

# An Introduction to CAMer package

**CAMer** package (Continuous Admixture Modeler) does Continuous Admixture Modeling (CAM) and related summary based on the result of *iMAAPs*. It introduces three new S3 classes, **CAM.single**, **CAM** and **CAM.conclusion**, and some corresponding methods. It also contains some utility functions and two simulated data sets (*CGF\_50* and *GA\_I*) for illustration.

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## Computation

### Single LD Decay Curve

The function `singleCAM()` does CAM for a single LD decay curve. For example, let's use the *CGF\_50* data set (the admixture proportion for population 1 ( $m_1$ ) is 0.3) to do CAM with the most ancient generation concerned being 70 ( $T=70L$ ) and core models being HI, CGF1, CGF2 and GA (`isolation=FALSE`):

```
library(CAMer)
data(CGF_50)
d<-CGF_50$Distance
Z<-CGF_50$Combined_LD
fit<-singleCAM(d=d,Z=Z,m1=0.3,T=70L,isolation=FALSE)
fit

## Continuous Admixture Inference (CAM) for a Single LD Decay Curve
##
## Function call: singleCAM(d = d, Z = Z, m1 = 0.3, T = 70L, isolation = FALSE)
##
## Length of Used LD: 3497
##
##   Model Start End      msE
##   HI      23  23 8.912686e-06
##   CGF1     49   1 1.654922e-06
##   CGF2     60   1 2.750241e-06
##   GA      53   1 5.509048e-06
```

where parameter `d` corresponds to genetic distance and parameter `Z` corresponds to an LD decay curve.

One can also specify the file path of the .log file containing the information of `m1` in argument `m1=`.

Here the class of `fit` is **CAM.single**, and it has its own method for `print()`. `fit$summary` is a more comprehensive data frame containing the data frame printed.

Parallel computation is also supported provided that **parallel** package or **snow** package is installed. For newer versions of R ( $\geq 2.14.0$ ), **parallel** is in R-core. If only **snow** is available, it is recommended to library it before using the parallel computing functionality.

See the help page of `singleCAM()` for more examples.

### Multiple LD Decay Curves (.rawld File)

The function `CAM()` does CAM for a .rawld file with multiple LD decay curve. Parallel computation is also supported. For example, let's use the *GA* data set ((the admixture proportion for population 1 ( $m_1$ ) is 0.3)

with the most ancient generation concerned being 150 (T=150L) and core models being HI, CGF1-I, CGF2-I and GA-I (isolation=TRUE by default), without using parallel computation for the four models for each LD decay curve (single.parallel=FALSE):

```
data(GA_I)
fit<-CAM(rawld=GA_I,m1=0.3,T=150L,LD.parallel=TRUE,single.parallel=FALSE)
#Usually, one only needs to pass the paths to the .rawld file and to the .log file to CAM():
#fit<-CAM(rawld="path/to/GA_I.rawld",m1="path/to/GA_I.log",T=150L,LD.parallel=TRUE,single.parallel=FALSE)
fit
```

