$DOVE - \underline{D}urability \ \underline{O}f \ \underline{V}accine \ \underline{E}fficacy$

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Introduction

DOVE is an R package for evaluating the durability of vaccine efficacy (VE) in a randomized, placebocontrolled clinical trial with staggered enrollment of participants and potential crossover of placebo recipients (Lin et al., 2021a; 2021b). It inputs a rectangular dataset with the following information:

- Entry time: Calendar time when the participant enters the trial.
- Event time: Calendar time when the participant experiences the clinical event of interest (e.g., symptomatic COVID-19) or their follow-up ends, whichever occurs first.
- Event status: Binary indicator taking the value 1 if the clinical event of interest occurs before the end of follow-up and 0 otherwise.
- Vaccination status: Binary indicator taking the value 1 if vaccination occurs before the end of follow-up and 0 otherwise.
- Vaccination time: Calendar time when vaccination takes place, with an arbitrary non-negative value if the participant is not vaccinated.
- Covariates: Baseline covariates (e.g., priority group, age, sex, ethnicity).

Of note, an arbitrary number of baseline covariates can be included, and all time variables are measured from the start of the trial and are specified in units of days.

The two primary analysis tools of the package are dove() and dove2(), the formal argument structures of which are similar and were chosen to resemble that of the coxph() function of the **survival** package. The underlying methodologies for dove() and dove2() are detailed in Lin et al. (2021a) and Lin et al. (2021b), respectively. Function dove() allows the vaccine effect to be an arbitrary function of time, whereas function dove2() assumes that the log hazard ratio for the vaccine effect is a piecewise linear function of time. Both functions return the estimated hazard ratio for each baseline covariate, the estimated VE in reducing the attack rate (cumulative incidence), the estimated VE in reducing the hazard rate (instantaneous risk), and the estimated VE in reducing the attack rate over successive time periods.

We recommend *dove()* for exploratory analyses and *dove2()* for formal analyses. The latter yields more stable estimates, together with proper confidence intervals, for VE in reducing the hazard rate. Both functions handle potentially right-censored events (e.g., symptomatic COVID-19, severe COVID-19, death). For interval-censored infection endpoints, iDOVE should be used instead.

The DOVE package includes convenience functions print(), plot(), and vaccine(). Function vaccine() is used to simplify the specification of input variables required in the model statements of dove() and dove2(), similar in spirit to the cluster() function of the **survival** package. A simulated dataset is provided to illustrate the use of the software.

Functions

vaccine()

This convenience function is used as a component of the right-hand side of a formula object for the sole purpose of simplifying the specification of required input variables: entry time, vaccination status, and vaccination time. This function is not intended to be used as a stand-alone feature; although for completeness, the function ensures that the input data obey basic constraints and returns the data in a predictable format for use in internal functions.

The usage is

vaccine(entry_time, vaccination_status, vaccination_time)

where entry_time is the time when the participant enters the trial, vaccination_status is the binary indicator of vaccination, and vaccination_time is the time when vaccination takes place.

dove()

This function estimates VE as a nonparametric function of time. The value object returned contains the estimated hazard ratio for each baseline covariate, estimated VE in reducing the attack rate, $VE_a(t)$, and in reducing the hazard rate, $VE_h(t)$, where t is the time elapsed since vaccination, as well as the estimated VE in reducing the attack rate over m successive time periods, $VE_a(0, t_1), VE_a(t_1, t_2), \ldots, VE_a(t_{m-1}, t_m)$. By definition, $VE_a(0, t) = VE_a(t)$.

The function call takes the following form:

dove(formula, data, plots = TRUE, timePts = NULL, bandwidth = NULL)

where

- formula is a model statement. See below for further details.
- data is the data.frame object containing all required data as previously described.
- plots is a logical object indicating whether graphical forms of the $VE_a(t)$ and $VE_h(t)$ results are to be generated.
- timePts is an optional vector object specifying the time points (t_1, t_2, \ldots, t_m) for partitioning the study period.
- bandwidth is an optional numeric object specifying the tuning parameter for the bandwidth used in the kernel estimation of $VE_h(t)$.

