# Simulation and plotting code for "Optimizing the use of carbonate standards to minimize uncertainties in clumped isotope data"

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#### Contents

1	install required packages	3
2	optional packages	4
3	load libraries	4
4	multiple cores	4
5	set up plotting theme	4
6	set up stdinfo etc. 6.1 setup axes	<b>5</b>
7	calculate temperature sensitivity as a function of temperature 7.1 temp_sens_pot	5
8	create standard intro plot	8
9	micro benchmark	11
10	example sims	11
11	stddis 11.1 create a list of all possible stddis combinations	14 14 15

	11.3	stddis runall
	11.4	arrange results for plot
	11.5	create barcharts for proportion axes 18
		stddis_pl
		prep text
10	.4.1	0.0
12		rssmp 22
		setup
		run sims
		save results
		tidy it up
		stdvssmp_pl 2
	12.6	supplementary figure plot with all input standard deviations . $20$
13	prop	o-eth3 28
		prop-eth3 for continuous sample range
		prop-eth3 expand experimental matrices and run simulations 29
		13.2.1 run many sims
		13.2.2 save results
	13.3	smp_out_uu
		best_dat
		smp_out_comb
		prop eth3 pl
		13.6.1 plot_best
		13.6.2 plot_prop
		13.6.3 combine and print
		13.6.4 3d rayshader plot
	13.7	best_prop_diff_pl
	13.8	calculate some summary statistics for use in-text
11	pror	o-eth3 with a very very very cold and hot standard?
		prop-eth3 for continuous sample range
		propeth sepand experimental matrices and run simulations 39
		run many sims
	11.0	14.3.1 save results
	14 4	brrr_best_dat
		smp out comb
		prop eth3 pl
	11.0	14.6.1 plot_best
		14.6.2 plot prop

The supplementary code to the manuscript "Optimizing the use of carbonate standards to minimize uncertainties in clumped isotope data".

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See https://github.com/japhir/stdstats for potentially updated versions and Emacs org source files.

Please note that it takes quite some time to run all of the code, so if you want to evaluate it, make sure to do so line-by-line!

We set the seed here in an attempt to make these exact simulations reproducible, but this only works for the next random sampling call. Thus, you will not get identical results. Please load the included simulation output .rds files to generate identical plots (see below).

set.seed(1563435)

#### 1 install required packages

Install all below packages. Some of them (e.g. furrr) rely on the dev version available on GitHub. Install them with

devtools::install\_github("DavisVaughan/furrr")

Install the new package stdsim similarly, from

devtools::install\_github("japhir/stdsim")

or point to the directory in which you downloaded the stdsim folder with

devtools::install\_local("/path/to/stdsim")

#### 2 optional packages

I use this for my theme.

```
devtools::install_github("baptiste/egg")
```

#### 3 load libraries

We use the following packages for this project:

```
library(dplyr)
                   # for piping (%>%), mutate, and nice data manipulation
library(tidyr)
                   # for making tibbles tidy, (gather, spread etc.)
library(tibble)
                   # a better thing than a dataframe
library(ggplot2)
                   # plotting
library(tictoc)
                   # to keep track of how long the functions took
library(purrr)
                   # functional programming to map variables from lists of dataframes
library(magrittr)
                   # only used once for extract, but original package of the pipe
library(patchwork) # combine plots
library(furrr)
                   # furrr allows parallell purr functions w/ progress bars!
                   # the new R package created just for this paper!
library(stdsim)
```

#### 4 multiple cores

run simulations on as many cores as possible. Note that I read somewhere that the random number generator gets less random when using multiple cores, but it is much faster and since I perform all simulations often this shouldn't effect the results.

```
plan(multiprocess)
```

#### 5 set up plotting theme

We can use a different theme, but I like the one in the dev package egg. Install it or ignore this section of code.

```
theme_set(egg::theme_article(base_size = 11, base_family="Helvetica"))
```

#### 6 set up stdinfo etc.

This uses the default functions in stdsim to generate a tibble with standard and sample information. It doesn't add  $\delta_{47}$  values by default since they differ between labs and are not important for these simulations.

```
eth.info <- make_std_table()
smpinfo <- make_smp_info(c(0, 40))
stdev <- 14
# append d47 values based on actual measurement results for our MOTU
eth.info$d47 <- c(15.6, -13.2, 16.2, -13.1, NA_real_)</pre>
```

#### 6.1 setup axes

# 7 calculate temperature sensitivity as a function of temperature

Now we are interested in calculating the rate of change as a function of D47, so that we can calculate the change in temperature. So we take the derivative of the original T(D47) function.

```
tempcal_simplified <- function(Tc, slp=0.0449, int=0.167, kkelvin=273.15) {
    (slp * 1e6) / (Tc + kkelvin)^2 + int
}

tempcal_derivative <- function(Tc, slp=0.0449, int=0.167, kkelvin=273.15) {
    -((2 * slp * 1e6) / ((kkelvin + Tc) ^ 3))
}

revcal_simplified <- function(D47, slp=0.0449, int=0.167, kkelvin=273.15) {
    sqrt((slp * 1e6) / (D47 - int)) - kkelvin
}

revcal_derivative <- function(D47, slp=0.0449, int=0.167) {
    (sqrt(-(slp * 1e6) / (int - D47))) / (2 * int - 2 * D47)
}
    See
?revcal
?tempcal</pre>
```

for the actual function documentation.

