

Chapter 2

Preliminaries I: Life History Models

“You know,” said Arthur, “it’s at times like this, when I’m trapped in a Vogon airlock with a man from Betelgeuse, and about to die of asphyxiation in deep space that I really wish I’d listened to what my mother told me when I was young.”

— Douglas Adams, *The Hitchhiker’s Guide to the Galaxy*

All organisms grow and develop. Development from birth proceeds through stages of life, some of which are easily defined and recognizable, while others are less so. For example, the transition from pupa to adult in many insects involves a distinct transition and metamorphosis step, making it easy to identify both the pupa and the adult. In contrast, childhood in humans is more difficult to define and its recognized, official end varies from culture to culture. Demography concerns itself with documenting and explaining these transitions, particularly the rates at which they occur.

Demographic analysis generally requires a model of development for the organism being studied, because vital rates generally differ across developmental stages. We call such a model a **life history model**. At the very least, this model needs to include the major, biologically notable stages of life, where a stage is defined as a portion of the life cycle defined by a unique set of biological, developmental, or demographic characteristics. The stages that a life history model includes may therefore change given the circumstances of the research.

Developing a proper life history model may sound easy, but it is actually among the most difficult things to accomplish in demography. Indeed, although life history models underpin population ecology generally, and matrix projection model (MPM) analyses specifically, recent analyses suggest that the majority of demographic analyses of MPMs utilize incorrect life history models (Kendall et al., 2019). These errors lead to biases that can misinform biological inference, with profound consequences for conservation management, evolutionary prediction, or whatever the goal of the study happens to be.

In this chapter, we consider how to develop a reasonable life history model for use in matrix population modeling using `lefko3`. We begin with a discussion of what is required of a good life history model for MPM analysis, including some of the more common errors in model development. We will then move on to create some life history models for use in examples later in this book. Our discussion will focus on the most commonly utilized life history model, the **life cycle graph**.

2.1 The life cycle graph

Matrix projection models (MPMs) are representations of the dynamics of a population across all life history stages deemed relevant, across the most relevant time interval (typically one year, but sometimes different, and assumed to be consistent within each analysis). Each MPM requires a complete model of the organism’s life history prior to construction, and this model must explicitly show all life stages and all life history transitions that are biologically possible. Each stage is mutually exclusive

across time, meaning that an individual can only be in a single stage at a given time, and each transition takes exactly one full time step, which needs to be defined consistently and should be equivalent to the approximate amount of time between consecutive monitoring occasions. Each stage is represented in the matrix by a single column and a single row, and each transition by a single element. Matrix elements (a_{kj}) show either the probability of transition for an individual in stage j at occasion t (along the columns), to stage k at occasion $t + 1$ (along the rows), or the mean rate of production of new recruits into stage k at occasion $t + 1$ (along the rows) by individuals in reproductive stage j in occasion t (along the columns). Conceptually, each individual is in a particular stage in the instance of monitoring or observation, and then either dies or completes a transition in the interval between that occasion's observation and the next occasion's observation. Death is not an explicit life stage and so is not modeled as such, instead becoming a potential endpoint of each transition.

Here we utilize a **life cycle graph** approach to build a life history model. A life cycle graph shows the life history of an organism as a series of nodes and arrows, using the simplifying assumptions that time is discrete and stages represent the state of the individual at each discrete point in time. The **nodes**, typically shown as circles in the graph, represent unique stages that the individual may go through during development, and individuals are noted as occurring in these stages during monitoring sessions conducted over a roughly or strictly regular frequency, such as at the start of the growing season every year, or over specific dates each year. **Arrows** represent transitions from stage to stage between consecutive occasions. These may be interpreted either as *probabilities of survival-transition* when reflecting the individual's stage across time, or as *fecundity rates* when reflecting the production of offspring by individuals in reproductive stages. A good life cycle graph, and in general a good life history model, must explicitly include representations of all stages possible for the organism to enter, and all biologically plausible transitions between these stages. Figure 2.1 (a) shows an example of a life cycle graph.

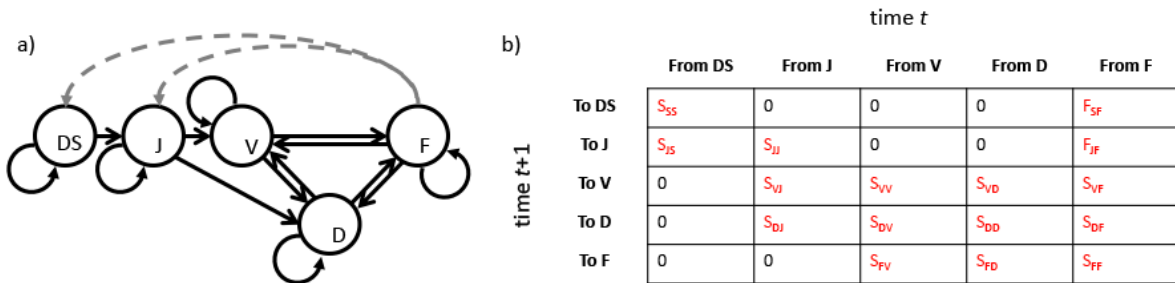


Figure 2.1: Simple life history model (a) and a historical MPM (b) for *Cypridium candidum*, a North American herbaceous plant species

Figure 2.1 is a simple example of a stage-based life history model viewed as a life cycle graph, and its associated Lefkovich matrix for a terrestrial orchid species, *Cypridium candidum* (Shefferson et al., 2001). Each stage is shown as a node, and each transition is shown as an uni-directional arrow (a). The rates and probabilities are shown as mathematical symbols in (b), with S_{kj} denoting survival-transition probability from stage j in occasion t to stage k in occasion $t + 1$, and F_{kj} denoting the fecundity of reproductive stage j in occasion t into recruit stage k in occasion $t + 1$ in this life history.

Stages are generally defined under the assumption that they are unique moments in an organism's life. This uniqueness is assumed to extend to the demography of the organism during its time within that stage, meaning that the stage must have a mutually exclusive set of life history characteristics. In some cases, stages are defined as developmental stages, as happens with insect instars. In other cases, stages are defined almost purely on the basis of size, as occurs with perennial herbs that may exhibit a different number of stems, leaves, and flowers in each growing season. In still other cases, stages may be defined via other characteristics, for example via age in a Leslie MPM. As another example, the vegetative dormancy stage of some perennial herbs is defined by a lack of aboveground size, and so is

characteristically not observable and is defined by its lack of observability (Shefferson et al., 2001).

2.1.1 Stage duration

Stages always have durations, and this duration can have important influences on the life history model developed. Particularly, a life history model used for MPM analysis needs to employ a consistent time interval for transitions, and so stages can only be defined if they occur at roughly the time resolution of the life history model. To think about stage duration, users may wish to consider a simple model that relates the life history model to the timing of monitoring. In this model, individuals exist in some stage at the instance of monitoring. Monitoring happens at regular intervals, and stages are defined in ways so that their durations are at least the interval between consecutive monitoring occasions. Stage transitions can then be thought of conceptually as occurring either right before the monitoring occasion (the **pre-breeding** model), or immediately after (the **post-breeding** model). To simplify analysis, we also assume that death happens at the time of transition.

Erroneous accounting of stage duration is commonplace in MPM analyses. The most common problem is that one or more stages within a life history model are actually characterized by different durations than other stages. Often, a stage's actual duration is longer than acknowledged. As an example, we might wish to create a seedling stage that yields a small adult. We might initially create a simple three-stage life history in which a seed stage is followed by the seedling, and this is followed by a small adult stage, which may be capable of reproduction. Imagine, however, that the seedling stage is almost never shorter than two years, and that we have a dataset based on an annual census (i.e. our MPM will consist of annual matrices). In this circumstance, the seedling stage must be split into at least two seedling stages in order to force the analysis to include the right minimum number of years to keep the organism a seedling. Failure to do so would create a model that assumes a shorter juvenile period than actually occurs, and would likely lead to biased estimates of population dynamics parameters, such as the asymptotic population growth rate.

Stage duration mismatch may also occur when actual stage durations are shorter than assumed by the model. For example, if I wish to study an insect that lives potentially up to ten years, and I will employ a matrix projection model in which matrices represent one year of transition, then I cannot divide the insect instars into multiple stages if they only last several days or weeks each in succession. Instead, I would need to lump those stages together in a way that incorporates a realistic amount of instar development over the course of a single year, and use that conglomerate stage within the life history model. Alternatively, I can deal with this issue by changing the set time interval used in the MPM, making a model with a much finer time resolution (although this would require monitoring many more times than just once a year, and might require redefining other stages).

2.1.2 The process and timing of fecundity

The timing of monitoring relative to the reproductive season greatly impacts the structure of life history models. Life history models are typically categorized as either **pre-breeding** or **post-breeding**. Here, *breeding* refers to the production of offspring, and so a pre-breeding model assumes that monitoring is conducted immediately before the new recruits are born, while a post-breeding model assumes that the monitoring is conducted just after new recruits are born (in both cases, breeding is assumed to be seasonal). In a pre-breeding model, no organism is monitored prior to an age of approximately one year, or one of whatever time unit is being used in the study. Further, in pre-breeding models fecundity equals the production of newborns multiplied by the survival of newborns to age 1, since fecundity must take place over a full time step and the timing of the census misses both the birth event itself and the time in which the organism is a newborn (Kendall et al., 2019). In a post-breeding model, organisms are first monitored at age 0, and fecundity equals the survival of the parent from the stage/age preceding reproduction to reproduction itself, multiplied by the number of newborns produced (Kendall et al., 2019).

Population ecologists often mistakenly forget important components of fecundity because of the inherent complexity of life cycle models. Often, ecologists will estimate fecundity in pre-breeding studies erroneously as only the production of newborns, without multiplying by the survival of newborns to age 1 (Kendall et al., 2019). Ecologists working with post-breeding models often erroneously omit the survival of the parent from the stage immediately preceding reproduction, since including this term generally requires creating reproductive versions of all possible preceding stages even if, counterintuitively, they are not considered reproductive stages (Kendall et al., 2019). Both of these omissions result in biased analytical results, of which the most common problem is that population growth rate estimates that are too high.

A comparison of the same system developed in pre-breeding vs. post-breeding format might help illustrate these issues. Figure 2.2 shows the same model as in figure 2.1, but given in both pre-breeding and post-breeding format. The flowering stage is the only reproductive stage in the pre-breeding model, and fecundity there is the expected number of seeds produced multiplied by either the probability of survival as a dormant seed (in production of dormant seed), or the probability of germination (in the production of juveniles). In the post-breeding model, the probability of survival from the stage preceding flowering is multiplied by the expected number of seeds, leading to a single offspring stage instead of two, and three forms of the flowering stage to deal with different expected fecundities in each case.

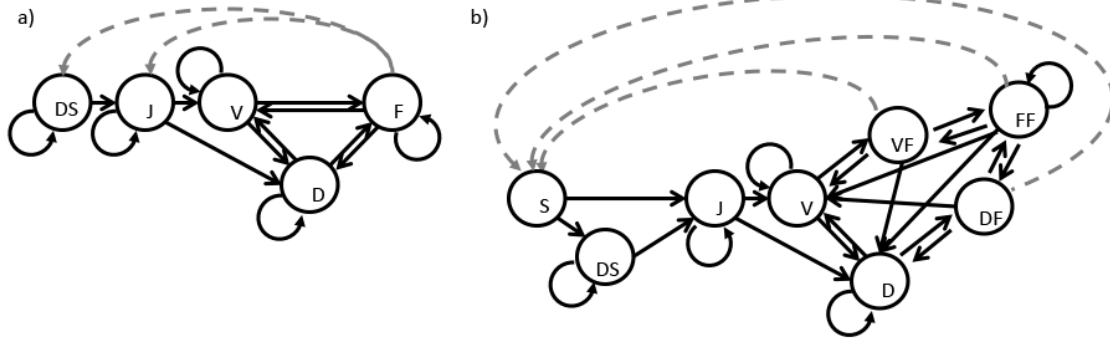


Figure 2.2: Simple life history pre-breeding model (a) and post-breeding model (b) for *Cypripedium candidum*, a North American herbaceous plant species. Stages include: S, seed; DS, dormant seed; J, juvenile (including protocorm and seedling); D, vegetatively dormant adult; V, sprouting but not flowering adult; F, flowering adult

In monitoring studies of plant populations, the typical strategy taken is that of a pre-breeding model, in which fecundity is estimated as the production of seeds in a given year multiplied by their over-winter survival probability as seeds and their germination probability in the following year. If the life history includes seed dormancy, then it can be modeled by multiplying seed production by seed survival (which is given as the probability of maintaining seed viability from one year to the next). In more complex situations, different seed survival probabilities may be set depending on whether the seed is freshly produced or dormant within the seedbank. Added complexity can arise if there are multiple fecundity pathways, for example when clonal reproduction is also possible, or if multiple propagule stages exist, or if recruitment can yield seeds that immediately germinate as well as seeds that are dormant for one or more years. We urge users to be careful with this step, as properly defining the life history model has important implications for all analyses of population dynamics (Kendall et al.,

2019).