## Continuous Admixture Inference (CAM) for a .rawlf File

##

## Function call:CAM(rawld = GA\_I, m1 = 0.3, T = 150L, LD.parallel = TRUE, single.parallel = FALSE)

##

## Total Length of LD: 3497

##

##	LD	Model	Start	End	msE	quasi.F
##	Combined_LD	HI	62	62	3.269912e-06	1.448423
##	Combined_LD	CGF1-I	110	18	2.350439e-06	1.041138
##	Combined_LD	CGF2-I	121	22	2.313658e-06	1.024846
##	Combined_LD	GA-I	101	26	2.290190e-06	1.014450
##	Jack1	HI	62	62	3.331873e-06	NA
##	Jack1	CGF1-I	109	19	2.532073e-06	NA
##	Jack1	CGF2-I	119	23	2.498988e-06	NA
##	Jack1	GA-I	102	26	2.487870e-06	NA
##	Jack2	HI	61	61	3.227448e-06	NA
##	Jack2	CGF1-I	110	18	2.361470e-06	NA
##	Jack2	CGF2-I	118	23	2.310203e-06	NA
##	Jack2	GA-I	101	26	2.299594e-06	NA
##	Jack3	HI	61	61	3.477126e-06	NA
##	Jack3	CGF1-I	111	17	2.457821e-06	NA
##	Jack3	CGF2-I	122	21	2.418382e-06	NA
##	Jack3	GA-I	102	25	2.392573e-06	NA
##	Jack4	HI	62	62	3.363831e-06	NA
##	Jack4	CGF1-I	112	17	2.387836e-06	NA
##	Jack4	CGF2-I	121	22	2.348988e-06	NA
##	Jack4	GA-I	104	25	2.343282e-06	NA
##	Jack5	HI	62	62	3.411285e-06	NA
##	Jack5	CGF1-I	109	19	2.402213e-06	NA
##	Jack5	CGF2-I	124	21	2.390199e-06	NA
##	Jack5	GA-I	104	25	2.357778e-06	NA
##	Jack6	HI	62	62	3.289421e-06	NA
##	Jack6	CGF1-I	108	19	2.457056e-06	NA
##	Jack6	CGF2-I	121	22	2.440099e-06	NA
##	Jack6	GA-I	100	27	2.402203e-06	NA
##	Jack7	HI	62	62	3.428745e-06	NA
##	Jack7	CGF1-I	110	18	2.429383e-06	NA
##	Jack7	CGF2-I	121	22	2.406796e-06	NA
##	Jack7	GA-I	103	25	2.398758e-06	NA
##	Jack8	HI	62	62	3.283614e-06	NA
##	Jack8	CGF1-I	107	20	2.527757e-06	NA
##	Jack8	CGF2-I	119	23	2.491798e-06	NA
##	Jack8	GA-I	102	26	2.471097e-06	NA
##	Jack9	HI	61	61	3.479743e-06	NA

##	Jack9	CGF1-I	111	17	2.418923e-06	NA
##	Jack9	CGF2-I	122	21	2.380390e-06	NA
##	Jack9	GA-I	105	24	2.367834e-06	NA
##	Jack10	HI	61	61	3.253384e-06	NA
##	Jack10	CGF1-I	108	19	2.330627e-06	NA
##	Jack10	CGF2-I	118	23	2.291267e-06	NA
##	Jack10	GA-I	103	25	2.289236e-06	NA

One can also specify the file path of the `.rawld` file in argument `rawld=` and the file path of the `.log` file containing the information of `m1` in argument `m1=`.

Here the class of `fit` is **CAM**, and it has its own method for `print()` and `plot()`. `fit$summary` is a more comprehensive data frame containing the data frame printed. A **CAM** object has an element named `CAM.list` consisting of the **CAM.single** objects for each LD decay curve.

Parallel computation is also supported as in the example, provided that **parallel** package or **snow** package is installed. For newer versions of R ( $\geq 2.14.0$ ), **parallel** is in R-core. If only **snow** is available, it is recommended to library it before using the parallel computing functionality.

See help page of `CAM()` for more examples and details.

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## Summary Plots

A new method of `plot()` for **CAM** class is introduced in this package (`plot.CAM()`). This function generates three plots in a device. The plot on the top left is the estimated time intervals/points for the four models. The color depth of segments/points corresponds to how many intervals/points covers this part in Jackknives. The deeper the color, the more estimates from Jackknives cover this part. The plot on the top right is the boxplot of `msE` for the four models. The third plot shows the fitting of four models to `Combined_LD` in the `.rawld` file. The numbers after model names in the legend are quasi-F values of the four models for `Combined_LD`. For example, let's plot the previous result:

```
plot(fit)
```

One can also run `plot(fit,"GA_I.pdf")` to plot to a `.pdf` file, which is recommended.