To obtain reliable estimates of VE_a (t_{j-1}, t_j) (j = 1, ..., m), we suggest using broad time periods, such as every month, every two months, or every quarter. If timePts is not provided, the default time periods are every 60 days. (If $\tau < 60$ days, timePts must be provided.) We suggest choosing bandwidth between 0.1 and 1.0: a smaller bandwidth yields a less biased estimate of $VE_h(t)$, whereas a larger bandwidth yields a smoother estimate of the VE_h curve. The default value of bandwidth is 0.3. This input is ignored if plots is FALSE.

The model statement is a formula object. The left-hand side is a survival analysis object as returned by the *Surv()* function of the **survival** package and specifies the event time and event status. The right-hand side is a combination of baseline covariates and the previously described *vaccine()* function. Categorical baseline covariates can be specified, and if provided, all other categories are compared to the first category. The **formula** input takes the following general structure

where 'event_time', 'event_status', 'covariates', 'entry_time' 'vaccination_status' and 'vaccination_time' are place holders indicating the data that are to be provided; they will be replaced by the variable names in the header of the input data.

The two VE measures, $VE_a(t)$ and $VE_h(t)$, are estimated up to the last observed event time. However, the estimates near the end of crossover where there are very few placebo participants under follow-up may not be reliable. We also constrain VE_h to be 0 at day 0 and non-decreasing at the right tail. For estimating VE_a over successive time periods, the last time period should not extend beyond the point that there are still a few placebo participants under follow-up.

dove2()

This function estimates VE under the assumption that the log hazard ratio for the vaccine effect is a piecewise linear function of time. The value object returned is similar to that returned by dove() and contains the estimated hazard ratio for each baseline covariate, estimated VE in reducing the attack rate, $VE_a(t)$, and in reducing the hazard rate, $VE_h(t)$, where t is the time elapsed since vaccination, as well as the estimated VE in reducing the attack rates over m successive time periods, $VE_a(0, t_1), VE_a(t_1, t_2), \ldots, VE_a(t_{m-1}, t_m)$. The 95% confidence intervals for all three measures of VE are provided.

The function call takes the following form:

where formula, data, plots, and timePts are as described above for *dove()*. Input changePts is an optional numerical vector to specify the change points, in units of days, of the piece-wise log-linear hazard ratio for the vaccine effect. If no change points are provided, one change point will automatically be selected among Weeks 4, 5, 6, 7, 8 by the Akaike information criterion (AIC). Input constantVE is a logical object indicating the VE trend after the last change point. If specified as TRUE, VE is assumed to be constant in the period after the last change point; otherwise it is allowed to vary after the last change point. If timePts is not specified, the default sequence of the multiples of the first change point will be used.

The model statement is as previously described for dove(). Specifically, the formula input takes the following general structure

To ensure stability, we suggest placing change points at the time points where the events are relatively frequent and not placing change points near the right tail. We estimate $VE_a(t)$ and $VE_h(t)$ up to the maximum of all observed event times.

plot()

When provided a value object of class DOVE (the object returned by dove() and dove2(), this convenience function creates/recreates plots of the estimated VE in reducing the attack rate, $VE_a(t)$, and in reducing the hazard rate, $VE_h(t)$.

print()

When provided a value object of class DOVE (the object returned by dove() and dove2()), the tabular results are displayed.

Examples

We use the dataset provided with the package, doveData, to illustrate the analyses. This dataset was simulated under a priority-tier dependent crossover design with a ramping vaccine effect between dose 1 and dose 2 and contains the following observations for each of the 40,000 participants:

- entry.time: The entry time in days.
- event.time: The event time in days.
- event.status: The event indicator (1=observed; 0=censored).
- vaccine.time: The time of vaccination in days; NA if not vaccinated.
- vaccine.status: The indicator of vaccination (1=vaccinated; 0 = not vaccinated).
- priority: A composite baseline risk score taking values 1-5.
- sex: A binary indicator of sex (male/female).

