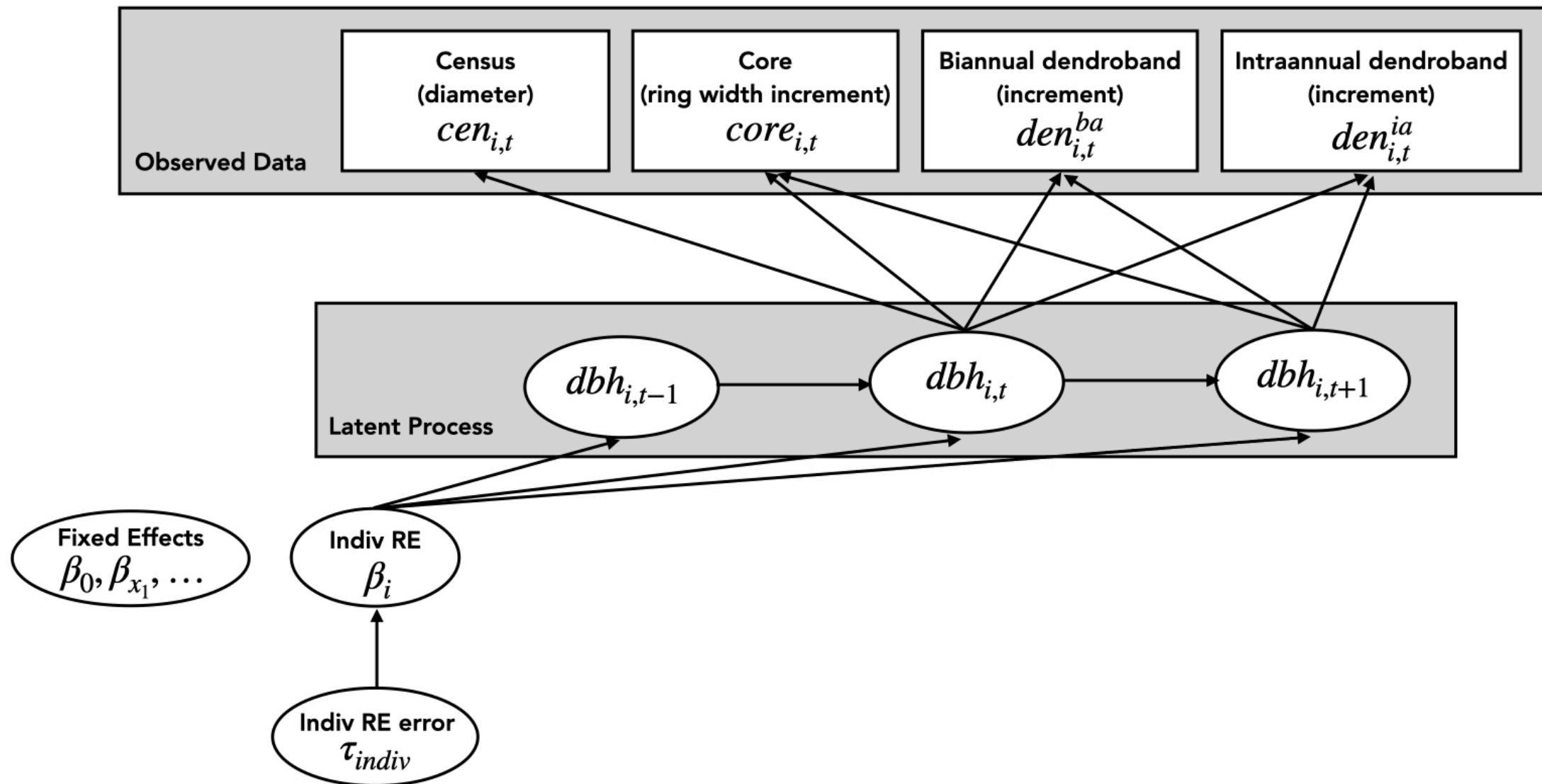


Model



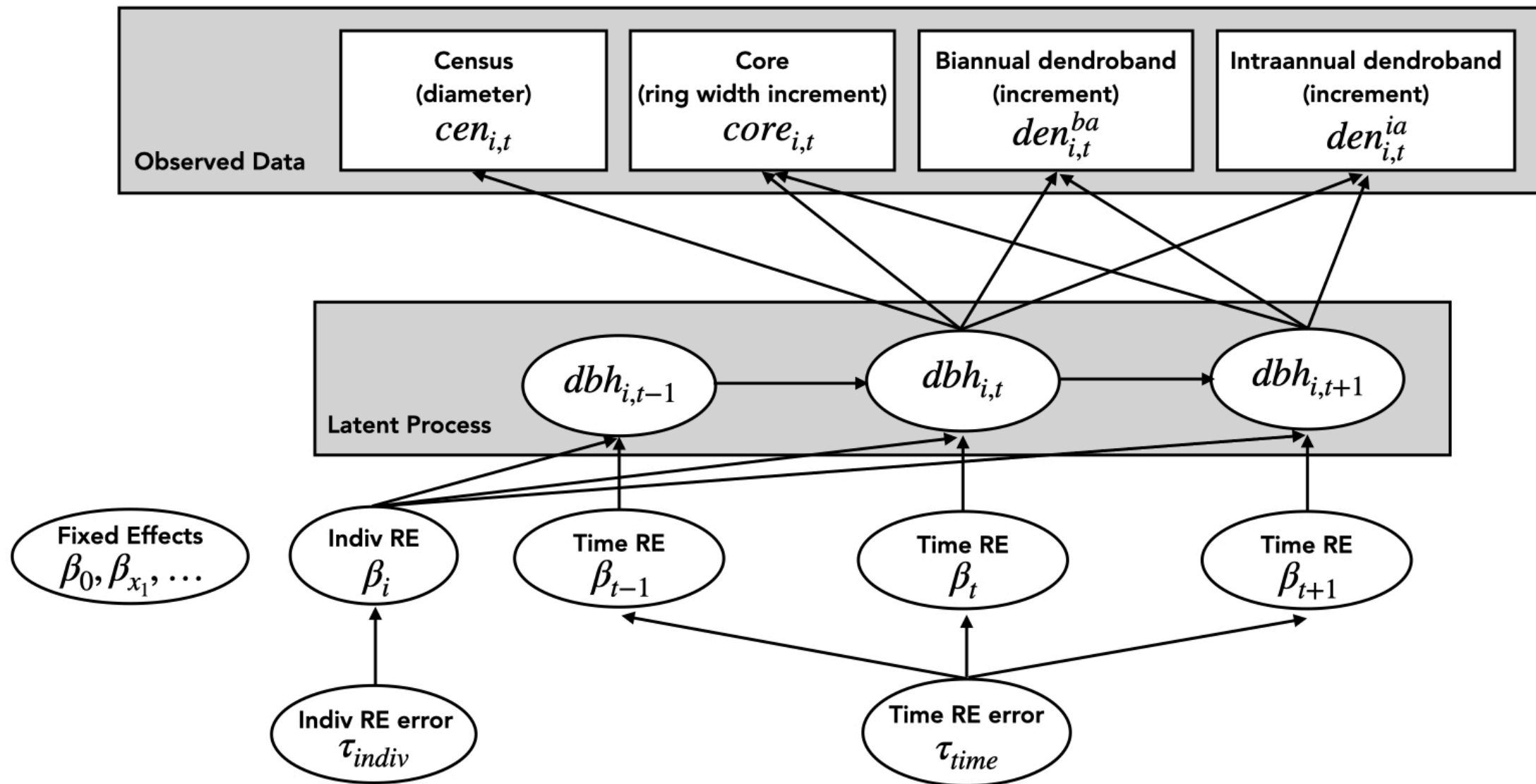
Fixed Effects
 $\beta_0, \beta_{x_1}, \dots$