To change the colors of models, one can pass a  $3 \times 4$  matrix of colors:

```
plot(fit,model.cols=matrix(c("pink","red","pink",
                             "lightseagreen","green","green",
                             "skyblue","blue","blue",
                             "yellow","orange","orange"),ncol=4))
```

See help page of `plot.CAM()` for more details.

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## Draw Conclusions on Best Model(s)

The function `conclude.model()` can draw conclusions on which models are the significantly best ones and find their estimated time intervals/points. It takes a "CAM" class object or its summary table as input. For example, let's find out the best model(s) from the previous CAM analysis:

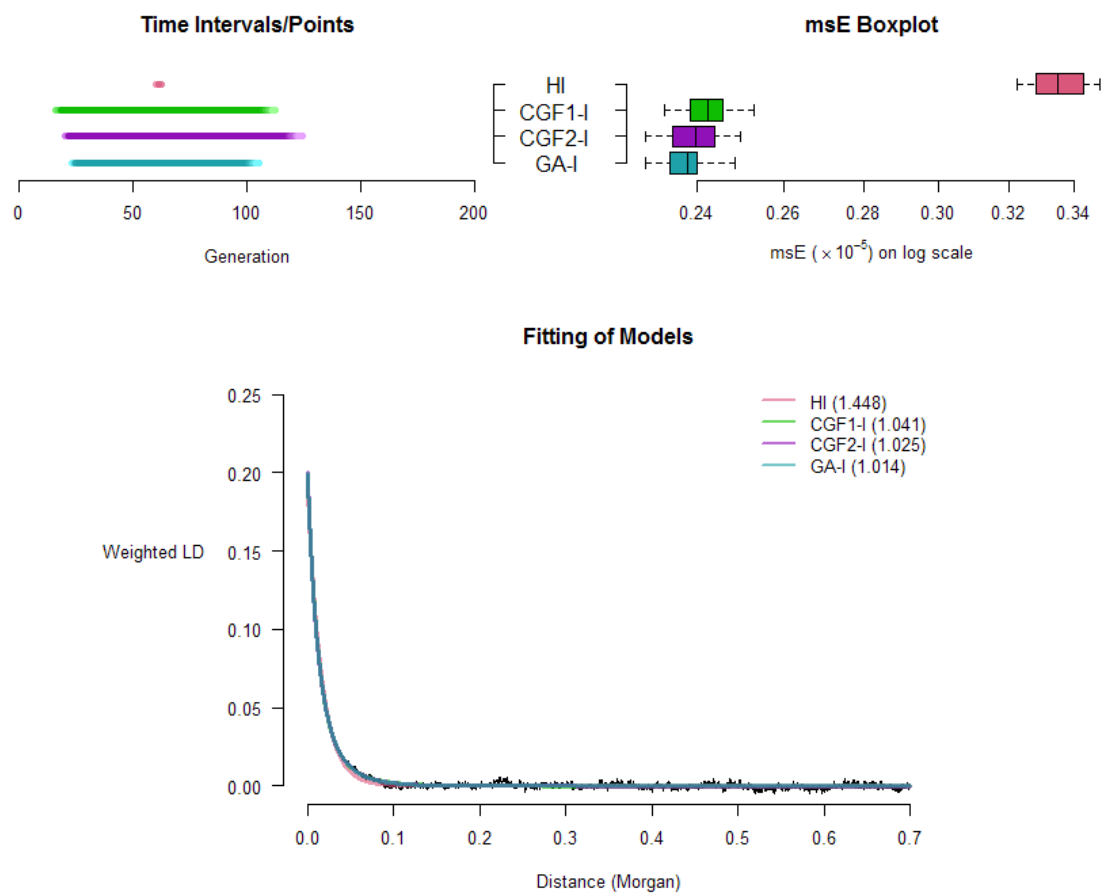


Figure 1:

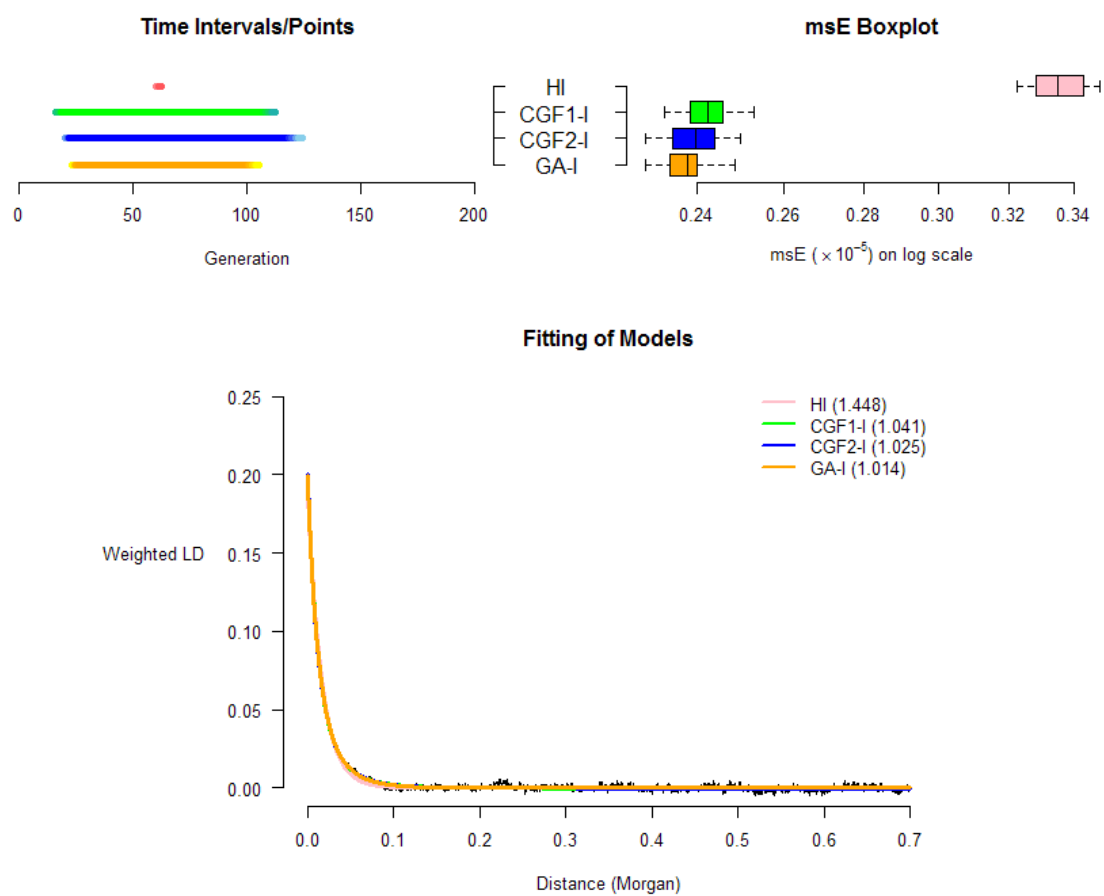


Figure 2:

```

conclusion<-conclude.model(fit)
conclusion<-conclude.model(fit$summary)
conclusion

## CAM Best Model(s) Conclusion:
##
## Function call: conclude.model(x = fit$summary)
##
## Familiwise Error Rate: 0.05
##
## Best Model(s) and Time Estimation:
## Best.Models End Start
##      CGF2-I  22   121
##      GA-I   25   102
##
## Group Medians of pseudo log(msE)/msE:
##      HI      CGF1-I      CGF2-I      GA-I
## -12.84265 -13.23881 -13.30074 -13.33372
##
## Adjusted p-value:
##      HI      CGF1-I      CGF2-I      GA-I
## HI      NA 0.01171875 0.01171875 0.01171875
## CGF1-I 0.01171875      NA 0.05468750 0.01171875
## CGF2-I 0.01171875 0.05468750      NA 0.05468750
## GA-I   0.01171875 0.01171875 0.05468750      NA

```

The function returns an object of **CAM.conclusion** class, which has a special method for `print()`.