2.1.3 Dataset and analytical influences on life history model design

Life history models can be influenced by factors seemingly external to the life history of the organism. Some of the most important considerations outside of the species' biology include the size of the dataset, and the decision of whether to create a raw MPM or a function-based MPM. All else being equal, a larger dataset and a function-based MPM will allow more stages to be used in the life history. Assuming a dataset of fixed size, raw MPMs will include more and more zeros as the number of stages in the life history model increases. This happens because the chance of having any individuals actually move through a specific transition decreases as the number of possible transitions increases. These increasingly common zeros can cause problems in analysis, because they suggest that some transitions are impossible when, in fact, individuals moving through them were missing just by chance. Even a transition with extremely high probability may be missing in some years just by chance. One impact of this would be the reduction of the mean transition within an element-wise mean matrix to an unrealistic level, causing both the survival probability of the associated stage and the population growth rate to be estimated too low.

Function-based MPMs are better able to handle large numbers of matrix elements because overparameterization is prevented by parsimonious model selection when vital rate models are determined. However, the size of the dataset influences the statistical power of these vital rate models, with smaller datasets yielding larger error associated with matrix elements. Loss of statistical power in vital rate models can lead to the loss of process variance, and can make population dynamics appear more static than they really are. In addition, adding more stages than necessary in the function-based case can create conditions in which stages themselves are estimated as having much higher survival probability than they actually do (this happens when stages that do not occur are modeled, as they are typically estimated as having trivial but non-zero probabilities of survival). The survival of a stage within a matrix is the column sum of for that stage in a survival-transition (i.e. a projection matrix without any fecundity terms). Vital rate functions generally produce non-zero values no matter what input is provided, even if sizes and conditions are entered that do not occur in the dataset, or that occur only rarely. Such situations may result in the artificial inflation of survival, and in extreme cases can result in predicted survival probability above 1.0. So, even in the function-based case, great effort needs to be taken to prevent the inclusion of too many stages, and of stages that do not actually occur in the dataset.

Chapter 6 in Morris and Doak (2002) provides a good description of techniques that may be used to define stages and to determine the exact number to use given the context imposed by a given dataset of a given size. Ellner and Rees (2006) provides some discussion of this issue within the context of integral projection model development.

Now let's move on to create a life history model and define it in R using `lefk03`.

2.2 Life history model development in `lefk03`

The basic workflow to analysis with package `lefk03` starts with the development of a life history model that encapsulates all of the appropriate life stages relevant to population dynamics. We will illustrate life history model development with the *Cypripedium candidum* dataset described in section 1.6.1. This plant begins life by germinating from a dust seed, and then develops into a protocorm, which is a special subterranean life stage found in orchids and pyroloids. During this stage, the plant is non-photosynthetic and completely parasitic on its mycorrhizal fungi. It spends several years as a protocorm, and previous studies suggest that it typically spends three years before becoming a seedling. As a seedling, it may or may not produce aboveground sprouts, often remaining entirely subterranean and continuing its parasitic lifestyle. It may persist this way for many years before attaining adult size, at which point it may sprout with or without flowers, or may remain underground in a condition

referred to as **vegetative dormancy**. The latter condition may continue for many years, with over a decade of continuous dormancy documented in the literature (Shefferson et al., 2018).

We will now load `lefko3` and the dataset, called `cypdata`, after clearing memory.

```
rm(list=ls(all=TRUE))
library(lefko3)

data(cypdata)
```

The first key decision in developing a **life cycle graph** is to decide on which life history stages to include. This is not a simple task. We might at first believe that any clearly defined developmental stage should be its own life history stage. However, life history stages need to be defined relative to the time interval used in the MPM. They also need to be defined uniquely, and to have transitions from other stages, leading either to further life history stages or back to themselves. Finally, they also need to be defined with consideration given to the size of the dataset, since smaller datasets will allow fewer stages.

It might be easiest to begin our discussion with consideration of the function-based MPM, and what sort of life history model works best if we choose this approach. With a function-based MPM (which includes the integral projection model, or IPM), there are really only two considerations that we need to be concerned about. First, what stages should be included based on a purely biological understanding of the organism's life history, and our decision on the time interval? Second, what is the range of stages and/or sizes that actually occur in the dataset and are typically encountered in nature? Let's discuss these points in turn.

The first point - taking the biology of the organism into account - seems intuitive, but it sometimes breaks down when we consider the time interval of the study. Key issues have already been mentioned - the timing of reproduction relative to the monitoring occasion means that fecundity itself must include either newborn survival or the survival of new parents from the preceding stage, but not both. This means that the same life history might lead to a single reproductive stage if a pre-breeding model is used, or several if a post-breeding model is used (since reproduction would happen after each possible preceding stage). If newborns are propagules, such as seeds, then the same choices determine whether the first stage in the life cycle graph is a recruited juvenile (as in the pre-breeding case), or a propagule (as in the post-breeding case). Users should take great care with this step.

The second point - identifying stages that actually occur in the dataset and that should occur typically in nature - is also important. Some stages may never be monitored, and so these will have to be included in the model but dealt with via proxy rates later. In our *Cypripedium* case, for example, we never monitored germinated seeds, protocorms, and seedlings, because it is essentially impossible to do so. We will include these stages in our life history model, but we will use proxy rates for their survival transitions, since we cannot use our dataset to estimate them. Conversely, there may be some stages that appear to be outliers, appearing only once and representing unusually large sizes or other unusual conditions. Such stages should not be included, because their inclusion may yield odd impacts on analysis, such as the artificial inflation of stage survival.

This question is really focused on determining which survival and fecundity transitions we have the ability to estimate given our dataset. We should develop our life cycle graph with a reasonable range of sizes and stages in mind, since we do not wish to create stages in our life cycle model that go beyond the limits of what was actually observed (models that make predictions outside of the range of data they were parameterized with have a tendency to produce erroneous and at times egregiously strange predictions). So let's explore the range of sizes actually occurring in our dataset. There is no one variable that encapsulates the entire size of the individual in our dataset, so we will create a series of vectors that sums the numbers of sprouts that are single-flowered flowering, double-flowered flowering, and non-flowering, and use these sums as plant sizes. If we use the horizontal dataset for this purpose, then we can assess size as follows.

```

size.04 <- cypdata$Inf2.04 + cypdata$Inf.04 + cypdata$Veg.04
size.05 <- cypdata$Inf2.05 + cypdata$Inf.05 + cypdata$Veg.05
size.06 <- cypdata$Inf2.06 + cypdata$Inf.06 + cypdata$Veg.06
size.07 <- cypdata$Inf2.07 + cypdata$Inf.07 + cypdata$Veg.07
size.08 <- cypdata$Inf2.08 + cypdata$Inf.08 + cypdata$Veg.08
size.09 <- cypdata$Inf2.09 + cypdata$Inf.09 + cypdata$Veg.09

summary(c(size.04, size.05, size.06, size.07, size.08, size.09))
>   Min. 1st Qu.  Median    Mean 3rd Qu.   Max.   NA's
>  1.000  1.000   2.000   3.644  5.000  24.000   156

```

The summary shows that the smallest recorded size is a single sprout, and the largest is 24 sprouts. The 97 NAs are a combination of vegetative dormancy and instances of individuals being dead or simply not yet born. Given this, we might utilize the following life history model for our function-based MPMs (figure 2.3).

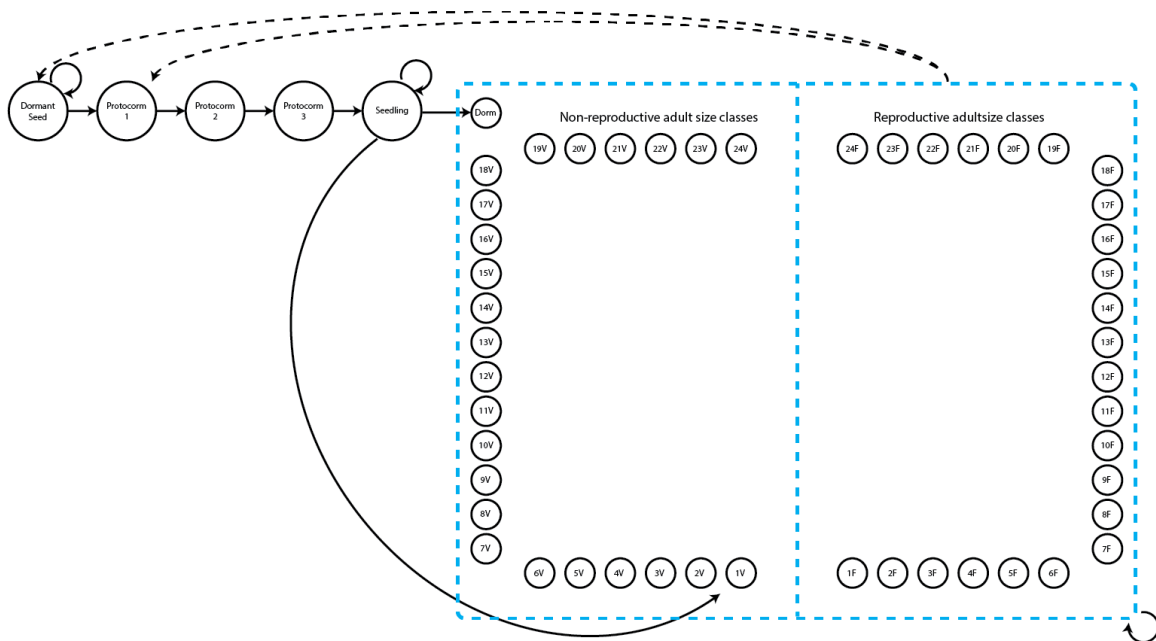


Figure 2.3: Life history model of *Cypripedium candidum* for use in function-based MPMs

We can see a variety of transitions within this figure. The propagule and juvenile stages are the five stages in the top-left corner and have fairly simple transitions. New recruits may enter the population directly from the germination of a seed produced the previous year (**Protocorm 1** stage), or they may begin as dormant seed (**Dormant seed** stage). Dormant seed may remain dormant, die, or germinate into the **Protocorm 1** stage. Protocorms exist for up to three years without transitioning further, and individuals cannot have fewer than three years as protocorms, yielding the **Protocorm 1**, **Protocorm 2**, and **Protocorm 3** stages. **Protocorm 3** leads to the **Seedling** stage, in which the plant may persist for at least one year and potentially many years before becoming mature. Here, maturity does not really refer to reproduction *per se*, but rather to a morphology indistinguishable from a reproductive plant except for the lack of a flower. The first mature stage is usually either vegetative dormancy (**Dorm**), during which time the plant does not sprout, or a small, non-flowering adult (**1V**). Once in this mature portion of the life history (designated in the hashed blue box), the

plant may transition among 49 mature stages, including vegetative dormancy, 1-24 shoots without flowers, or 1-24 shoots with at least one flower.

2.2.1 Building the stageframe

Now that we have our life history model, we will need to describe the life history characterizing the dataset, matching it to our analyses properly with a `stageframe` for our *Cypridium candidum* dataset. This is a vitally important step, and most instances of errors occurring in the use of `lefko3` originate from an inappropriate `stageframe` being used in an analysis. Since this analysis will be function-based, we will include all possible size classes here. If constructing raw matrices, all sizes that occur in the dataset need to be accounted for in a way that is both natural and parsimonious with respect to the numbers of individuals moving through actual transitions. If constructing function-based matrices, such as IPMs, then representative sizes at systematic increments will be satisfactory. Since size is count-based in the *Cypridium candidum* case, we will use all numbers of stems that might occur from zero to the maximum in the dataset, representing the life history diagram shown in the beginning of this chapter.