Model

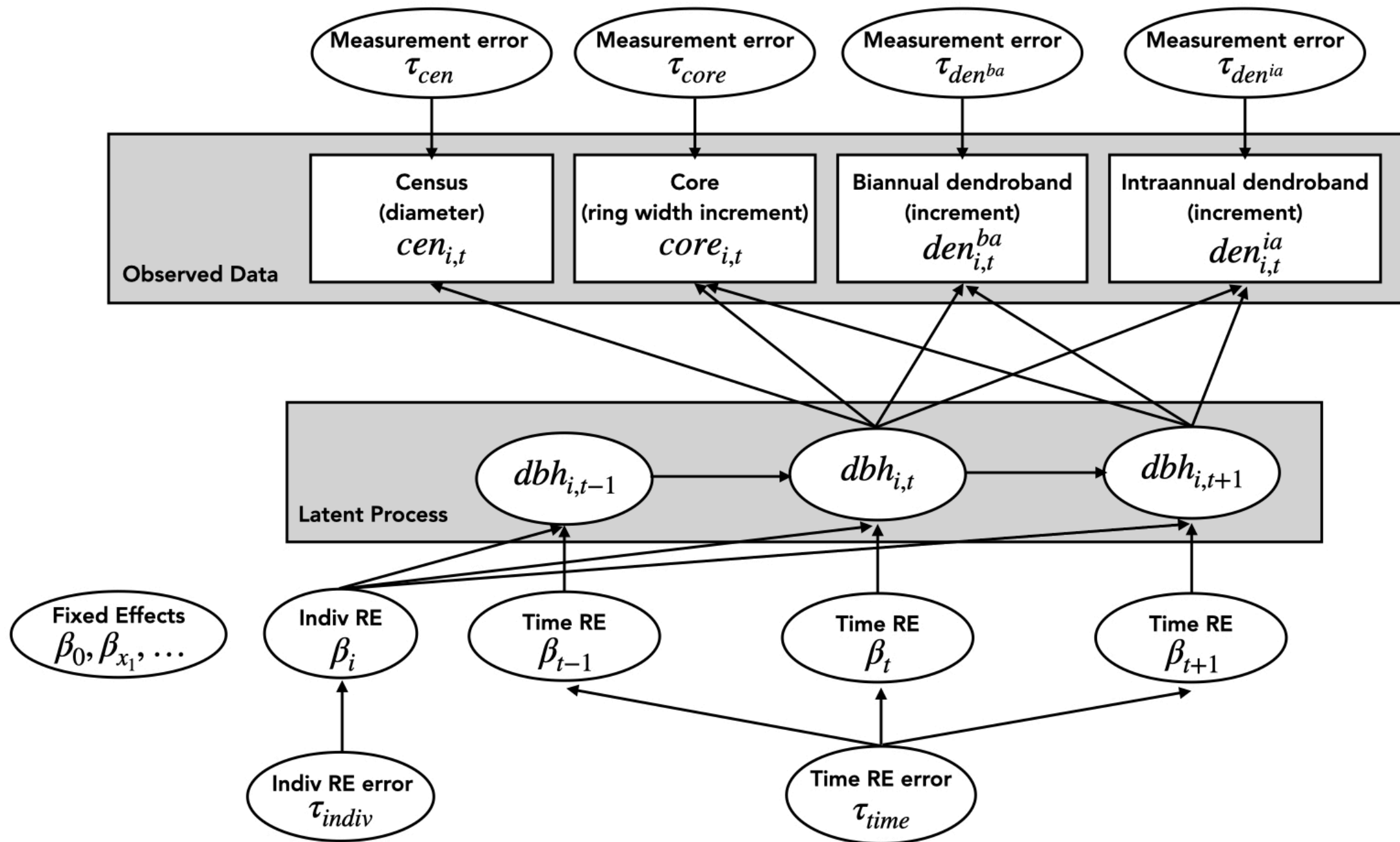


FYI: Express variances via precision $\tau = \frac{1}{\sigma^2}$