Note that this function only selects the significantly best model(s), i.e. the one(s) that are significantly the closest to what is observed. It does **NOT** check if the best model(s) are credible or not. The user should check the quasi-F value and msE in the summary table or plot of a “CAM” class object for this purpose.

See the help page of `conclude.model()` for further information.

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## Miscellany

### Construct a Simple CAM object

Sometimes maybe only the summary table of an object of **CAM** class is saved. The function `construct.CAM()` can construct a simple **CAM** object given the original .rawld file, the summary table of the original **CAM** object and the admixture proportion of population 1  $m_1$ , which can be passed to `plot.CAM()` function and `conclude.model()` function. For example, let's “save” the summary table of the previous result (`fit$summary`), then use this function to construct a **CAM** class object and do some further analysis from it:

```

summarydata<-fit$summary
rm(fit)
fit<-construct.CAM(rawld=GA_I,m1=0.3,dataset=summarydata)
fit

## Continuous Admixture Inference (CAM) for a .rawlf File
##

```

```

## Total Length of LD: 3497
##
##      LD  Model Start End      msE  quasi.F
## Combined_LD  HI      62  62 3.269912e-06 1.448423
## Combined_LD CGF1-I    110  18 2.350439e-06 1.041138
## Combined_LD CGF2-I    121  22 2.313658e-06 1.024846
## Combined_LD  GA-I    101  26 2.290190e-06 1.014450
##      Jack1    HI      62  62 3.331873e-06      NA
##      Jack1 CGF1-I    109  19 2.532073e-06      NA
##      Jack1 CGF2-I    119  23 2.498988e-06      NA
##      Jack1  GA-I    102  26 2.487870e-06      NA
##      Jack2    HI      61  61 3.227448e-06      NA
##      Jack2 CGF1-I    110  18 2.361470e-06      NA
##      Jack2 CGF2-I    118  23 2.310203e-06      NA
##      Jack2  GA-I    101  26 2.299594e-06      NA
##      Jack3    HI      61  61 3.477126e-06      NA
##      Jack3 CGF1-I    111  17 2.457821e-06      NA
##      Jack3 CGF2-I    122  21 2.418382e-06      NA
##      Jack3  GA-I    102  25 2.392573e-06      NA
##      Jack4    HI      62  62 3.363831e-06      NA
##      Jack4 CGF1-I    112  17 2.387836e-06      NA
##      Jack4 CGF2-I    121  22 2.348988e-06      NA
##      Jack4  GA-I    104  25 2.343282e-06      NA
##      Jack5    HI      62  62 3.411285e-06      NA
##      Jack5 CGF1-I    109  19 2.402213e-06      NA
##      Jack5 CGF2-I    124  21 2.390199e-06      NA
##      Jack5  GA-I    104  25 2.357778e-06      NA
##      Jack6    HI      62  62 3.289421e-06      NA
##      Jack6 CGF1-I    108  19 2.457056e-06      NA
##      Jack6 CGF2-I    121  22 2.440099e-06      NA
##      Jack6  GA-I    100  27 2.402203e-06      NA
##      Jack7    HI      62  62 3.428745e-06      NA
##      Jack7 CGF1-I    110  18 2.429383e-06      NA
##      Jack7 CGF2-I    121  22 2.406796e-06      NA
##      Jack7  GA-I    103  25 2.398758e-06      NA
##      Jack8    HI      62  62 3.283614e-06      NA
##      Jack8 CGF1-I    107  20 2.527757e-06      NA
##      Jack8 CGF2-I    119  23 2.491798e-06      NA
##      Jack8  GA-I    102  26 2.471097e-06      NA
##      Jack9    HI      61  61 3.479743e-06      NA
##      Jack9 CGF1-I    111  17 2.418923e-06      NA
##      Jack9 CGF2-I    122  21 2.380390e-06      NA
##      Jack9  GA-I    105  24 2.367834e-06      NA
##      Jack10   HI      61  61 3.253384e-06      NA
##      Jack10 CGF1-I    108  19 2.330627e-06      NA
##      Jack10 CGF2-I    118  23 2.291267e-06      NA
##      Jack10  GA-I    103  25 2.289236e-06      NA