To start, we will create two vectors, one showing the names of all stages in our life history model, and one showing whether the stage is included in the dataset. Both vectors utilize the same order of stages.

```
stagevector <- c("SD", "P1", "P2", "P3", "SL", "D", "V1", "V2", "V3", "V4",
  "V5", "V6", "V7", "V8", "V9", "V10", "V11", "V12", "V13", "V14", "V15", "V16",
  "V17", "V18", "V19", "V20", "V21", "V22", "V23", "V24", "F1", "F2", "F3",
  "F4", "F5", "F6", "F7", "F8", "F9", "F10", "F11", "F12", "F13", "F14", "F15",
  "F16", "F17", "F18", "F19", "F20", "F21", "F22", "F23", "F24")
indataset <- c(0, 0, 0, 0, 0, rep(1, 49))
```

We included all stages in this step, and the `indataset` vector allows us to tell R that the first five stages ("SD", "P1", "P2", "P3", "SL") do not actually exist in the dataset. Next, we will create a few more vectors to describe the life history fully. The vectors will include the representative size (typically the mean or median size), reproductive status, observation status (i.e. whether it is technically possible to observe the stage, or whether the stage is essentially invisible and must be inferred to be present), and maturity status of each of these stages.

```
sizevector <- c(0, 0, 0, 0, 0, seq(from = 0, t = 24), seq(from = 1, to = 24))
repvector <- c(0, 0, 0, 0, 0, rep(0, 25), rep(1, 24))
obsvector <- c(0, 0, 0, 0, 0, 0, rep(1, 48))
matvector <- c(0, 0, 0, 0, 0, rep(1, 49))
```

Now we will create just three more vectors. The first will be a vector for status as an immature stage. While this vector might seem to be simply the logical inverse and so the mathematical complement to `matvector`, in truth there are cases where a stage may be both mature and immature, depending on the circumstance, and cases where a stage may actually be neither. One stage that is neither is **SD**, the dormant seed stage - that is marked as 0 in both cases. The second vector is a vector showing status as a propagule stage. The final vector is a comment vector, providing text descriptions of the stages.

```
immvector <- c(0, 1, 1, 1, 1, rep(0, 49))
propvector <- c(1, rep(0, 53))
comments <- c("Dormant seed", "Yr1 protocorm", "Yr2 protocorm", "Yr3 protocorm",
  "Seedling", "Veg dorm", "Veg adult 1 stem", "Veg adult 2 stems",
  "Veg adult 3 stems", "Veg adult 4 stems", "Veg adult 5 stems",
```



```
"Veg adult 6 stems", "Veg adult 7 stems", "Veg adult 8 stems",
"Veg adult 9 stems", "Veg adult 10 stems", "Veg adult 11 stems",
"Veg adult 12 stems", "Veg adult 13 stems", "Veg adult 14 stems",
"Veg adult 15 stems", "Veg adult 16 stems", "Veg adult 17 stems",
"Veg adult 18 stems", "Veg adult 19 stems", "Veg adult 20 stems",
"Veg adult 21 stems", "Veg adult 22 stems", "Veg adult 23 stems",
"Veg adult 24 stems", "Flo adult 1 stem", "Flo adult 2 stems",
"Flo adult 3 stems", "Flo adult 4 stems", "Flo adult 5 stems",
"Flo adult 6 stems", "Flo adult 7 stems", "Flo adult 8 stems",
"Flo adult 9 stems", "Flo adult 10 stems", "Flo adult 11 stems",
"Flo adult 12 stems", "Flo adult 13 stems", "Flo adult 14 stems",
"Flo adult 15 stems", "Flo adult 16 stems", "Flo adult 17 stems",
"Flo adult 18 stems", "Flo adult 19 stems", "Flo adult 20 stems",
"Flo adult 21 stems", "Flo adult 22 stems", "Flo adult 23 stems",
"Flo adult 24 stems")
```

Next we will create the stageframe using the `sf_create()` function.

```
cypframe_fb <- sf_create(sizes = sizevector, stagenames = stagevector,
  repstatus = repvector, obsstatus = obsvector, matstatus = matvector,
  propstatus = propvector, immstatus = immvector, indataset = indataset,
  comments = comments)
```

```
cypframe_fb
>   stage size size_b size_c min_age max_age repstatus obsstatus propstatus
> 1   SD    0    NA    NA      NA      NA          0          0          1
> 2   P1    0    NA    NA      NA      NA          0          0          0
> 3   P2    0    NA    NA      NA      NA          0          0          0
> 4   P3    0    NA    NA      NA      NA          0          0          0
> 5   SL    0    NA    NA      NA      NA          0          0          0
> 6   D     0    NA    NA      NA      NA          0          0          0
> 7   V1    1    NA    NA      NA      NA          0          1          0
> 8   V2    2    NA    NA      NA      NA          0          1          0
> 9   V3    3    NA    NA      NA      NA          0          1          0
> 10  V4    4    NA    NA      NA      NA          0          1          0
> 11  V5    5    NA    NA      NA      NA          0          1          0
> 12  V6    6    NA    NA      NA      NA          0          1          0
> 13  V7    7    NA    NA      NA      NA          0          1          0
> 14  V8    8    NA    NA      NA      NA          0          1          0
> 15  V9    9    NA    NA      NA      NA          0          1          0
> 16  V10   10   NA    NA      NA      NA          0          1          0
> 17  V11   11   NA    NA      NA      NA          0          1          0
> 18  V12   12   NA    NA      NA      NA          0          1          0
> 19  V13   13   NA    NA      NA      NA          0          1          0
> 20  V14   14   NA    NA      NA      NA          0          1          0
> 21  V15   15   NA    NA      NA      NA          0          1          0
> 22  V16   16   NA    NA      NA      NA          0          1          0
> 23  V17   17   NA    NA      NA      NA          0          1          0
> 24  V18   18   NA    NA      NA      NA          0          1          0
> 25  V19   19   NA    NA      NA      NA          0          1          0
> 26  V20   20   NA    NA      NA      NA          0          1          0
> 27  V21   21   NA    NA      NA      NA          0          1          0
> 28  V22   22   NA    NA      NA      NA          0          1          0
```

```

> 29 V23 23 NA NA NA NA 0 1 0
> 30 V24 24 NA NA NA NA 0 1 0
> 31 F1 1 NA NA NA NA 1 1 0
> 32 F2 2 NA NA NA NA 1 1 0
> 33 F3 3 NA NA NA NA 1 1 0
> 34 F4 4 NA NA NA NA 1 1 0
> 35 F5 5 NA NA NA NA 1 1 0
> 36 F6 6 NA NA NA NA 1 1 0
> 37 F7 7 NA NA NA NA 1 1 0
> 38 F8 8 NA NA NA NA 1 1 0
> 39 F9 9 NA NA NA NA 1 1 0
> 40 F10 10 NA NA NA NA 1 1 0
> 41 F11 11 NA NA NA NA 1 1 0
> 42 F12 12 NA NA NA NA 1 1 0
> 43 F13 13 NA NA NA NA 1 1 0
> 44 F14 14 NA NA NA NA 1 1 0
> 45 F15 15 NA NA NA NA 1 1 0
> 46 F16 16 NA NA NA NA 1 1 0
> 47 F17 17 NA NA NA NA 1 1 0
> 48 F18 18 NA NA NA NA 1 1 0
> 49 F19 19 NA NA NA NA 1 1 0
> 50 F20 20 NA NA NA NA 1 1 0
> 51 F21 21 NA NA NA NA 1 1 0
> 52 F22 22 NA NA NA NA 1 1 0
> 53 F23 23 NA NA NA NA 1 1 0
> 54 F24 24 NA NA NA NA 1 1 0
> immstatus matstatus indataset binhalfwidth_raw sizebin_min sizebin_max
> 1 0 0 0 0.5 -0.5 0.5
> 2 1 0 0 0.5 -0.5 0.5
> 3 1 0 0 0.5 -0.5 0.5
> 4 1 0 0 0.5 -0.5 0.5
> 5 1 0 0 0.5 -0.5 0.5
> 6 0 1 1 0.5 -0.5 0.5
> 7 0 1 1 0.5 0.5 1.5
> 8 0 1 1 0.5 1.5 2.5
> 9 0 1 1 0.5 2.5 3.5
> 10 0 1 1 0.5 3.5 4.5
> 11 0 1 1 0.5 4.5 5.5
> 12 0 1 1 0.5 5.5 6.5
> 13 0 1 1 0.5 6.5 7.5
> 14 0 1 1 0.5 7.5 8.5
> 15 0 1 1 0.5 8.5 9.5
> 16 0 1 1 0.5 9.5 10.5
> 17 0 1 1 0.5 10.5 11.5
> 18 0 1 1 0.5 11.5 12.5
> 19 0 1 1 0.5 12.5 13.5
> 20 0 1 1 0.5 13.5 14.5
> 21 0 1 1 0.5 14.5 15.5
> 22 0 1 1 0.5 15.5 16.5
> 23 0 1 1 0.5 16.5 17.5
> 24 0 1 1 0.5 17.5 18.5

```

> 25	0	1	1	0.5	18.5	19.5
> 26	0	1	1	0.5	19.5	20.5
> 27	0	1	1	0.5	20.5	21.5
> 28	0	1	1	0.5	21.5	22.5
> 29	0	1	1	0.5	22.5	23.5
> 30	0	1	1	0.5	23.5	24.5
> 31	0	1	1	0.5	0.5	1.5
> 32	0	1	1	0.5	1.5	2.5
> 33	0	1	1	0.5	2.5	3.5
> 34	0	1	1	0.5	3.5	4.5
> 35	0	1	1	0.5	4.5	5.5
> 36	0	1	1	0.5	5.5	6.5
> 37	0	1	1	0.5	6.5	7.5
> 38	0	1	1	0.5	7.5	8.5
> 39	0	1	1	0.5	8.5	9.5
> 40	0	1	1	0.5	9.5	10.5
> 41	0	1	1	0.5	10.5	11.5
> 42	0	1	1	0.5	11.5	12.5
> 43	0	1	1	0.5	12.5	13.5
> 44	0	1	1	0.5	13.5	14.5
> 45	0	1	1	0.5	14.5	15.5
> 46	0	1	1	0.5	15.5	16.5
> 47	0	1	1	0.5	16.5	17.5
> 48	0	1	1	0.5	17.5	18.5
> 49	0	1	1	0.5	18.5	19.5
> 50	0	1	1	0.5	19.5	20.5
> 51	0	1	1	0.5	20.5	21.5
> 52	0	1	1	0.5	21.5	22.5
> 53	0	1	1	0.5	22.5	23.5
> 54	0	1	1	0.5	23.5	24.5
>	sizebin_center	sizebin_width	binhalfwidth	b_raw	sizebinb_min	sizebinb_max
> 1	0		1	NA	NA	NA
> 2	0		1	NA	NA	NA
> 3	0		1	NA	NA	NA
> 4	0		1	NA	NA	NA
> 5	0		1	NA	NA	NA
> 6	0		1	NA	NA	NA
> 7	1		1	NA	NA	NA
> 8	2		1	NA	NA	NA
> 9	3		1	NA	NA	NA
> 10	4		1	NA	NA	NA
> 11	5		1	NA	NA	NA
> 12	6		1	NA	NA	NA
> 13	7		1	NA	NA	NA
> 14	8		1	NA	NA	NA
> 15	9		1	NA	NA	NA
> 16	10		1	NA	NA	NA
> 17	11		1	NA	NA	NA
> 18	12		1	NA	NA	NA
> 19	13		1	NA	NA	NA
> 20	14		1	NA	NA	NA

> 21	15	1	NA	NA	NA
> 22	16	1	NA	NA	NA
> 23	17	1	NA	NA	NA
> 24	18	1	NA	NA	NA
> 25	19	1	NA	NA	NA
> 26	20	1	NA	NA	NA
> 27	21	1	NA	NA	NA
> 28	22	1	NA	NA	NA
> 29	23	1	NA	NA	NA
> 30	24	1	NA	NA	NA
> 31	1	1	NA	NA	NA
> 32	2	1	NA	NA	NA
> 33	3	1	NA	NA	NA
> 34	4	1	NA	NA	NA
> 35	5	1	NA	NA	NA
> 36	6	1	NA	NA	NA
> 37	7	1	NA	NA	NA
> 38	8	1	NA	NA	NA
> 39	9	1	NA	NA	NA
> 40	10	1	NA	NA	NA
> 41	11	1	NA	NA	NA
> 42	12	1	NA	NA	NA
> 43	13	1	NA	NA	NA
> 44	14	1	NA	NA	NA
> 45	15	1	NA	NA	NA
> 46	16	1	NA	NA	NA
> 47	17	1	NA	NA	NA
> 48	18	1	NA	NA	NA
> 49	19	1	NA	NA	NA
> 50	20	1	NA	NA	NA
> 51	21	1	NA	NA	NA
> 52	22	1	NA	NA	NA
> 53	23	1	NA	NA	NA
> 54	24	1	NA	NA	NA
>	sizebinb_center	sizebinb_width	binhalfwidthc_raw	sizebinc_min	sizebinc_max
> 1	NA	NA	NA	NA	NA
> 2	NA	NA	NA	NA	NA
> 3	NA	NA	NA	NA	NA
> 4	NA	NA	NA	NA	NA
> 5	NA	NA	NA	NA	NA
> 6	NA	NA	NA	NA	NA
> 7	NA	NA	NA	NA	NA
> 8	NA	NA	NA	NA	NA
> 9	NA	NA	NA	NA	NA
> 10	NA	NA	NA	NA	NA
> 11	NA	NA	NA	NA	NA
> 12	NA	NA	NA	NA	NA
> 13	NA	NA	NA	NA	NA
> 14	NA	NA	NA	NA	NA
> 15	NA	NA	NA	NA	NA
> 16	NA	NA	NA	NA	NA