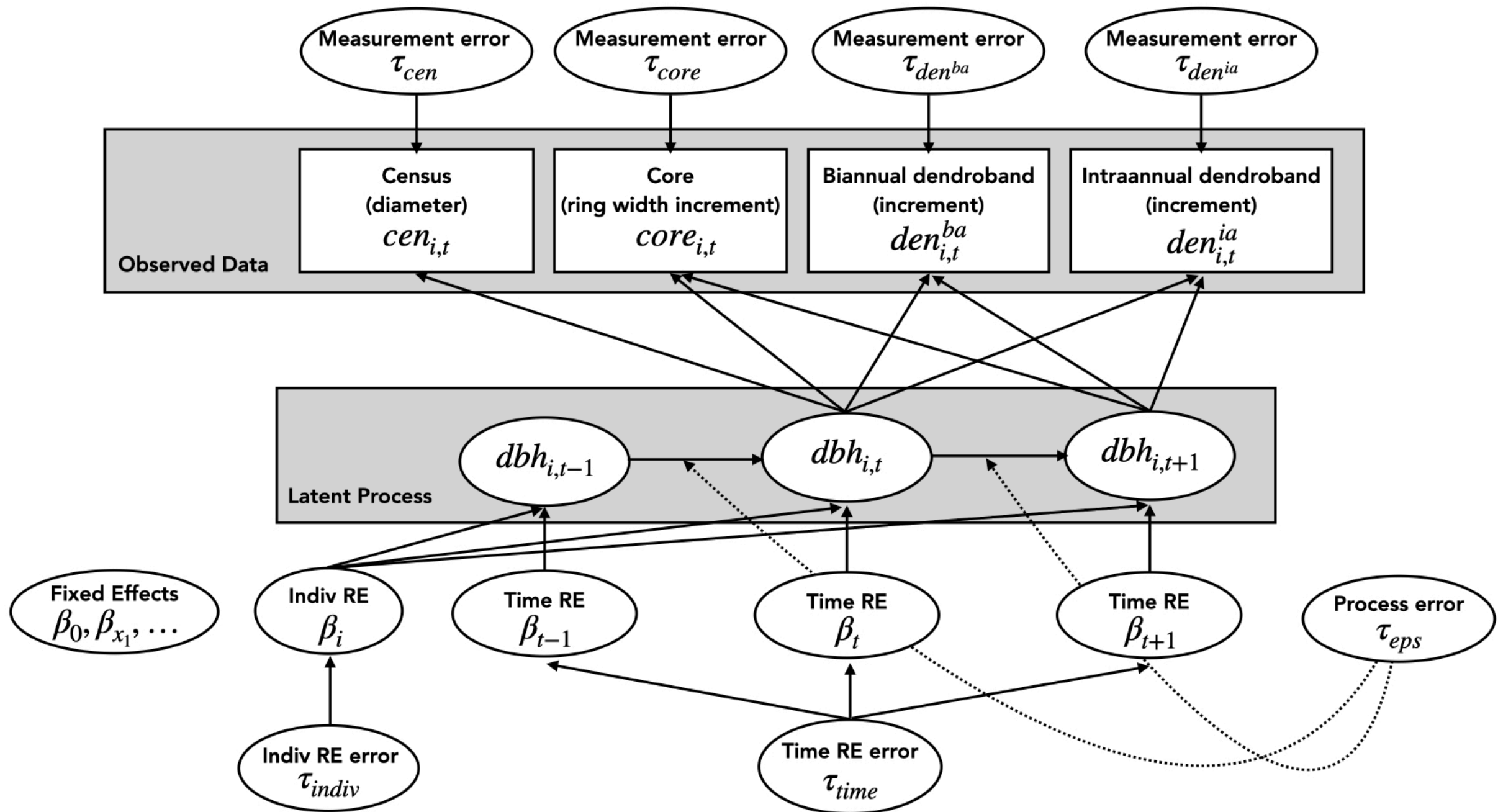
Model



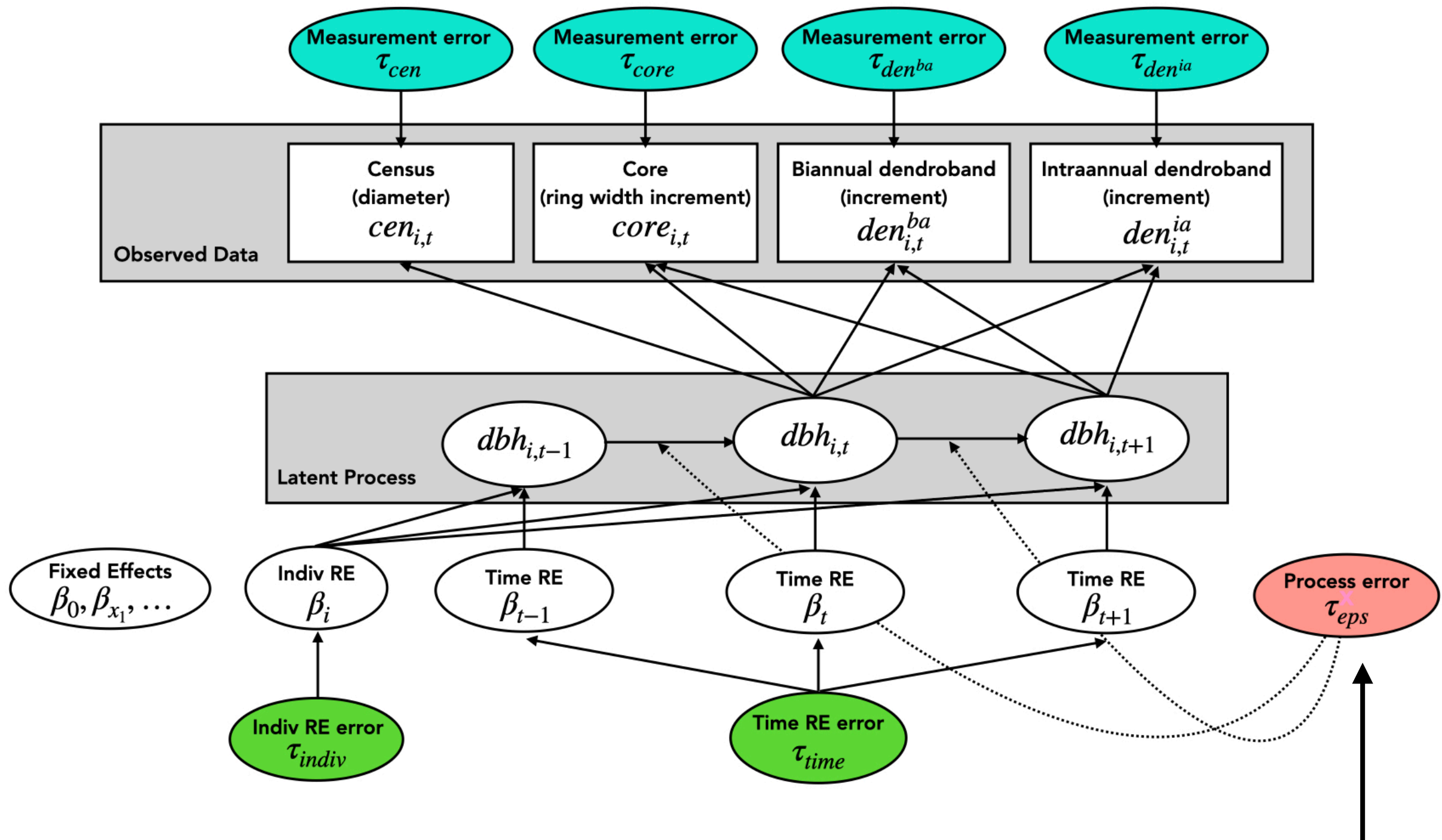
Model



Model