```

```
plot(fit)
```

```
conclude.model(fit)
```

```
## CAM Best Model(s) Conclusion:
```

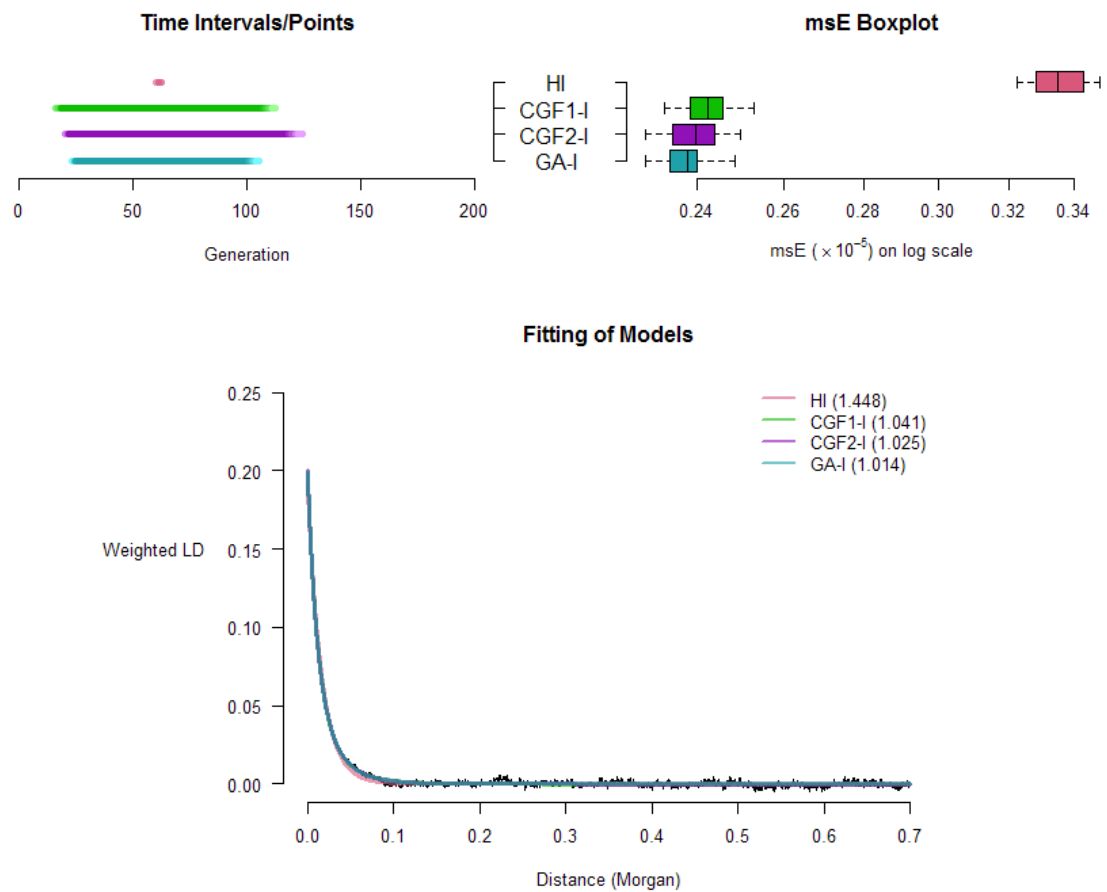


Figure 3:



```
##
## Function call: conclude.model(x = fit)
##
## Familiwise Error Rate: 0.05
##
## Best Model(s) and Time Estimation:
## Best.Models End Start
##      CGF2-I  22   121
##      GA-I   25   102
##
## Group Medians of pseudo log(msE)/msE:
##      HI      CGF1-I      CGF2-I      GA-I
## -12.84265 -13.23881 -13.30074 -13.33372
##
## Adjusted p-value:
##      HI      CGF1-I      CGF2-I      GA-I
## HI      NA 0.01171875 0.01171875 0.01171875
## CGF1-I 0.01171875      NA 0.05468750 0.01171875
## CGF2-I 0.01171875 0.05468750      NA 0.05468750
## GA-I   0.01171875 0.01171875 0.05468750      NA
```