The data can be loaded in the usual way

data(doveData)

head(doveData)

##	entry.time	event.time	event.status	vaccine.time	vaccine.status	priority	sex
## 1	69	212	1	NA	0	4	1
## 2	92	320	0	250	1	3	0
## 3	21	137	1	NA	0	3	0
## 4	31	320	0	186	1	5	1
## 5	32	320	0	251	1	3	0
## 6	72	320	0	188	1	5	1

The summary statistics are shown below

```
summary(doveData)
```

##	entry.time	event.time	event.status	vaccine.time
##	Min. : 0.00	Min. : 9.0	Min. :0.00000	Min. : 0.0
##	1st Qu.: 31.00	1st Qu.:320.0	1st Qu.:0.00000	1st Qu.: 57.0
##	Median : 60.00	Median :320.0	Median :0.00000	Median :112.0
##	Mean : 60.25	Mean :310.8	Mean :0.06657	Mean :148.1
##	3rd Qu.: 90.00	3rd Qu.:320.0	3rd Qu.:0.00000	3rd Qu.:247.0
##	Max. :120.00	Max. :320.0	Max. :1.00000	Max. :319.0
##				NA's :3234
##	vaccine.status	priority	sex	
##	Min. :0.0000	Min. :1.000	Min. :0.0000	
##	1st Qu.:1.0000	1st Qu.:2.000	1st Qu.:0.0000	
##	Median :1.0000	Median :3.000	Median :1.0000	
##	Mean :0.9192	Mean :3.005	Mean :0.5035	
##	3rd Qu.:1.0000	3rd Qu.:4.000	3rd Qu.:1.0000	
##	Max. :1.0000	Max. :5.000	Max. :1.0000	

##

We see that participants were enrolled in the study over a 4-month period ($0 \le \text{entry.time} \le 120 \text{ days}$), the follow-up ended on day 320 (event.time $\le 320 \text{ days}$) with a $\sim 6.7\%$ event rate, and $\sim 92\%$ of the participants were vaccinated over the course of the study period. In addition, the priority (risk) score is evenly distributed across participants, who are equally distributed between the two sex groups.

dove()

First, we illustrate a dove() analysis. Here, we will include in our model statement baseline covariates, priority and sex. In addition, we will use the default partitioning of the study period and the default tuning parameter for the bandwidth. The function call takes the following form

tau = 320

timePts not given; default values will be used

method converged after 13 iterations

The function returns an object of class DOVE containing a list with the following elements. For brevity, we show only a snapshot of the large tabular results.

call: The unevaluated call.

Covariate Effects: The estimated hazard ratio for each covariate, together with the (estimated) standard error, the 95% confidence interval, and the two-sided p-value for testing no covariate effect.

result1\$covariates

##	coef	se(coef)	z	Pr(z)	exp(coef)	lower .95	upper .95
## priority	0.2139999	0.01444336	14.816485	0	1.238622	1.204050	1.274188
## sex	0.3645546	0.03940875	9.250601	0	1.439873	1.332842	1.555498

Vaccine Efficacy: Element **\$efficacy** contains the estimated VE in reducing the attack rate at each observed event time, together with its standard error and the 95% confidence interval. In addition, the raw estimate of the hazard ratio at each observed event time is provided.

```
head(result1$vaccine$efficacy)
```

##	time	VE_a	se	lower .95	upper .95	hazardRatio
## [1	L,] O	0.000000	0.000000	0.0000000	0.000000	1.0000000
## [2	2,] 1	0.3529937	0.2056809	-0.20644423	0.6530157	0.6470063
## [3	3,] 2	0.2547967	0.1569215	-0.12595279	0.5067929	0.8434002
## [4	1,] 3	0.1559060	0.1377541	-0.16227101	0.3869806	1.0418753
## [5	5,] 4	0.2198001	0.1153130	-0.04235380	0.4160218	0.5885176
## [6	5,] 5	0.2577284	0.1011185	0.03055578	0.4316670	0.5905584

```
tail(result1$vaccine$efficacy)
```

##timeVE_aselower.95upper.95hazardRatio##[284,]2980.56143140.025870780.50767710.60931660.9504455##[285,]2990.55958330.026203950.50510900.60806140.9911702##[286,]3010.55882410.026502510.50369750.60782751.1093593