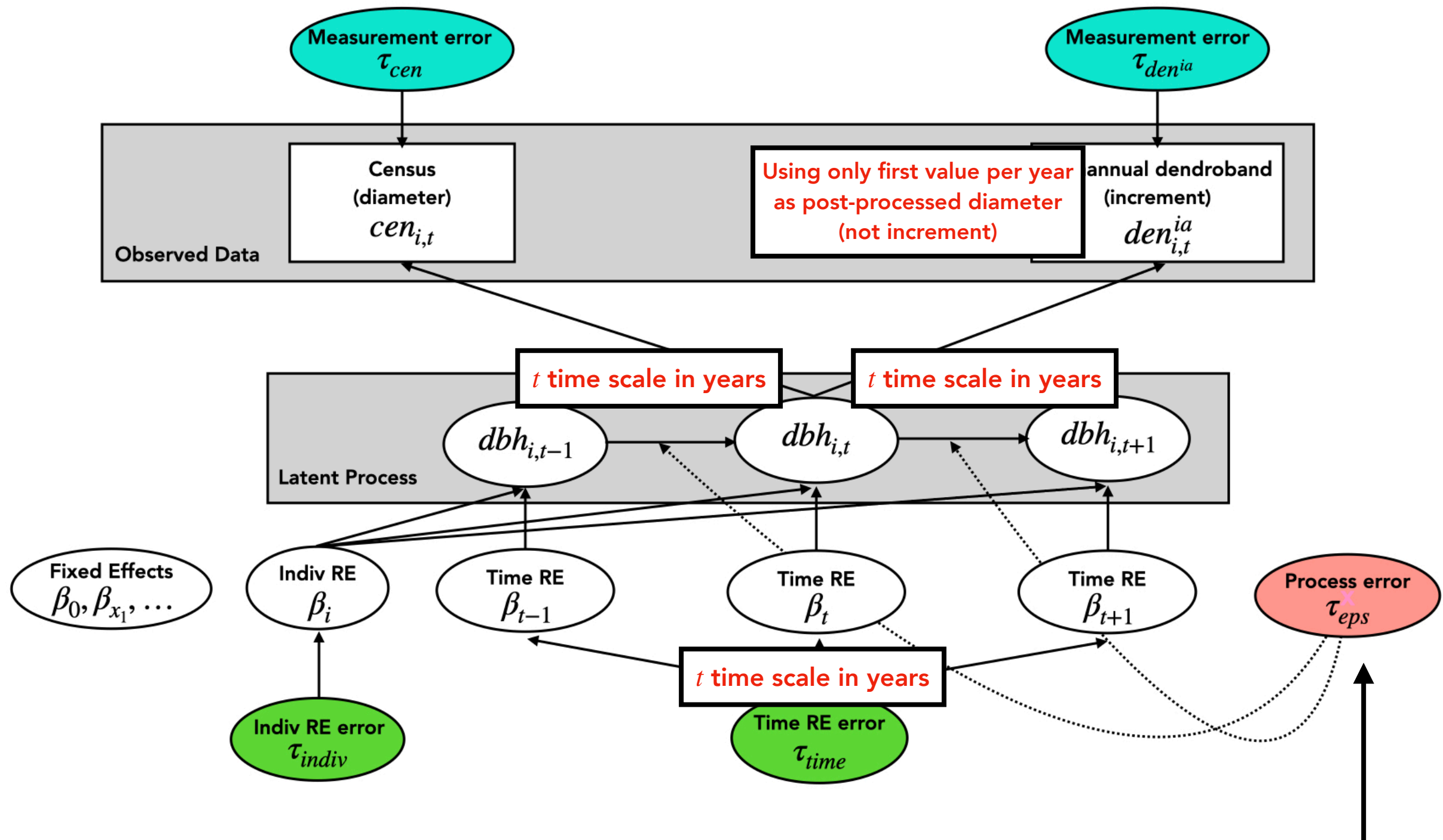


Model



Moral: only error this propagates across time in forecasts