## Reconstruct Fitted LD Decay Curves

One may want to get the fitted LD decay curves. The function `reconstruct.fitted()` takes a **CAM.single** class object and returns a list containing the best-fit curves for the four models. It can take the **CAM.single** class objects in the constructed a **CAM** class object from `construct.CAM()` as input. For example, let's use the **CAM** class object just constructed and reconstruct the fitted curves:

```
fitted<-reconstruct.fitted(fit$CAM.list[[1]])
str(fitted)

## List of 4
## $ HI.fitted      : num [1:3497] 0.191 0.189 0.187 0.184 0.182 ...
## $ CGF1-I.fitted: num [1:3497] 0.2 0.197 0.194 0.191 0.188 ...
## $ CGF2-I.fitted: num [1:3497] 0.2 0.198 0.195 0.192 0.189 ...
## $ GA-I.fitted   : num [1:3497] 0.199 0.197 0.194 0.191 0.188 ...
```

## HI Modle for Single LD Decay Curve

The function `singleHI()` does time inference, of HI model only, for a single LD decay curve. The algorithm is the same as the HI model part of `singleCAM()`. For example, let's use the Combined LD in the *CGF\_50* data set and use only HI as the core model:

```
fit<-singleHI(d,Z,m1=0.3,T=70L)
fit

## Continuous Admixture Inference (CAM) for a Single LD Decay Curve
##
## Function call: singleHI(d = d, Z = Z, m1 = 0.3, T = 70L)
##
## Length of Used LD: 3497
##
```

```
## Model Start End          msE
##   HI      23  23 8.912686e-06
```

This function also returns an object of **CAM.single** class, and can be passed to `reconstruct.fitted()`:

```
fitted<-reconstruct.fitted(fit)
str(fitted)
```

```
## List of 1
## $ HI.fitted: num [1:3497] 0.195 0.194 0.193 0.193 0.192 ...
```

It is recommended to use this function when only HI model is concerned. See the help page of `singleHI()` for further details.

### HI Model for Multiple LD Decay Curves (.rawld File)

The function `HI()` does time inference, of HI model only, for a .rawld file. The algorithm is the same as the HI model part of `CAM()`. For example, let's again use the *GA\_I* data set with the most ancient generation concerned being 150 (T=150L), but this time only HI is the core model:

```
fit<-HI(GA_I,m1=.3,T=150L)
fit
```

```
## Continuous Admixture Inference (CAM) for a .rawlf File
##
## Function call:HI(rawld = GA_I, m1 = 0.3, T = 150L)
##
## Total Length of LD: 3497
##
##          LD Model Start End          msE  quasi.F
## Combined_LD   HI      62  62 3.269912e-06 1.448423
##      Jack1     HI      62  62 3.331873e-06      NA
##      Jack2     HI      61  61 3.227448e-06      NA
##      Jack3     HI      61  61 3.477126e-06      NA
##      Jack4     HI      62  62 3.363831e-06      NA
##      Jack5     HI      62  62 3.411285e-06      NA
##      Jack6     HI      62  62 3.289421e-06      NA
##      Jack7     HI      62  62 3.428745e-06      NA
##      Jack8     HI      62  62 3.283614e-06      NA
##      Jack9     HI      61  61 3.479743e-06      NA
##      Jack10    HI      61  61 3.253384e-06      NA
```

The output is also an object of **CAM** class. However, it should *NOT* be passed to `plot()`, and its summary table should *NOT* be passed to `construct.CAM()`.

It is recommended to use this function when only HI model is concerned. See the help page of `HI()` for further details.