 ## [287,]
 304 0.5588767 0.02682445 0.5030393 0.6084403
 1.3075374

 ## [288,]
 306 0.5569388 0.02734465 0.4999669 0.6074196
 1.4752206

 ## [289,]
 307 0.5532190 0.02807463 0.4946606 0.6049916
 1.5850495

Element **\$period_efficacy** contains the estimated VE in reducing the attack rate over each time period, its standard error, and the 95% confidence interval.

```
result1$vaccine$period_efficacy
```

left right VE_a lower .95 upper .95 se ## [1,] 0 60 0.6821729 0.02126667 0.63763337 0.7212379 [2,] 120 0.7239177 0.02133139 0.67877631 0.7627153 ## 60 [3,] 120 180 0.6755593 0.02529073 0.62200229 0.7215281 ## [4,] 180 240 0.4436504 0.04417056 0.34997670 0.5238249 ## 300 0.2799564 0.09087456 0.07787798 0.4377503 ## [5,] 240

The graphical depictions of estimates returned in **vaccine**sefficacy are generated by default by dove() and are shown in Figure 1. This figure can be regenerated using plot() as follows:

```
plot(x = result1)
```



Figure 1: Plots auto-generated by dove(). On the left, the estimated VE curve in reducing the attack rate, $VE_a(t)$ (black) and its 95% confidence intervals (green) are shown as a function of the time since vaccination. On the right, the estimated VE curve in reducing the hazard ratio, $VE_h(t)$, is shown as a function of the time since vaccination.

dove2()

In the first analysis illustrating dove2(), we set Week 4 as the change point and assume a potentially waning VE after 4 weeks. We estimate VE_a over 0-4, 4-16, 16-28, 28-40 weeks. Note that all times must be provided in the unit of days. The function call takes the following form

```
changePts = 4*7,
timePts = c(4, 16, 28, 40)*7)
```

tau = 320

```
## Number of subjects: 40000
```

```
## log partial-likelihood at final estimates: -27351.81
```

The function returns an S3 object of class DOVE, which contains a list object with the following information.

call: The unevaluated call.

result2\$call

```
## dove2(formula = Surv(event.time, event.status) ~ priority + sex +
## vaccine(entry.time, vaccine.status, vaccine.time), data = doveData,
## changePts = 4 * 7, timePts = c(4, 16, 28, 40) * 7)
```

changePts: The changePts of the analysis.

result2\$changePts

[1] 28

Covariate Effects: The estimated (log) hazard ratio of each covariate, together with the estimated standard error, the 95% confidence interval, and the two-sided p-value for testing no covariate effect.

result2\$covariates

```
## coef se(coef) z Pr(>|z|) exp(coef) lower .95
## priority 0.2145486 0.01438123 14.918649 2.492693e-50 1.239302 1.204858
## sex 0.3646241 0.03940478 9.253296 2.176748e-20 1.439973 1.332945
## upper .95
## priority 1.274732
## sex 1.555594
```

When no baseline covariates are provided, this element will be NA.

Vaccine Efficacy: Element \$VE_a contains the daily VE estimate in reducing the attack rate, together with its standard error and the 95% confidence interval. Element \$VE_h contains the daily VE estimate in reducing the hazard rate, together with its standard error and the 95% confidence interval.

head(result2\$vaccine\$VE_a)

```
##
                             se lower .95 upper .95
       time
                 VE_a
         ## [1,]
## [2,]
         1 0.03024399 0.001222094 0.02784573 0.03263634
## [3,]
         2 0.05927458 0.002346379 0.05466442 0.06386226
         3 0.08714643 0.003379452 0.08049861 0.09374618
## [4,]
## [5,]
         4 0.11391156 0.004327491 0.10538895 0.12235298
         5 0.13961952 0.005196276 0.12937430 0.14974418
## [6,]
```

tail(result2\$vaccine\$VE_a)

time VE_a se lower .95 upper .95
[316,] 315 0.5385651 0.02499808 0.4868731 0.5850498
[317,] 316 0.5370153 0.02518384 0.4849277 0.5838355