Model as of 2021/1/22



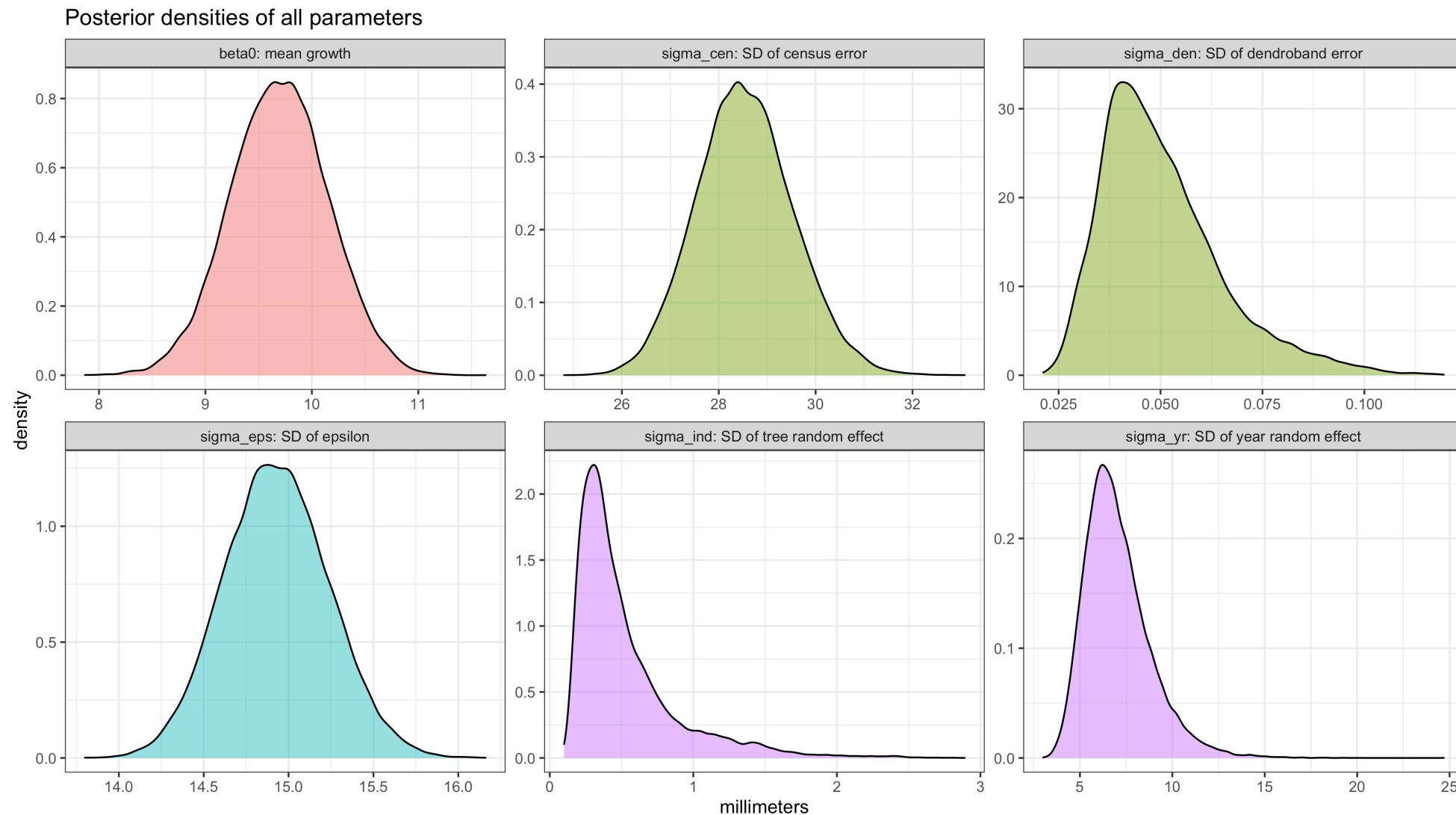
Moral: only error this propagates across time in forecasts