> 17	NA	NA	NA	NA	NA
> 18	NA	NA	NA	NA	NA
> 19	NA	NA	NA	NA	NA
> 20	NA	NA	NA	NA	NA
> 21	NA	NA	NA	NA	NA
> 22	NA	NA	NA	NA	NA
> 23	NA	NA	NA	NA	NA
> 24	NA	NA	NA	NA	NA
> 25	NA	NA	NA	NA	NA
> 26	NA	NA	NA	NA	NA
> 27	NA	NA	NA	NA	NA
> 28	NA	NA	NA	NA	NA
> 29	NA	NA	NA	NA	NA
> 30	NA	NA	NA	NA	NA
> 31	NA	NA	NA	NA	NA
> 32	NA	NA	NA	NA	NA
> 33	NA	NA	NA	NA	NA
> 34	NA	NA	NA	NA	NA
> 35	NA	NA	NA	NA	NA
> 36	NA	NA	NA	NA	NA
> 37	NA	NA	NA	NA	NA
> 38	NA	NA	NA	NA	NA
> 39	NA	NA	NA	NA	NA
> 40	NA	NA	NA	NA	NA
> 41	NA	NA	NA	NA	NA
> 42	NA	NA	NA	NA	NA
> 43	NA	NA	NA	NA	NA
> 44	NA	NA	NA	NA	NA
> 45	NA	NA	NA	NA	NA
> 46	NA	NA	NA	NA	NA
> 47	NA	NA	NA	NA	NA
> 48	NA	NA	NA	NA	NA
> 49	NA	NA	NA	NA	NA
> 50	NA	NA	NA	NA	NA
> 51	NA	NA	NA	NA	NA
> 52	NA	NA	NA	NA	NA
> 53	NA	NA	NA	NA	NA
> 54	NA	NA	NA	NA	NA
>	sizebinc_center	sizebinc_width	group		comments
> 1	NA	NA	0		Dormant seed
> 2	NA	NA	0		Yr1 protocorm
> 3	NA	NA	0		Yr2 protocorm
> 4	NA	NA	0		Yr3 protocorm
> 5	NA	NA	0		Seedling
> 6	NA	NA	0		Veg dorm
> 7	NA	NA	0		Veg adult 1 stem
> 8	NA	NA	0		Veg adult 2 stems
> 9	NA	NA	0		Veg adult 3 stems
> 10	NA	NA	0		Veg adult 4 stems
> 11	NA	NA	0		Veg adult 5 stems
> 12	NA	NA	0		Veg adult 6 stems

```

> 13      NA      NA      0 Veg adult 7 stems
> 14      NA      NA      0 Veg adult 8 stems
> 15      NA      NA      0 Veg adult 9 stems
> 16      NA      NA      0 Veg adult 10 stems
> 17      NA      NA      0 Veg adult 11 stems
> 18      NA      NA      0 Veg adult 12 stems
> 19      NA      NA      0 Veg adult 13 stems
> 20      NA      NA      0 Veg adult 14 stems
> 21      NA      NA      0 Veg adult 15 stems
> 22      NA      NA      0 Veg adult 16 stems
> 23      NA      NA      0 Veg adult 17 stems
> 24      NA      NA      0 Veg adult 18 stems
> 25      NA      NA      0 Veg adult 19 stems
> 26      NA      NA      0 Veg adult 20 stems
> 27      NA      NA      0 Veg adult 21 stems
> 28      NA      NA      0 Veg adult 22 stems
> 29      NA      NA      0 Veg adult 23 stems
> 30      NA      NA      0 Veg adult 24 stems
> 31      NA      NA      0 Flo adult 1 stem
> 32      NA      NA      0 Flo adult 2 stems
> 33      NA      NA      0 Flo adult 3 stems
> 34      NA      NA      0 Flo adult 4 stems
> 35      NA      NA      0 Flo adult 5 stems
> 36      NA      NA      0 Flo adult 6 stems
> 37      NA      NA      0 Flo adult 7 stems
> 38      NA      NA      0 Flo adult 8 stems
> 39      NA      NA      0 Flo adult 9 stems
> 40      NA      NA      0 Flo adult 10 stems
> 41      NA      NA      0 Flo adult 11 stems
> 42      NA      NA      0 Flo adult 12 stems
> 43      NA      NA      0 Flo adult 13 stems
> 44      NA      NA      0 Flo adult 14 stems
> 45      NA      NA      0 Flo adult 15 stems
> 46      NA      NA      0 Flo adult 16 stems
> 47      NA      NA      0 Flo adult 17 stems
> 48      NA      NA      0 Flo adult 18 stems
> 49      NA      NA      0 Flo adult 19 stems
> 50      NA      NA      0 Flo adult 20 stems
> 51      NA      NA      0 Flo adult 21 stems
> 52      NA      NA      0 Flo adult 22 stems
> 53      NA      NA      0 Flo adult 23 stems
> 54      NA      NA      0 Flo adult 24 stems

```

A close look at the resulting object, `cypframe`, shows a data frame that includes the following information in order for each stage: the stage's name, the associated size (in terms of up to three size metrics), its minimum and maximum age of occurrence (only used if creating an age-by-stage MPM), its reproductive status, its status as an observable stage, its status as a propagule stage, its status as an immature stage, its status as a mature stage, whether it occurs in the dataset, the half-width of its size class bin, the minimum and maximum of its size class bin, the centroid of its size class bin (currently the arithmetic mean), its full size class bin width (these last three variables are given for up to three size metrics in order), the stage group that the stage belongs to, and a comments field describing the stage. Stage names and combinations of characteristics for all stages occurring in the

dataset must be unique to prevent estimation errors, and the comments field may be edited to include any information deemed pertinent. Note that we did not include information on secondary and tertiary size, stage group, and the ages at which stages occur, but these may be supplied as well. We will cover these topics later in this book.

At this stage, we can build our function-based MPM. However, we might ask: How does the life history model differ if we wish to develop a raw MPM? The key difference is that we need to consider how large our dataset is, and to create only as many stages as can be routinely transitioned to and from based on the data. The life history model above in Figure 2.3, for example, is not usable for a raw MPM because we have cut the size bins too finely - it is likely that in a typical year, only some of these stages will have individuals actually transitioning between them. The impact of this is that we will end up with many zeros for transitions that, in a sufficiently large population, should not equal zero, and this fact will force our estimates of the population growth rate lower than necessary.

To deal with this problem, we need to explore the dataset to determine a reasonable number of life history stages and where the breaks should occur between these stages. A number of means exist to do this, and users should see Caswell (2001) and Kendall et al. (2019) for good discussions of the topic. Here, we suggest plotting a size distribution and assessing a series of numbers of natural breaks using the Jenks natural breaks algorithm (Jenks, 1967), or another such algorithm. Let's take a look at a distribution plot of size to help us with this process (figure 2.4).

```
plot(density(c(size.04, size.05, size.06, size.07, size.08, size.09),
  na.rm = TRUE), main = "", xlab = "Size (# of sprouts)", bty = "n")
```

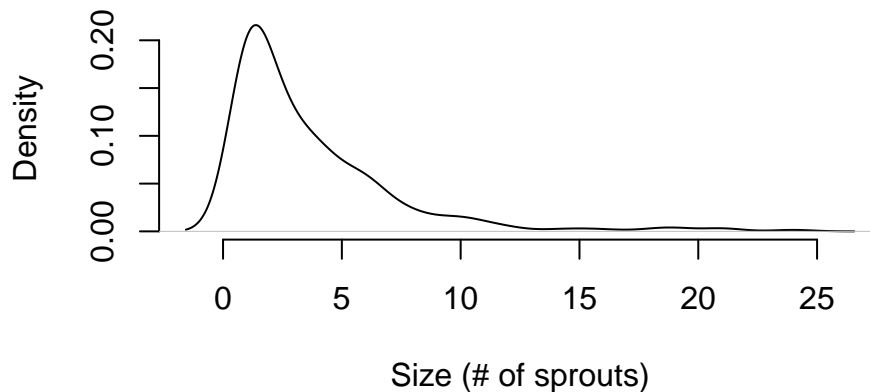


Figure 2.4: Distribution of size in *Cypridium candidum*

We can see that most individuals are small, and so our size data is densest around 1-2 sprouts or so. Large individuals are rare, so we will need to make size bins wider for large plants than for small plants. Let's try finding some natural breaks with the Jenks algorithm, separating into three, four, five, and six stages. To separate the size data into these numbers of stages, we need to identify a total of four, five, six, and seven breaks including the minimum and maximum. First, let's install the package `BAMMtools`, which includes the `getJenksBreaks()` function.

```
install.packages("BAMMtools", dependencies = TRUE)
```

Now let's see where we might fit natural breaks, and how many there might need to be.

```
BAMMtools::getJenksBreaks(c(size.04, size.05, size.06, size.07, size.08,
  size.09), k = 4)
> [1] 1 4 12 24
```

```

BAMMtools::getJenksBreaks(c(size.04, size.05, size.06, size.07, size.08,
  size.09), k = 5)
> [1] 1 3 7 14 24
BAMMtools::getJenksBreaks(c(size.04, size.05, size.06, size.07, size.08,
  size.09), k = 6)
> [1] 1 2 4 7 14 24
BAMMtools::getJenksBreaks(c(size.04, size.05, size.06, size.07, size.08,
  size.09), k = 7)
> [1] 1 2 4 7 11 16 24

```

The Jenks method gives us the borders of the size classes under different numbers of stages. The first line shows a three stage model, yielding four breaks including the minimum and maximum. This model has the first stage include 1-3 sprouts, the second stage include 4-11 sprouts, and the 3rd stage include 12-24 sprouts. The fourth line is a six stage model with seven breaks shown. Given what we know about the size distribution, we will try to separate the data into five stages (one sprout, 2-4 sprouts, 5-7 sprouts, 8-14 sprouts, and 15-24 sprouts), which will result in 11 total life history stages in our life history (one dormant seed, three protocorm stages of different age, one seedling stage, one vegetatively dormant stage, and five size-classified adult stages). Here is our new life history model (figure 2.5). Note that the hashed blue box here refers to the mature stages that are actually observable in our population.

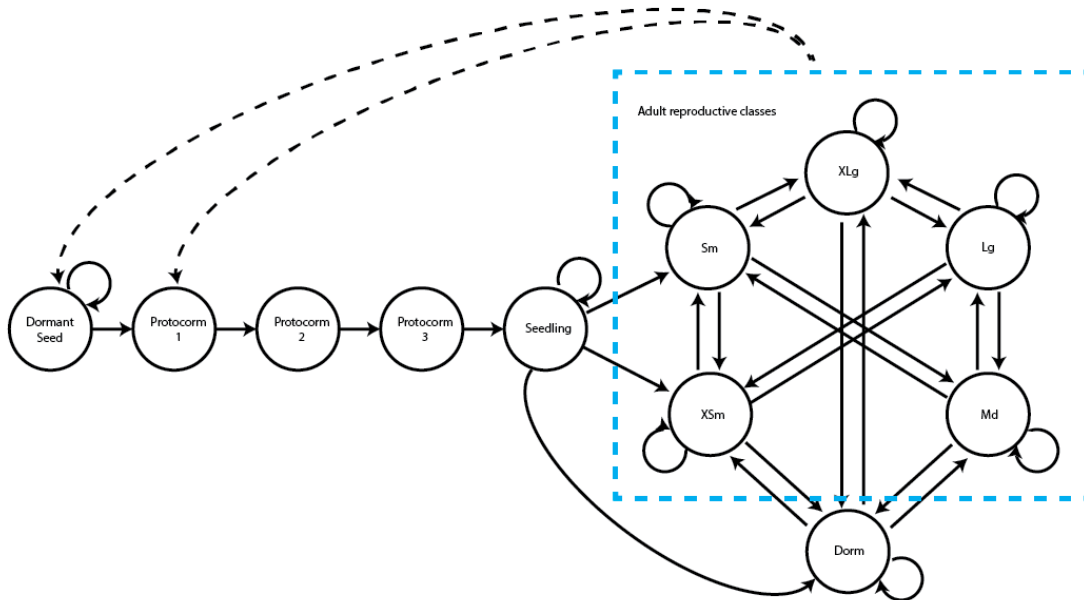


Figure 2.5: Life history model of *Cypripedium candidum* for use in raw MPMs

Let's now build a stageframe using these breaks. We will build vectors designating the same kinds of information as before, plus a new vector designating the halfwidth of the size bins associated with each stage (the `binvec` vector). The default halfwidth is 0.5, and while we used the default for the function-based stageframe, we cannot assume the default here.

```

sizevector <- c(0, 0, 0, 0, 0, 0, 1, 3, 6, 11, 19.5)
stagevector <- c("SD", "P1", "P2", "P3", "SL", "D", "XSm", "Sm", "Md", "Lg",
  "XLg")