[318,] 317 0.5354577 0.02537152 0.4829703 0.5826167
[319,] 318 0.5338922 0.02556114 0.4810008 0.5813935
[320,] 319 0.5323189 0.02575270 0.4790191 0.5801658
[321,] 320 0.5307377 0.02594622 0.4770253 0.5789336

head(result2\$vaccine\$VE_h)

```
##
       time
                VE_h
                             se lower .95 upper .95
## [1,]
         1 0.05987195 0.002394147 0.05516769 0.06455278
## [2,]
## [3,]
         2 0.11615925 0.004501610 0.10729190 0.12493851
## [4,]
         3 0.16907651 0.006348134 0.15654055 0.18142616
## [5,]
         4 0.21882552 0.007957412 0.20307225 0.23426738
## [6,]
         5 0.26559596 0.009351233 0.24703692 0.28369756
```

tail(result2\$vaccine\$VE_h)

##		time	VE_h	se	lower .95	upper .95
##	[316,]	315	0.05159091	0.09347856	-0.1505214	0.2181981
##	[317,]	316	0.04603717	0.09443096	-0.1582220	0.2142741
##	[318,]	317	0.04045090	0.09539185	-0.1659755	0.2103312
##	[319,]	318	0.03483192	0.09636127	-0.1737821	0.2063694
##	[320,]	319	0.02918004	0.09733931	-0.1816423	0.2023886
##	[321,]	320	0.02349506	0.09832601	-0.1895563	0.1983886

Element **\$VE_period** contains the estimated VE in reducing the attack rate over each time period, its standard error, and the 95% confidence interval.

result2\$vaccine\$VE_period

left right se lower .95 upper .95 VE_a ## [1,] 0 28 0.5242172 0.01230293 0.4994819 0.5477300 ## [2,] 28 112 0.7708700 0.01303526 0.7438421 0.7950462 ## [3,] 112 196 0.6258165 0.01823842 0.5883060 0.6599094 ## [4,] 196 280 0.3889351 0.04248634 0.2997210 0.4667834

The graphical depictions of VE_a and VE_h estimates are generated by default by dove2() and are shown in Figure 2. This figure can be regenerated using plot() as follows:

plot(x = result2)

In the final analysis, we have the software use AIC to choose a change point among Weeks 4, 5, 6, 7, 8. We assume a constant VE after the change point, and thus only the constant VE is estimated. The function call takes the following form



Figure 2: Plots auto-generated by dove2(). On the left, the estimated VE curve in reducing the attack rate, $VE_a(t)$ (black) and its 95% confidence intervals (green) are shown as a function of the time since vaccination. On the right, the estimated VE curve in reducing the hazard ratio, $VE_h(t)$ (black) and its 95% confidence intervals (green) are shown as a function of the time since vaccination.

The function returns a list object containing the following items.

changePts: The change point selected.
result3\$changePts

[1] 28

Covariate Effects: The estimated (log) hazard ratio of each covariate, together with the estimated standard error, the 95% confidence interval, and the two-sided p-value for testing no covariate effect.

result3\$covariates

coef se(coef) Pr(|z|) exp(coef) lower .95 z ## priority 0.2127205 0.01450545 14.664868 1.082278e-48 1.237039 1.202364 ## sex 0.3652731 0.03940404 9.269939 1.862523e-20 1.440907 1.333812 upper .95 ## 1.272713 ## priority 1.556602 ## sex

Vaccine Efficacy: Element \mathbf{SVE} contains the estimated constant VE, together with its standard error and the 95% confidence interval.

result3\$vaccine\$VE

VE se lower .95 upper .95
0.68233226 0.01443303 0.65274570 0.70939801

Plots cannot be generated when constantVE = TRUE.

plot(x = result3)

plot() is not available for the provided analysis (constantVE = TRUE)

References

Lin DY, Zeng D, Gilbert PB (2021a). Evaluating the long-term efficacy of COVID-19 vaccines. Clinical Infectious Diseases, ciab226, https://doi.org/10.1093/cid/ciab226

Lin, D-Y, Gu, Y., Zeng, D., Janes, H. E., and Gilbert, P. B. (2021b). Evaluating Vaccine Efficacy Against SARS-CoV-2 Infection. Clinical Infectious Diseases, ciab630, https://doi.org/10.1093/cid/ciab630