Results

MCMC specifications

- $n = 155$ trees with intraannual dendrobands (tend to be larger canopy trees)
- Implemented in JAGS
- 30k draws from posterior minus 10% burn-in
- Empirical Bayes (data informed) prior parameters
- Forecast into 2020 - 2022 by treating these years as missing values

Posterior Distributions



- mean $\beta_0 \sim 1\text{cm}$ growth per year (needs sanity checking)
- $\sigma_{cen} \gg \sigma_{den}$
- Year-to-year variation in growth > Between individuals variation
- σ_ϵ = remaining process error that propagates in forecasts across time

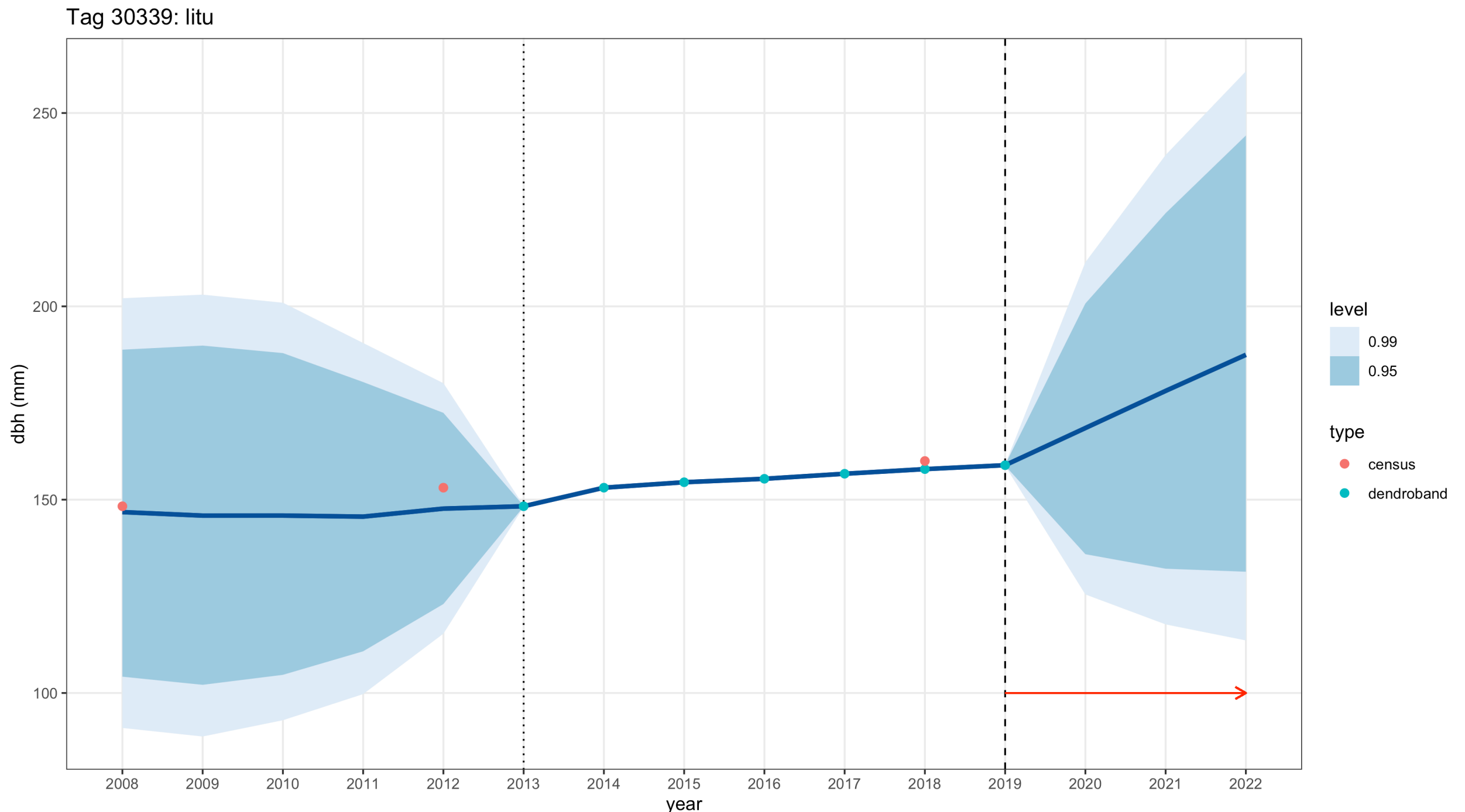
One particular tulip poplar

tag_stem	type	sp	`2007`	`2008`	`2009`	`2010`	`2011`	`2012`	`2013`	`2014`	`2015`	`2016`	`2017`	`2018`	`2019`
<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
30339_3	census	litu	NA	148.	NA	NA	NA	153.	NA	NA	NA	NA	NA	160	NA
30339_3	dendroband	litu	NA	NA	NA	NA	NA	NA	149.	155.	156.	157.	157.	159.	160.