```



```

repvector <- c(0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1)
obsvector <- c(0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1)
matvector <- c(0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1)
immvector <- c(0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0)
propvector <- c(1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0)
indataset <- c(0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1)
binvec <- c(0, 0, 0, 0, 0, 0.5, 0.5, 1.5, 1.5, 3.5, 5)

comments <- c("Dormant seed", "1st yr protocorm", "2nd yr protocorm",
  "3rd yr protocorm", "Seedling", "Dormant adult",
  "Extra small adult (1 shoot)", "Small adult (2-4 shoots)",
  "Medium adult (5-7 shoots)", "Large adult (8-14 shoots)",
  "Extra large adult (>14 shoots)")

cypframe_raw <- sf_create(sizes = sizevector, stagenames = stagevector,
  repstatus = repvector, obsstatus = obsvector, matstatus = matvector,
  propstatus = propvector, immstatus = immvector, indataset = indataset,
  binhalfwidth = binvec, comments = comments)

cypframe_raw
>   stage size size_b size_c min_age max_age repstatus obsstatus propstatus
> 1   SD  0.0    NA    NA    NA    NA    0          0          1
> 2   P1  0.0    NA    NA    NA    NA    0          0          0
> 3   P2  0.0    NA    NA    NA    NA    0          0          0
> 4   P3  0.0    NA    NA    NA    NA    0          0          0
> 5   SL  0.0    NA    NA    NA    NA    0          0          0
> 6    D  0.0    NA    NA    NA    NA    0          0          0
> 7  XSm  1.0    NA    NA    NA    NA    1          1          0
> 8   Sm  3.0    NA    NA    NA    NA    1          1          0
> 9   Md  6.0    NA    NA    NA    NA    1          1          0
> 10  Lg 11.0    NA    NA    NA    NA    1          1          0
> 11  XLg 19.5    NA    NA    NA    NA    1          1          0
>   immstatus matstatus indataset binhalfwidth_raw sizebin_min sizebin_max
> 1          0          0          0          0.0          0.0          0.0
> 2          1          0          0          0.0          0.0          0.0
> 3          1          0          0          0.0          0.0          0.0
> 4          1          0          0          0.0          0.0          0.0
> 5          1          0          0          0.0          0.0          0.0
> 6          0          1          1          0.5         -0.5          0.5
> 7          0          1          1          0.5          0.5          1.5
> 8          0          1          1          1.5          1.5          4.5
> 9          0          1          1          1.5          4.5          7.5
> 10         0          1          1          3.5          7.5         14.5
> 11         0          1          1          5.0         14.5         24.5
>   sizebin_center sizebin_width binhalfwidthb_raw sizebinb_min sizebinb_max
> 1              0.0             0              NA              NA              NA
> 2              0.0             0              NA              NA              NA
> 3              0.0             0              NA              NA              NA
> 4              0.0             0              NA              NA              NA
> 5              0.0             0              NA              NA              NA
> 6              0.0             1              NA              NA              NA

```

```

> 7          1.0          1          NA          NA          NA
> 8          3.0          3          NA          NA          NA
> 9          6.0          3          NA          NA          NA
> 10         11.0         7          NA          NA          NA
> 11         19.5         10         NA          NA          NA
>   sizebinb_center sizebinb_width binhalfwidthc_raw sizebinc_min sizebinc_max
> 1              NA              NA              NA              NA              NA
> 2              NA              NA              NA              NA              NA
> 3              NA              NA              NA              NA              NA
> 4              NA              NA              NA              NA              NA
> 5              NA              NA              NA              NA              NA
> 6              NA              NA              NA              NA              NA
> 7              NA              NA              NA              NA              NA
> 8              NA              NA              NA              NA              NA
> 9              NA              NA              NA              NA              NA
> 10             NA              NA              NA              NA              NA
> 11             NA              NA              NA              NA              NA
>   sizebinc_center sizebinc_width group          comments
> 1              NA              NA      0          Dormant seed
> 2              NA              NA      0          1st yr protocorm
> 3              NA              NA      0          2nd yr protocorm
> 4              NA              NA      0          3rd yr protocorm
> 5              NA              NA      0          Seedling
> 6              NA              NA      0          Dormant adult
> 7              NA              NA      0          Extra small adult (1 shoot)
> 8              NA              NA      0          Small adult (2-4 shoots)
> 9              NA              NA      0          Medium adult (5-7 shoots)
> 10             NA              NA      0          Large adult (8-14 shoots)
> 11             NA              NA      0          Extra large adult (>14 shoots)

```

2.3 Stage classification with multiple size metrics

So far, we have seen the simplest case of stage classification with size, where each stage is defined by a mutually exclusive size bin of a single size variable. However, more complicated cases certainly exist, in which stages should be defined on the basis of multiple size variables. Let's rethink the *Cypripedium candidum* example with this in mind.

In the previous two stageframes, we defined stages on the basis of the number of sprouts. However, many ecologists would argue that a flowering sprout is biologically quite different from a non-flowering sprout. Certainly, the amount of resources needed to produce a flowering sprout should be greater than that required for a non-flowering sprout. Of course, flowering sprouts actually produce offspring as well, making them demographically different from non-flowering sprouts. And the population actually includes an occasional, rare two-flowered sprout, as well. Can we redefine our life history model to use the numbers of non-flowering, one-flowered, and two-flowered sprouts to define stages?

To conduct this exercise, we first need to explore our data to see what combinations of different numbers of these three kinds of sprouts actually exist. We will do that using the `verticalize3()` function to standardize and format our demographic data first. We will describe this function in detail in the next chapter and so will not do so here. However, we will just point out that the three kinds of sprouts are input as `sizeacol`, `sizebcol`, and `sizeccol` in this function call, with `Veg.04` designating the number of non-flowering sprouts in 2004, `Inf.04` designating the number of single-flowered sprouts in 2004, and `Inf2.04` designating the number of double-flowered sprouts in 2004

(other years are incorporated via the blocksize option, but more on that in the next chapter).

```

vert.data.check <- verticalize3(cypdata, noyears = 6, firstyear = 2004,
  patchidcol = "patch", individcol = "plantid", blocksize = 4,
  sizeacol = "Veg.04", sizebcol = "Inf.04", sizeccol = "Inf2.04",
  repstracol = "Inf.04", repstrbcol = "Inf2.04", fecacol = "Pod.04",
  censorcol = "censor", censorkeep = 1, censorRepeat = FALSE,
  NAas0 = TRUE, age_offset = 4, censor = FALSE)
summary(vert.data.check)
>      rowid      popid patchid  individ      year2
> Min.   : 1.00    :320  A: 93   Length:320   Min.   :2004
> 1st Qu.:21.00          B:154  Class :character 1st Qu.:2005
> Median :37.50          C: 73   Mode  :character Median :2006
> Mean   :38.45                                     Mean   :2006
> 3rd Qu.:56.00                                     3rd Qu.:2007
> Max.   :77.00                                     Max.   :2008
>      firstseen      lastseen      obsage      obslifespan      sizea1
> Min.   :2004   Min.   :2004   Min.   :5.000   Min.   :0.000   Min.   : 0.0
> 1st Qu.:2004   1st Qu.:2009   1st Qu.:6.000   1st Qu.:5.000   1st Qu.: 0.0
> Median :2004   Median :2009   Median :7.000   Median :5.000   Median : 1.0
> Mean   :2004   Mean   :2009   Mean   :6.853   Mean   :4.556   Mean   : 1.9
> 3rd Qu.:2004   3rd Qu.:2009   3rd Qu.:8.000   3rd Qu.:5.000   3rd Qu.: 3.0
> Max.   :2008   Max.   :2009   Max.   :9.000   Max.   :5.000   Max.   :13.0
>      sizeb1      sizec1      size1added      repstral
> Min.   : 0.0000   Min.   :0.000000   Min.   : 0.000   Min.   : 0.0000
> 1st Qu.: 0.0000   1st Qu.:0.000000   1st Qu.: 0.000   1st Qu.: 0.0000
> Median : 0.0000   Median :0.000000   Median : 2.000   Median : 0.0000
> Mean   : 0.7469   Mean   :0.009375   Mean   : 2.656   Mean   : 0.7469
> 3rd Qu.: 1.0000   3rd Qu.:0.000000   3rd Qu.: 4.000   3rd Qu.: 1.0000
> Max.   :18.0000   Max.   :1.000000   Max.   :21.000   Max.   :18.0000
>      repstrb1      fecal      juvgiven1      obsstatus1
> Min.   :0.000000   Min.   :0.0000   Min.   :0   Min.   :0.0000
> 1st Qu.:0.000000   1st Qu.:0.0000   1st Qu.:0   1st Qu.:0.0000
> Median :0.000000   Median :0.0000   Median :0   Median :1.0000
> Mean   :0.009375   Mean   :0.2656   Mean   :0   Mean   :0.7469
> 3rd Qu.:0.000000   3rd Qu.:0.0000   3rd Qu.:0   3rd Qu.:1.0000
> Max.   :1.000000   Max.   :7.0000   Max.   :0   Max.   :1.0000
>      repstatus1      fecstatus1      matstatus1      alive1
> Min.   :0.0000   Min.   :0.0000   Min.   :1   Min.   :0.0000
> 1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:1   1st Qu.:1.0000
> Median :0.0000   Median :0.0000   Median :1   Median :1.0000
> Mean   :0.2875   Mean   :0.1344   Mean   :1   Mean   :0.7688
> 3rd Qu.:1.0000   3rd Qu.:0.0000   3rd Qu.:1   3rd Qu.:1.0000
> Max.   :1.0000   Max.   :1.0000   Max.   :1   Max.   :1.0000
>      stage1      stage1index      sizea2      sizeb2
> Length:320   Min.   :0   Min.   : 0.000   Min.   : 0.0000
> Class :character 1st Qu.:0   1st Qu.: 1.000   1st Qu.: 0.0000
> Mode  :character Median :0   Median : 2.000   Median : 0.0000
>      Mean   :0   Mean   : 2.416   Mean   : 0.8969
>      3rd Qu.:0   3rd Qu.: 3.000   3rd Qu.: 1.0000
>      Max.   :0   Max.   :13.000   Max.   :18.0000
>      sizec2      size2added      repstra2      repstrb2

```

```

> Min. :0.000000 Min. : 0.000 Min. : 0.0000 Min. :0.000000
> 1st Qu.:0.000000 1st Qu.: 1.000 1st Qu.: 0.0000 1st Qu.:0.000000
> Median :0.000000 Median : 2.000 Median : 0.0000 Median :0.000000
> Mean :0.009375 Mean : 3.322 Mean : 0.8969 Mean :0.009375
> 3rd Qu.:0.000000 3rd Qu.: 4.000 3rd Qu.: 1.0000 3rd Qu.:0.000000
> Max. :1.000000 Max. :24.000 Max. :18.0000 Max. :1.000000
> feca2 juvgiven2 obsstatus2 repstatus2
> Min. :0.0000 Min. :0 Min. :0.0000 Min. :0.0000
> 1st Qu.:0.0000 1st Qu.:0 1st Qu.:1.0000 1st Qu.:0.0000
> Median :0.0000 Median :0 Median :1.0000 Median :0.0000
> Mean :0.2906 Mean :0 Mean :0.9531 Mean :0.3688
> 3rd Qu.:0.0000 3rd Qu.:0 3rd Qu.:1.0000 3rd Qu.:1.0000
> Max. :7.0000 Max. :0 Max. :1.0000 Max. :1.0000
> fecstatus2 matstatus2 alive2 stage2 stage2index
> Min. :0.0000 Min. :1 Min. :1 Length:320 Min. :0
> 1st Qu.:0.0000 1st Qu.:1 1st Qu.:1 Class :character 1st Qu.:0
> Median :0.0000 Median :1 Median :1 Mode :character Median :0
> Mean :0.1562 Mean :1 Mean :1 Mean :0
> 3rd Qu.:0.0000 3rd Qu.:1 3rd Qu.:1 3rd Qu.:0
> Max. :1.0000 Max. :1 Max. :1 Max. :0
> sizea3 sizeb3 sizec3 size3added
> Min. : 0.000 Min. : 0.000 Min. :0.000000 Min. : 0.000
> 1st Qu.: 1.000 1st Qu.: 0.000 1st Qu.:0.000000 1st Qu.: 1.000
> Median : 1.000 Median : 0.000 Median :0.000000 Median : 2.000
> Mean : 2.209 Mean : 1.069 Mean :0.009375 Mean : 3.288
> 3rd Qu.: 3.000 3rd Qu.: 1.000 3rd Qu.:0.000000 3rd Qu.: 4.000
> Max. :13.000 Max. :18.000 Max. :1.000000 Max. :24.000
> repstra3 repstrb3 fec3 juvgiven3 obsstatus3
> Min. : 0.000 Min. :0.000000 Min. :0.0000 Min. :0 Min. :0.0
> 1st Qu.: 0.000 1st Qu.:0.000000 1st Qu.:0.0000 1st Qu.:0 1st Qu.:1.0
> Median : 0.000 Median :0.000000 Median :0.0000 Median :0 Median :1.0
> Mean : 1.069 Mean :0.009375 Mean :0.4562 Mean :0 Mean :0.9
> 3rd Qu.: 1.000 3rd Qu.:0.000000 3rd Qu.:0.0000 3rd Qu.:0 3rd Qu.:1.0
> Max. :18.000 Max. :1.000000 Max. :8.0000 Max. :0 Max. :1.0
> repstatus3 fecstatus3 matstatus3 alive3
> Min. :0.0 Min. :0.0000 Min. :1 Min. :0.0000
> 1st Qu.:0.0 1st Qu.:0.0000 1st Qu.:1 1st Qu.:1.0000
> Median :0.0 Median :0.0000 Median :1 Median :1.0000
> Mean :0.4 Mean :0.2219 Mean :1 Mean :0.9469
> 3rd Qu.:1.0 3rd Qu.:0.0000 3rd Qu.:1 3rd Qu.:1.0000
> Max. :1.0 Max. :1.0000 Max. :1 Max. :1.0000
> stage3 stage3index
> Length:320 Min. :0
> Class :character 1st Qu.:0
> Mode :character Median :0
> Mean :0
> 3rd Qu.:0
> Max. :0