One particular tulip poplar diameter

$y = \text{modeled true latent } dbh_{i,t}$



Dendroband installed in 2013

Future Work

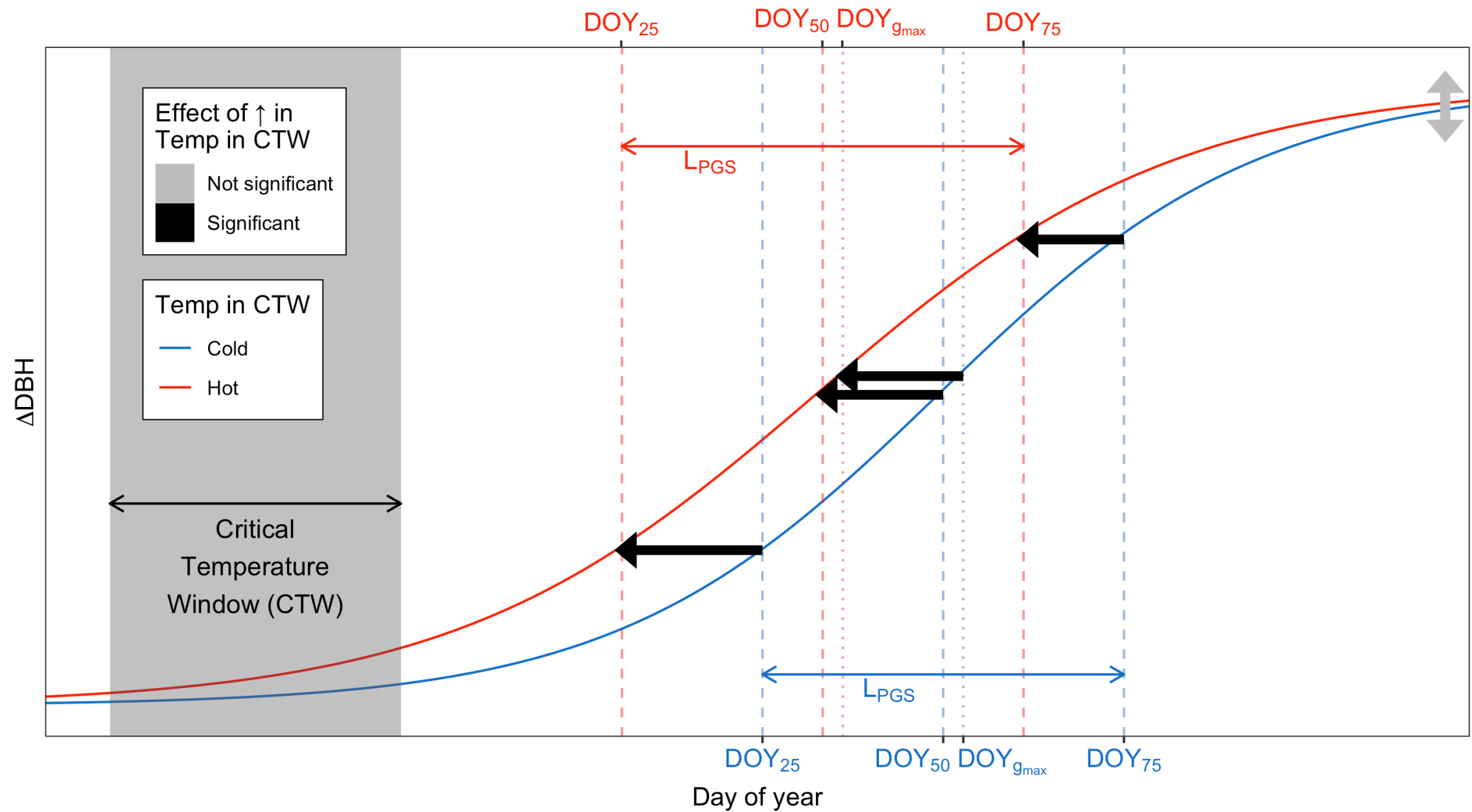
TODO

- Add remaining data:
 - All, not just first yearly observation, intra-annual dendroband
 - All biannual dendroband
 - Tree coring data
- Merge dendroband after comparing τ_{den}^{ba} versus τ_{den}^{ia} ?
- Improve model for latent variable
 - $dbh_{i,t} = dbh_{i,t-1} + \beta_0 + \dots + \epsilon$