```

The resulting object is a data frame, and we need to see what sorts of occurrences of each size actually occur within this data frame. For this purpose, we will use the `xtabs()` function in R's `stats` package to create contingency tables of the different sizes at time t .

```

xtabs(~ sizea2 + sizeb2 + sizec2, vert.data.check)
> , , sizec2 = 0
>
>      sizeb2
> sizea2  0  1  2  3  4  5  6  7  8 10 11 18
>    0  15 25  4  2  1  1  1  0  1  0  0  0
>    1  83 12  6  4  1  1  0  0  0  0  0  0
>    2  39  8  3  2  2  0  0  0  0  0  0  0
>    3  18  7  3  2  0  0  0  0  0  0  0  2
>    4  15  3  2  1  1  0  1  0  0  0  0  0
>    5  10  2  2  0  0  1  0  0  0  0  0  0
>    6  10  1  1  0  1  0  0  1  0  0  0  0
>    7   4  2  0  0  0  0  0  0  0  0  0  0
>    8   4  0  1  0  0  0  0  0  0  0  0  0
>    9   1  1  0  0  0  0  0  0  0  1  0  0
>   10   1  0  0  0  0  0  0  0  0  0  0  0
>   11   1  0  0  0  0  0  0  0  0  0  0  0
>   12   1  0  0  0  1  0  0  0  0  0  0  0
>   13   0  2  1  0  0  0  0  0  0  0  1  0
>
> , , sizec2 = 1
>
>      sizeb2
> sizea2  0  1  2  3  4  5  6  7  8 10 11 18
>    0  0  0  0  0  0  0  0  0  0  0  0  0
>    1  0  0  1  0  0  0  0  0  0  0  0  0
>    2  0  0  0  0  0  0  0  2  0  0  0  0
>    3  0  0  0  0  0  0  0  0  0  0  0  0
>    4  0  0  0  0  0  0  0  0  0  0  0  0
>    5  0  0  0  0  0  0  0  0  0  0  0  0
>    6  0  0  0  0  0  0  0  0  0  0  0  0
>    7  0  0  0  0  0  0  0  0  0  0  0  0
>    8  0  0  0  0  0  0  0  0  0  0  0  0
>    9  0  0  0  0  0  0  0  0  0  0  0  0
>   10  0  0  0  0  0  0  0  0  0  0  0  0
>   11  0  0  0  0  0  0  0  0  0  0  0  0
>   12  0  0  0  0  0  0  0  0  0  0  0  0
>   13  0  0  0  0  0  0  0  0  0  0  0  0

```

Clearly, not all size combinations are possible, and many combinations never occur. If we set up all combinations of sizes, including even a stage for 13 non-flowering sprouts and 18 single-flowered sprouts, then that stage will represent something that never actually happened in our dataset and may never happen in nature. The resulting matrices would likely include unrealistic statistics for these stages, potentially including artificially high survival. So, we will only build stages for combinations that seem likely. We might view these stages as occurring above a diagonal line running across the top contingency table, perhaps from around 11 or 12 non-flowering sprouts with no flowering sprouts, to 10 or 11 single-flowered sprouts with no non-flowering sprouts. Only the smallest two-flowered combinations will be included, as well.

```

sizevector.f <- c(0, 0, 0, 0, 0, 0, seq(1, 12, by = 1), seq(0, 9, by = 1),
  seq(0, 8, by = 1), seq(0, 7, by = 1), seq(0, 6, by = 1), seq(0, 5, by = 1),
  seq(0, 4, by = 1), seq(0, 3, by = 1), 0, 1, 2, 0, 1, 0,

```

```

0, 0, 1, 0)
sizebvector.f <- c(0, 0, 0, 0, 0, 0, rep(0, 12), rep(1, 10), rep(2, 9),
  rep(3, 8), rep(4, 7), rep(5, 6), rep(6, 5), rep(7, 4), rep(8, 3), 9, 9, 10,
  0, 1, 1, 2)
sizecvector.f <- c(0, 0, 0, 0, 0, 0, rep(0, 12), rep(0, 10), rep(0, 9),
  rep(0, 8), rep(0, 7), rep(0, 6), rep(0, 5), rep(0, 4), 0, 0, 0, 0, 0, 0,
  1, 1, 1, 1)
stagevector.f <- c("DS", "P1", "P2", "P3", "Sd1", "Dorm", "V1 I0 D0",
  "V2 I0 D0", "V3 I0 D0", "V4 I0 D0", "V5 I0 D0", "V6 I0 D0", "V7 I0 D0",
  "V8 I0 D0", "V9 I0 D0", "V10 I0 D0", "V11 I0 D0", "V12 I0 D0", "V0 I1 D0",
  "V1 I1 D0", "V2 I1 D0", "V3 I1 D0", "V4 I1 D0", "V5 I1 D0", "V6 I1 D0",
  "V7 I1 D0", "V8 I1 D0", "V9 I1 D0", "V0 I2 D0", "V1 I2 D0", "V2 I2 D0",
  "V3 I2 D0", "V4 I2 D0", "V5 I2 D0", "V6 I2 D0", "V7 I2 D0", "V8 I2 D0",
  "V0 I3 D0", "V1 I3 D0", "V2 I3 D0", "V3 I3 D0", "V4 I3 D0", "V5 I3 D0",
  "V6 I3 D0", "V7 I3 D0", "V0 I4 D0", "V1 I4 D0", "V2 I4 D0", "V3 I4 D0",
  "V4 I4 D0", "V5 I4 D0", "V6 I4 D0", "V0 I5 D0", "V1 I5 D0", "V2 I5 D0",
  "V3 I5 D0", "V4 I5 D0", "V5 I5 D0", "V0 I6 D0", "V1 I6 D0", "V2 I6 D0",
  "V3 I6 D0", "V4 I6 D0", "V0 I7 D0", "V1 I7 D0", "V2 I7 D0", "V3 I7 D0",
  "V0 I8 D0", "V1 I8 D0", "V2 I8 D0", "V0 I9 D0", "V1 I9 D0", "V0 I10 D0",
  "V0 I0 D1", "V0 I1 D1", "V1 I1 D1", "V0 I2 D1")
repvector.f <- c(0, 0, 0, 0, 0, 0, rep(0, 13), rep(1, 59))
obsvector.f <- c(0, 0, 0, 0, 0, 0, rep(1, 71))
matvector.f <- c(0, 0, 0, 0, 0, 0, rep(1, 72))
immvector.f <- c(0, 1, 1, 1, 1, 1, rep(0, 72))
propvector.f <- c(1, rep(0, 76))
indataset.f <- c(0, 0, 0, 0, 0, 0, rep(1, 72))
binvec.f <- c(0, 0, 0, 0, 0, 0, rep(0.5, 72))
binbvec.f <- c(0, 0, 0, 0, 0, 0, rep(0.5, 72))
bincvec.f <- c(0, 0, 0, 0, 0, 0, rep(0.5, 72))

vertframe.3f <- sf_create(sizes = sizevector.f, sizesb = sizebvector.f,
  sizesc = sizecvector.f, stagenames = stagevector.f, repstatus = repvector.f,
  obsstatus = obsvector.f, propstatus = propvector.f, immstatus = immvector.f,
  matstatus = matvector.f, indataset = indataset.f, binhalfwidth = binvec.f,
  binhalfwidthb = binbvec.f, binhalfwidthc = bincvec.f)

vertframe.3f
>
>   stage size size_b size_c min_age max_age repstatus obsstatus propstatus
> 1     DS    0     0     0      NA     NA           0           0           1
> 2     P1    0     0     0      NA     NA           0           0           0
> 3     P2    0     0     0      NA     NA           0           0           0
> 4     P3    0     0     0      NA     NA           0           0           0
> 5     Sd1   0     0     0      NA     NA           0           0           0
> 6     Dorm  0     0     0      NA     NA           0           0           0
> 7  V1 I0 D0  1     0     0      NA     NA           0           1           0
> 8  V2 I0 D0  2     0     0      NA     NA           0           1           0
> 9  V3 I0 D0  3     0     0      NA     NA           0           1           0
> 10 V4 I0 D0  4     0     0      NA     NA           0           1           0
> 11 V5 I0 D0  5     0     0      NA     NA           0           1           0
> 12 V6 I0 D0  6     0     0      NA     NA           0           1           0
> 13 V7 I0 D0  7     0     0      NA     NA           0           1           0

```

> 14	V8	I0	D0	8	0	0	NA	NA	0	1	0
> 15	V9	I0	D0	9	0	0	NA	NA	0	1	0
> 16	V10	I0	D0	10	0	0	NA	NA	0	1	0
> 17	V11	I0	D0	11	0	0	NA	NA	0	1	0
> 18	V12	I0	D0	12	0	0	NA	NA	0	1	0
> 19	V0	I1	D0	0	1	0	NA	NA	1	1	0
> 20	V1	I1	D0	1	1	0	NA	NA	1	1	0
> 21	V2	I1	D0	2	1	0	NA	NA	1	1	0
> 22	V3	I1	D0	3	1	0	NA	NA	1	1	0
> 23	V4	I1	D0	4	1	0	NA	NA	1	1	0
> 24	V5	I1	D0	5	1	0	NA	NA	1	1	0
> 25	V6	I1	D0	6	1	0	NA	NA	1	1	0
> 26	V7	I1	D0	7	1	0	NA	NA	1	1	0
> 27	V8	I1	D0	8	1	0	NA	NA	1	1	0
> 28	V9	I1	D0	9	1	0	NA	NA	1	1	0
> 29	V0	I2	D0	0	2	0	NA	NA	1	1	0
> 30	V1	I2	D0	1	2	0	NA	NA	1	1	0
> 31	V2	I2	D0	2	2	0	NA	NA	1	1	0
> 32	V3	I2	D0	3	2	0	NA	NA	1	1	0
> 33	V4	I2	D0	4	2	0	NA	NA	1	1	0
> 34	V5	I2	D0	5	2	0	NA	NA	1	1	0
> 35	V6	I2	D0	6	2	0	NA	NA	1	1	0
> 36	V7	I2	D0	7	2	0	NA	NA	1	1	0
> 37	V8	I2	D0	8	2	0	NA	NA	1	1	0
> 38	V0	I3	D0	0	3	0	NA	NA	1	1	0
> 39	V1	I3	D0	1	3	0	NA	NA	1	1	0
> 40	V2	I3	D0	2	3	0	NA	NA	1	1	0
> 41	V3	I3	D0	3	3	0	NA	NA	1	1	0
> 42	V4	I3	D0	4	3	0	NA	NA	1	1	0
> 43	V5	I3	D0	5	3	0	NA	NA	1	1	0
> 44	V6	I3	D0	6	3	0	NA	NA	1	1	0
> 45	V7	I3	D0	7	3	0	NA	NA	1	1	0
> 46	V0	I4	D0	0	4	0	NA	NA	1	1	0
> 47	V1	I4	D0	1	4	0	NA	NA	1	1	0
> 48	V2	I4	D0	2	4	0	NA	NA	1	1	0
> 49	V3	I4	D0	3	4	0	NA	NA	1	1	0
> 50	V4	I4	D0	4	4	0	NA	NA	1	1	0
> 51	V5	I4	D0	5	4	0	NA	NA	1	1	0
> 52	V6	I4	D0	6	4	0	NA	NA	1	1	0
> 53	V0	I5	D0	0	5	0	NA	NA	1	1	0
> 54	V1	I5	D0	1	5	0	NA	NA	1	1	0
> 55	V2	I5	D0	2	5	0	NA	NA	1	1	0
> 56	V3	I5	D0	3	5	0	NA	NA	1	1	0
> 57	V4	I5	D0	4	5	0	NA	NA	1	1	0
> 58	V5	I5	D0	5	5	0	NA	NA	1	1	0
> 59	V0	I6	D0	0	6	0	NA	NA	1	1	0
> 60	V1	I6	D0	1	6	0	NA	NA	1	1	0
> 61	V2	I6	D0	2	6	0	NA	NA	1	1	0
> 62	V3	I6	D0	3	6	0	NA	NA	1	1	0
> 63	V4	I6	D0	4	6	0	NA	NA	1	1	0
> 64	V0	I7	D0	0	7	0	NA	NA	1	1	0

```

> 65 V1 I7 D0 1 7 0 NA NA 1 1 0
> 66 V2 I7 D0 2 7 0 NA NA 1 1 0
> 67 V3 I7 D0 3 7 0 NA NA 1 1 0
> 68 V0 I8 D0 0 8 0 NA NA 1 1 0
> 69 V1 I8 D0 1 8 0 NA NA 1 1 0
> 70 V2 I8 D0 2 8 0 NA NA 1 1 0
> 71 V0 I9 D0 0 9 0 NA NA 1 1 0
> 72 V1 I9 D0 1 9 0 NA NA 1 1 0
> 73 V0 I10 D0 0 10 0 NA NA 1 1 0
> 74 V0 I0 D1 0 0 1 NA NA 1 1 0
> 75 V0 I1 D1 0 1 1 NA NA 1 1 0
> 76 V1 I1 D1 1 1 1 NA NA 1 1 0
> 77 V0 I2 D1 0 2 1 NA NA 1 1 0
>   immstatus matstatus indataset binhalfwidth_raw sizebin_min sizebin_max
> 1 0 0 0 0.0 0.0 0.0
> 2 1 0 0 0.0 0.0 0.0
> 3 1 0 0 0.0 0.0 0.0
> 4 1 0 0 0.0 0.0 0.0
> 5 1 0 0 0.0 0.0 0.0
> 6 0 1 1 0.5 -0.5 0.5
> 7 0 1 1 0.5 0.5 1.5
> 8 0 1 1 0.5 1.5 2.5
> 9 0 1 1 0.5 2.5 3.5
> 10 0 1 1 0.5 3.5 4.5
> 11 0 1 1 0.5 4.5 5.5
> 12 0 1 1 0.5 5.5 6.5
> 13 0 1 1 0.5 6.5 7.5
> 14 0 1 1 0.5 7.5 8.5
> 15 0 1 1 0.5 8.5 9.5
> 16 0 1 1 0.5 9.5 10.5
> 17 0 1 1 0.5 10.5 11.5
> 18 0 1 1 0.5 11.5 12.5
> 19 0 1 1 0.5 -0.5 0.5
> 20 0 1 1 0.5 0.5 1.5
> 21 0 1 1 0.5 1.5 2.5
> 22 0 1 1 0.5 2.5 3.5
> 23 0 1 1 0.5 3.5 4.5
> 24 0 1 1 0.5 4.5 5.5
> 25 0 1 1 0.5 5.5 6.5
> 26 0 1 1 0.5 6.5 7.5
> 27 0 1 1 0.5 7.5 8.5
> 28 0 1 1 0.5 8.5 9.5
> 29 0 1 1 0.5 -0.5 0.5
> 30 0 1 1 0.5 0.5 1.5
> 31 0 1 1 0.5 1.5 2.5
> 32 0 1 1 0.5 2.5 3.5
> 33 0 1 1 0.5 3.5 4.5
> 34 0 1 1 0.5 4.5 5.5
> 35 0 1 1 0.5 5.5 6.5
> 36 0 1 1 0.5 6.5 7.5
> 37 0 1 1 0.5 7.5 8.5

```



```

> 38      0      1      1      0.5      -0.5      0.5
> 39      0      1      1      0.5      0.5      1.5
> 40      0      1      1      0.5      1.5      2.5
> 41      0      1      1      0.5      2.5      3.5
> 42      0      1      1      0.5      3.5      4.5
> 43      0      1      1      0.5      4.5      5.5
> 44      0      1      1      0.5      5.5      6.5
> 45      0      1      1      0.5      6.5      7.5
> 46      0      1      1      0.5      -0.5      0.5
> 47      0      1      1      0.5      0.5      1.5
> 48      0      1      1      0.5      1.5      2.5
> 49      0      1      1      0.5      2.5      3.5
> 50      0      1      1      0.5      3.5      4.5
> 51      0      1      1      0.5      4.5      5.5
> 52      0      1      1      0.5      5.5      6.5
> 53      0      1      1      0.5      -0.5      0.5
> 54      0      1      1      0.5      0.5      1.5
> 55      0      1      1      0.5      1.5      2.5
> 56      0      1      1      0.5      2.5      3.5
> 57      0      1      1      0.5      3.5      4.5
> 58      0      1      1      0.5      4.5      5.5
> 59      0      1      1      0.5      -0.5      0.5
> 60      0      1      1      0.5      0.5      1.5
> 61      0      1      1      0.5      1.5      2.5
> 62      0      1      1      0.5      2.5      3.5
> 63      0      1      1      0.5      3.5      4.5
> 64      0      1      1      0.5      -0.5      0.5
> 65      0      1      1      0.5      0.5      1.5
> 66      0      1      1      0.5      1.5      2.5
> 67      0      1      1      0.5      2.5      3.5
> 68      0      1      1      0.5      -0.5      0.5
> 69      0      1      1      0.5      0.5      1.5
> 70      0      1      1      0.5      1.5      2.5
> 71      0      1      1      0.5      -0.5      0.5
> 72      0      1      1      0.5      0.5      1.5
> 73      0      1      1      0.5      -0.5      0.5
> 74      0      1      1      0.5      -0.5      0.5
> 75      0      1      1      0.5      -0.5      0.5
> 76      0      1      1      0.5      0.5      1.5
> 77      0      1      1      0.5      -0.5      0.5
>      sizebin_center sizebin_width binhalfwidthb_raw sizebinb_min sizebinb_max
> 1          0          0          0.0          0.0          0.0
> 2          0          0          0.0          0.0          0.0
> 3          0          0          0.0          0.0          0.0
> 4          0          0          0.0          0.0          0.0
> 5          0          0          0.0          0.0          0.0
> 6          0          1          0.5          -0.5          0.5
> 7          1          1          0.5          -0.5          0.5
> 8          2          1          0.5          -0.5          0.5
> 9          3          1          0.5          -0.5          0.5
> 10         4          1          0.5          -0.5          0.5

```

> 11	5	1	0.5	-0.5	0.5
> 12	6	1	0.5	-0.5	0.5
> 13	7	1	0.5	-0.5	0.5
> 14	8	1	0.5	-0.5	0.5
> 15	9	1	0.5	-0.5	0.5
> 16	10	1	0.5	-0.5	0.5
> 17	11	1	0.5	-0.5	0.5
> 18	12	1	0.5	-0.5	0.5
> 19	0	1	0.5	0.5	1.5
> 20	1	1	0.5	0.5	1.5
> 21	2	1	0.5	0.5	1.5
> 22	3	1	0.5	0.5	1.5
> 23	4	1	0.5	0.5	1.5
> 24	5	1	0.5	0.5	1.5
> 25	6	1	0.5	0.5	1.5
> 26	7	1	0.5	0.5	1.5
> 27	8	1	0.5	0.5	1.5
> 28	9	1	0.5	0.5	1.5
> 29	0	1	0.5	1.5	2.5
> 30	1	1	0.5	1.5	2.5
> 31	2	1	0.5	1.5	2.5
> 32	3	1	0.5	1.5	2.5
> 33	4	1	0.5	1.5	2.5
> 34	5	1	0.5	1.5	2.5
> 35	6	1	0.5	1.5	2.5
> 36	7	1	0.5	1.5	2.5
> 37	8	1	0.5	1.5	2.5
> 38	0	1	0.5	2.5	3.5
> 39	1	1	0.5	2.5	3.5
> 40	2	1	0.5	2.5	3.5
> 41	3	1	0.5	2.5	3.5
> 42	4	1	0.5	2.5	3.5
> 43	5	1	0.5	2.5	3.5
> 44	6	1	0.5	2.5	3.5
> 45	7	1	0.5	2.5	3.5
> 46	0	1	0.5	3.5	4.5
> 47	1	1	0.5	3.5	4.5
> 48	2	1	0.5	3.5	4.5
> 49	3	1	0.5	3.5	4.5
> 50	4	1	0.5	3.5	4.5
> 51	5	1	0.5	3.5	4.5
> 52	6	1	0.5	3.5	4.5
> 53	0	1	0.5	4.5	5.5
> 54	1	1	0.5	4.5	5.5
> 55	2	1	0.5	4.5	5.5
> 56	3	1	0.5	4.5	5.5
> 57	4	1	0.5	4.5	5.5
> 58	5	1	0.5	4.5	5.5
> 59	0	1	0.5	5.5	6.5
> 60	1	1	0.5	5.5	6.5
> 61	2	1	0.5	5.5	6.5

```

> 62          3          1          0.5          5.5          6.5
> 63          4          1          0.5          5.5          6.5
> 64          0          1          0.5          6.5          7.5
> 65          1          1          0.5          6.5          7.5
> 66          2          1          0.5          6.5          7.5
> 67          3          1          0.5          6.5          7.5
> 68          0          1          0.5          7.5          8.5
> 69          1          1          0.5          7.5          8.5
> 70          2          1          0.5          7.5          8.5
> 71          0          1          0.5          8.5          9.5
> 72          1          1          0.5          8.5          9.5
> 73          0          1          0.5          9.5         10.5
> 74          0          1          0.5          -0.5          0.5
> 75          0          1          0.5          0.5          1.5
> 76          1          1          0.5          0.5          1.5
> 77          0          1          0.5          1.5          2.5
>  sizebinb_center sizebinb_width binhalfwidthhc_raw sizebinc_min sizebinc_max
> 1              0              0              0.0              0.0              0.0
> 2              0              0              0.0              0.0              0.0
> 3              0              0              0.0              0.0              0.0
> 4              0              0              0.0              0.0              0.0
> 5              0              0              0.0              0.0              0.0
> 6              0              1              0.5              -0.5              0.5
> 7              0              1              0.5              -0.5              0.5
> 8              0              1              0.5              -0.5              0.5
> 9              0              1              0.5              -0.5              0.5
> 10             0              1              0.5              -0.5              0.5
> 11             0              1              0.5              -0.5              0.5
> 12             0              1              0.5              -0.5              0.5
> 13             0              1              0.5              -0.5              0.5
> 14             0              1              0.5              -0.5              0.5
> 15             0              1              0.5              -0.5              0.5
> 16             0              1              0.5              -0.5              0.5
> 17             0              1              0.5              -0.5              0.5
> 18             0              1              0.5              -0.5              0.5
> 19             1              1              0.5              -0.5              0.5
> 20             1              1              0.5              -0.5              0.5
> 21             1              1              0.5              -0.5              0.5
> 22             1              1              0.5              -0.5              0.5
> 23             1              1              0.5              -0.5              0.5
> 24             1              1              0.5              -0.5              0.5
> 25             1              1              0.5              -0.5              0.5
> 26             1              1              0.5              -0.5              0.5
> 27             1              1              0.5              -0.5              0.5
> 28             1              1              0.5              -0.5              0.5
> 29             2              1              0.5              -0.5              0.5
> 30             2              1              0.5              -0.5              0.5
> 31             2              1              0.5              -0.5              0.5
> 32             2              1              0.5              -0.5              0.5
> 33             2              1              0.5              -0.5              0.5
> 34             2              1              0.5              -0.5              0.5

```

```

> 35      2      1      0.5      -0.5      0.5
> 36      2      1      0.5      -0.5      0.5
> 37      2      1      0.5      -0.5      0.5
> 38      3      1      0.5      -0.5      0.5
> 39      3      1      0.5      -0.5      0.5
> 40      3      1      0.5      -0.5      0.5
> 41      3      1      0.5      -0.5      0.5
> 42      3      1      0.5      -0.5      0.5
> 43      3      1      0.5      -0.5      0.5
> 44      3      1      0.5      -0.5      0.5
> 45      3      1      0.5      -0.5      0.5
> 46      4      1      0.5      -0.5      0.5
> 47      4      1      0.5      -0.5      0.5
> 48      4      1      0.5      -0.5      0.5
> 49      4      1      0.5      -0.5      0.5
> 50      4      1      0.5      -0.5      0.5
> 51      4      1      0.5      -0.5      0.5
> 52      4      1      0.5      -0.5      0.5
> 53      5      1      0.5      -0.5      0.5
> 54      5      1      0.5      -0.5      0.5
> 55      5      1      0.5      -0.5      0.5
> 56      5      1      0.5      -0.5      0.5
> 57      5      1      0.5      -0.5      0.5
> 58      5      1      0.5      -0.5      0.5
> 59      6      1      0.5      -0.5      0.5
> 60      6      1      0.5      -0.5      0.5
> 61      6      1      0.5      -0.5      0.5
> 62      6      1      0.5      -0.5      0.5
> 63      6      1      0.5      -0.5      0.5
> 64      7      1      0.5      -0.5      0.5
> 65      7      1      0.5      -0.5      0.5
> 66      7      1      0.5      -0.5      0.5
> 67      7      1      0.5      -0.5      0.5
> 68      8      1      0.5      -0.5      0.5
> 69      8      1      0.5      -0.5      0.5
> 70      8      1      0.5      -0.5      0.5
> 71      9      1      0.5      -0.5      0.5
> 72      9      1      0.5      -0.5      0.5
> 73     10      1      0.5      -0.5      0.5
> 74      0      1      0.5      0.5      1.5
> 75      1      1      0.5      0.5      1.5
> 76      1      1      0.5      0.5      1.5
> 77      2      1      0.5      0.5      1.5
>      sizebinc_center sizebinc_width group      comments
> 1      0      0      0      0 No description
> 2      0      0      0      0 No description
> 3      0      0      0      0 No description
> 4      0      0      0      0 No description
> 5      0      0      0      0 No description
> 6      0      1      0      0 No description
> 7      0      1      0      0 No description