 - Covariates: In particular species & starting diameter
- Choose appropriate time scale for t

Thanks!

**Slides on Twitter
@rudeboybert**

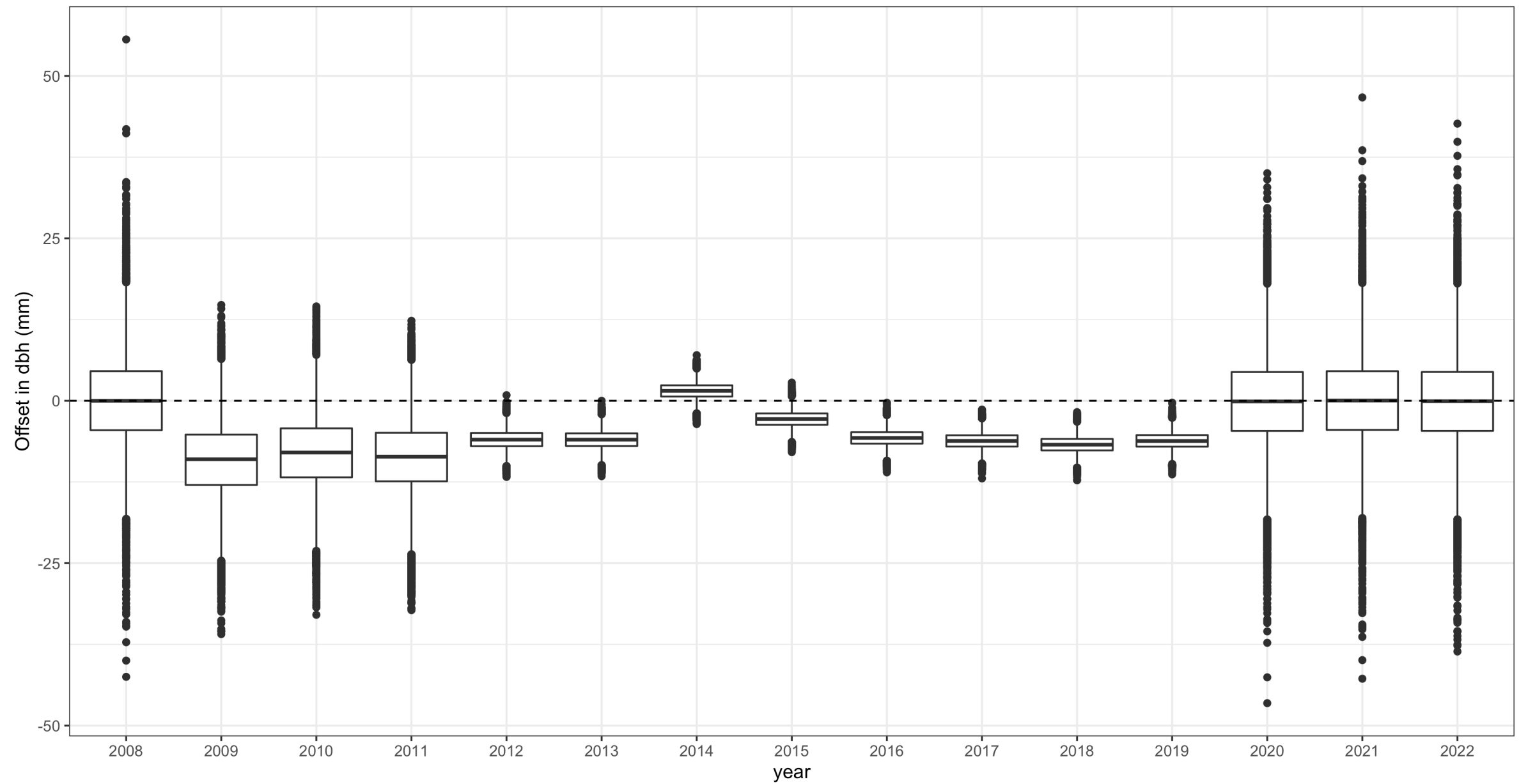
Intra-annual effect of climate



Year Random Effects

Year random effects

Distribution of all MCMC draws from posterior for each year



Individual Random Effects

Individual tree random effects

Distribution of all MCMC draws from posterior for each tree