```

> 8	0	1	0 No description
> 9	0	1	0 No description
> 10	0	1	0 No description
> 11	0	1	0 No description
> 12	0	1	0 No description
> 13	0	1	0 No description
> 14	0	1	0 No description
> 15	0	1	0 No description
> 16	0	1	0 No description
> 17	0	1	0 No description
> 18	0	1	0 No description
> 19	0	1	0 No description
> 20	0	1	0 No description
> 21	0	1	0 No description
> 22	0	1	0 No description
> 23	0	1	0 No description
> 24	0	1	0 No description
> 25	0	1	0 No description
> 26	0	1	0 No description
> 27	0	1	0 No description
> 28	0	1	0 No description
> 29	0	1	0 No description
> 30	0	1	0 No description
> 31	0	1	0 No description
> 32	0	1	0 No description
> 33	0	1	0 No description
> 34	0	1	0 No description
> 35	0	1	0 No description
> 36	0	1	0 No description
> 37	0	1	0 No description
> 38	0	1	0 No description
> 39	0	1	0 No description
> 40	0	1	0 No description
> 41	0	1	0 No description
> 42	0	1	0 No description
> 43	0	1	0 No description
> 44	0	1	0 No description
> 45	0	1	0 No description
> 46	0	1	0 No description
> 47	0	1	0 No description
> 48	0	1	0 No description
> 49	0	1	0 No description
> 50	0	1	0 No description
> 51	0	1	0 No description
> 52	0	1	0 No description
> 53	0	1	0 No description
> 54	0	1	0 No description
> 55	0	1	0 No description
> 56	0	1	0 No description
> 57	0	1	0 No description
> 58	0	1	0 No description

```

> 59      0      1      0 No description
> 60      0      1      0 No description
> 61      0      1      0 No description
> 62      0      1      0 No description
> 63      0      1      0 No description
> 64      0      1      0 No description
> 65      0      1      0 No description
> 66      0      1      0 No description
> 67      0      1      0 No description
> 68      0      1      0 No description
> 69      0      1      0 No description
> 70      0      1      0 No description
> 71      0      1      0 No description
> 72      0      1      0 No description
> 73      0      1      0 No description
> 74      1      1      0 No description
> 75      1      1      0 No description
> 76      1      1      0 No description
> 77      1      1      0 No description

```

The result is our biggest stageframe yet, but a realistic one that we will go back to later.

2.4 Automating the creation of large numbers of stages

There are times when we wish to create extremely large life history models, with many tens of life stages. For example, users wishing to develop IPMs may wish to create stageframes with over 100 stages, most of which would differ only by size. Developing stageframes for these situations can take some time and effort. Fortunately, `lefko3` includes a method to make this quick and easy.

The key innovation is to use an option in the `sf_create()` function that allows us to set the minimum and maximum sizes for a group of stages that are alike in every way except for size. A further option can set the number of total size classes to fit within these bounds. Let's try an example, using the *Cypripedium candidum* case.

The size metrics that we will use in the *Cypripedium candidum* dataset are all count variables. However, let us pretend that the total number of sprouts is actually a continuous, quantitative variable with a minimum of 1 and a maximum of 23. If we wished to create 100 stages within these bounds, we could do so by identifying the minimum and maximum size values with the "ipm" marker in the stage names. The `ipmbins` option defaults to 100, but let us say so explicitly here. We will then look at just a few key variables in the resulting stageframe.

```

sizevector <- c(0, 0, 0, 0, 0, 0, 1, 23)
stagevector <- c("SD", "P1", "P2", "P3", "SL", "D", "ipm", "ipm")
repvector <- c(0, 0, 0, 0, 0, 0, 1, 1)
obsvector <- c(0, 0, 0, 0, 0, 0, 1, 1)
matvector <- c(0, 0, 0, 0, 0, 1, 1, 1)
immvector <- c(0, 1, 1, 1, 1, 0, 0, 0)
propvector <- c(1, 0, 0, 0, 0, 0, 0, 0)
indataset <- c(0, 0, 0, 0, 0, 1, 1, 1)
binvec <- c(0, 0, 0, 0, 0, 0.5, NA, NA)

cyprframe_ipm <- sf_create(sizes = sizevector, stagenames = stagevector,
  repstatus = repvector, obsstatus = obsvector, matstatus = matvector,

```

```

propstatus = propvector, immstatus = immvector, indataset = indataset,
binhalfwidth = binvec, ipmbins = 100)

cypframe_ipm[,c("stage", "size", "sizebin_min", "sizebin_max")]
>
>      stage  size sizebin_min sizebin_max
> 1      SD  0.00         0.00         0.00
> 2      P1  0.00         0.00         0.00
> 3      P2  0.00         0.00         0.00
> 4      P3  0.00         0.00         0.00
> 5      SL  0.00         0.00         0.00
> 6       D  0.00        -0.50         0.50
> 7  sza_1.1100_0  1.11         1.00         1.22
> 8  sza_1.3300_0  1.33         1.22         1.44
> 9  sza_1.5500_0  1.55         1.44         1.66
> 10  sza_1.7700_0  1.77         1.66         1.88
> 11  sza_1.9900_0  1.99         1.88         2.10
> 12  sza_2.2100_0  2.21         2.10         2.32
> 13  sza_2.4300_0  2.43         2.32         2.54
> 14  sza_2.6500_0  2.65         2.54         2.76
> 15  sza_2.8700_0  2.87         2.76         2.98
> 16  sza_3.0900_0  3.09         2.98         3.20
> 17  sza_3.3100_0  3.31         3.20         3.42
> 18  sza_3.5300_0  3.53         3.42         3.64
> 19  sza_3.7500_0  3.75         3.64         3.86
> 20  sza_3.9700_0  3.97         3.86         4.08
> 21  sza_4.1900_0  4.19         4.08         4.30
> 22  sza_4.4100_0  4.41         4.30         4.52
> 23  sza_4.6300_0  4.63         4.52         4.74
> 24  sza_4.8500_0  4.85         4.74         4.96
> 25  sza_5.0700_0  5.07         4.96         5.18
> 26  sza_5.2900_0  5.29         5.18         5.40
> 27  sza_5.5100_0  5.51         5.40         5.62
> 28  sza_5.7300_0  5.73         5.62         5.84
> 29  sza_5.9500_0  5.95         5.84         6.06
> 30  sza_6.1700_0  6.17         6.06         6.28
> 31  sza_6.3900_0  6.39         6.28         6.50
> 32  sza_6.6100_0  6.61         6.50         6.72
> 33  sza_6.8300_0  6.83         6.72         6.94
> 34  sza_7.0500_0  7.05         6.94         7.16
> 35  sza_7.2700_0  7.27         7.16         7.38
> 36  sza_7.4900_0  7.49         7.38         7.60
> 37  sza_7.7100_0  7.71         7.60         7.82
> 38  sza_7.9300_0  7.93         7.82         8.04
> 39  sza_8.1500_0  8.15         8.04         8.26
> 40  sza_8.3700_0  8.37         8.26         8.48
> 41  sza_8.5900_0  8.59         8.48         8.70
> 42  sza_8.8100_0  8.81         8.70         8.92
> 43  sza_9.0300_0  9.03         8.92         9.14
> 44  sza_9.2500_0  9.25         9.14         9.36
> 45  sza_9.4700_0  9.47         9.36         9.58
> 46  sza_9.6900_0  9.69         9.58         9.80

```

> 47	sza_9.9100_0	9.91	9.80	10.02
> 48	sza_10.130_0	10.13	10.02	10.24
> 49	sza_10.350_0	10.35	10.24	10.46
> 50	sza_10.570_0	10.57	10.46	10.68
> 51	sza_10.790_0	10.79	10.68	10.90
> 52	sza_11.010_0	11.01	10.90	11.12
> 53	sza_11.230_0	11.23	11.12	11.34
> 54	sza_11.450_0	11.45	11.34	11.56
> 55	sza_11.670_0	11.67	11.56	11.78
> 56	sza_11.890_0	11.89	11.78	12.00
> 57	sza_12.110_0	12.11	12.00	12.22
> 58	sza_12.330_0	12.33	12.22	12.44
> 59	sza_12.550_0	12.55	12.44	12.66
> 60	sza_12.770_0	12.77	12.66	12.88
> 61	sza_12.990_0	12.99	12.88	13.10
> 62	sza_13.210_0	13.21	13.10	13.32
> 63	sza_13.430_0	13.43	13.32	13.54
> 64	sza_13.650_0	13.65	13.54	13.76
> 65	sza_13.870_0	13.87	13.76	13.98
> 66	sza_14.090_0	14.09	13.98	14.20
> 67	sza_14.310_0	14.31	14.20	14.42
> 68	sza_14.530_0	14.53	14.42	14.64
> 69	sza_14.750_0	14.75	14.64	14.86
> 70	sza_14.970_0	14.97	14.86	15.08
> 71	sza_15.190_0	15.19	15.08	15.30
> 72	sza_15.410_0	15.41	15.30	15.52
> 73	sza_15.630_0	15.63	15.52	15.74
> 74	sza_15.850_0	15.85	15.74	15.96
> 75	sza_16.070_0	16.07	15.96	16.18
> 76	sza_16.290_0	16.29	16.18	16.40
> 77	sza_16.510_0	16.51	16.40	16.62
> 78	sza_16.730_0	16.73	16.62	16.84
> 79	sza_16.950_0	16.95	16.84	17.06
> 80	sza_17.170_0	17.17	17.06	17.28
> 81	sza_17.390_0	17.39	17.28	17.50
> 82	sza_17.610_0	17.61	17.50	17.72
> 83	sza_17.830_0	17.83	17.72	17.94
> 84	sza_18.050_0	18.05	17.94	18.16
> 85	sza_18.270_0	18.27	18.16	18.38
> 86	sza_18.490_0	18.49	18.38	18.60
> 87	sza_18.710_0	18.71	18.60	18.82
> 88	sza_18.930_0	18.93	18.82	19.04
> 89	sza_19.150_0	19.15	19.04	19.26
> 90	sza_19.370_0	19.37	19.26	19.48
> 91	sza_19.590_0	19.59	19.48	19.70
> 92	sza_19.810_0	19.81	19.70	19.92
> 93	sza_20.030_0	20.03	19.92	20.14
> 94	sza_20.250_0	20.25	20.14	20.36
> 95	sza_20.470_0	20.47	20.36	20.58
> 96	sza_20.690_0	20.69	20.58	20.80
> 97	sza_20.910_0	20.91	20.80	21.02


```

> 98  sza_21.130_0 21.13      21.02      21.24
> 99  sza_21.350_0 21.35      21.24      21.46
> 100 sza_21.570_0 21.57      21.46      21.68
> 101 sza_21.790_0 21.79      21.68      21.90
> 102 sza_22.010_0 22.01      21.90      22.12
> 103 sza_22.230_0 22.23      22.12      22.34
> 104 sza_22.450_0 22.45      22.34      22.56
> 105 sza_22.670_0 22.67      22.56      22.78
> 106 sza_22.890_0 22.89      22.78      23.00

```

We see that we have created a total of 106 stages - six early-life stages, followed by 100 stages that were developed automatically. These stages have size bins that do not overlap with neighboring sizes, and cover all possible sizes in the range.

We will see more uses of this approach in chapter 7 on integral projection models. However, it is important to point out that this approach can be used for multiple groups of stages, each with its own unique combinations of stage characteristics and with their own size minima and maxima. This approach can also be used in multiple size classification scenarios. More on this later.

At this point, we have four stageframes and can move ahead to standardizing and formatting our demographic datasets and creating some MPMs with them. There are further options that we can use to develop more complicated life history models, including the use of stage grouping to help properly format MPMs, and the designation of ages for each stage for use in age-by-stage MPMs. We will explore such issues in later chapters. For further information on life history model development, see chapters 3 and 4 in Caswell (2001) for a good treatment useful for beginners. We also note Kendall et al. (2019) as a good reference detailing common problems in MPM construction and how to avoid them through proper development and operationalization of life history models. Beissinger and Westphal (1998), Wardle (1998), and Salguero-Gómez and Casper (2010) provide good discussions of the proper application of life history models to understand population dynamics through the MPM approach.

2.5 Points to remember

1. All MPMs and IPMs require appropriate and carefully thought out models of the life history of the organism.
2. Good life history models require appropriately defined life history stages, which are defined not simply by their size and reproductive characteristics, but also by their stage duration, their relationships to other stages, by the actual numbers of different combinations of life history characteristics occurring in the dataset, and by the choice of raw vs. function-based MPM.
3. Function `sf_create()` will develop a data frame identifying each stage in the life history model, and this data frame will be used at each step of the MPM development and analysis process.
4. Function `sf_create()` can also be used to define large suites of stages using shorthand designations for groups of stages. This is particularly useful in developing discretized IPMs.