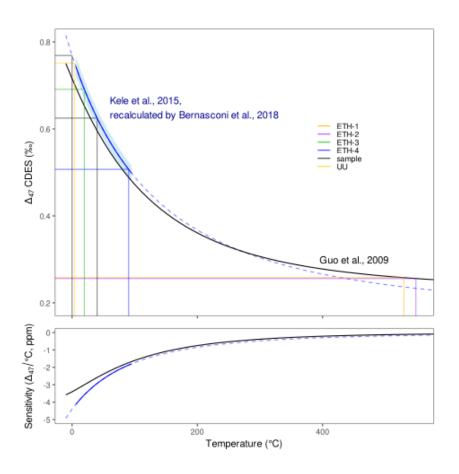
#### 7.1 temp\_sens\_pot

Calculate the sensitivity of the temperature calibration at the relevant temperature range, so that we can add an estimate of uncertainty in the temperature domain to plots.

The citeA:Kele2015 temperature calibration is only valid between 6 and 95  $^{\circ}\mathrm{C},$  so

```
# guo 2009 eqn. 18
# takes temperature in degrees celsius, converts to D47
guo_cal <- function(temp) {</pre>
  # convert degrees celsius to kelvin
  x <- temp + kkelvin
  # apply polynomial fit
  -3.33040e9 / x^4 + 2.32415e7 / x^3 - 2.91282e3 / x^2 - 5.54042 / x + 0.23252
}
# takes D47, converts to temperature in degrees celsius
guo_deriv <- function(temp) {</pre>
  x <- temp + kkelvin
  (5.54042 * x^3 + 5825.64 * x^2 - 69724500 * x + 13321600000)/x^5 * 1000
}
   Update standards to use Guo if ETH-1 or ETH-2. We hack it together
by numerically solving it.
guo_temp <- tibble(Tc = rng, D47 = guo_cal(Tc))</pre>
eth1_new_temp <- guo_temp$Tc[[which(near(guo_temp$D47, eth.info$D47[[1]], tol = .000000
eth2_new_temp <- guo_temp$Tc[[which(near(guo_temp$D47, eth.info$D47[[2]], tol = .000000
guo_std_temp <- bind_rows(eth.info, smpinfo) %>%
  mutate(temp = case_when(id == "ETH-1" ~ eth1_new_temp, #802.812 - kkelvin,
  id == "ETH-2" ~ eth2_new_temp, #822.2 - kkelvin,
  TRUE ~ temp))
plot_temp <- sensdf %>%
  filter(Tc >= 6, Tc <= 95) %>%
  ggplot(aes(y = D47, x = Tc)) +
  geom_ribbon(aes(ymin = lwr, ymax = upr), fill = "skyblue", alpha = .4) +
  geom_line(colour = "blue", linetype = 2, alpha = .5, data = sensdf) +
  geom_line(colour = "blue") +
```

```
## geom_line(colour = "black", data = guo_temp) +
  stat_function(fun = guo_cal, colour = "black") +
  geom_segment(aes(x = -Inf, xend = temp, y = D47, yend = D47, col = id),
       alpha=.5,
       inherit.aes=FALSE,
       data = guo_std_temp) +
  geom_segment(aes(x = temp, xend = temp, y = -Inf, yend = D47, col = id),
       alpha=.5,
       inherit.aes=FALSE,
       data = guo_std_temp) +
  annotate("text", x = 60, y = .65, label = "Kele et al., 2015, \nrecalculated by Berna
  annotate("text", x = 450, y = .3, label = "Guo et al., 2009") +
  scale_colour_manual(values = c(eth.info$col[-5], smpinfo$col[[1]], eth.info$col[[5]];
  labs(x = "Temperature (°C)", y = Delta[47] \sim "CDES (\u2030)") +
  coord_cartesian(ylim = c(.2, .8), xlim = c(0, 550)) +
  theme(axis.title.x=element_blank(),
axis.text.x=element_blank(),
legend.pos=c(.75, .6), legend.title = element_blank())
plot_sens <- sensdf %>%
  filter(Tc >= 6, Tc <= 95) %>%
  ggplot(aes(y = sens * 1e3, x = Tc)) +
  geom_line(colour = "blue", linetype = 2, alpha = .5, data = sensdf) +
  geom_line(colour = "blue") +
  stat_function(fun = guo_deriv, xlim = c(-10, 1000)) +
  coord_cartesian(ylim = c(-5, 0), xlim = c(0, 550)) +
  ## geom_vline(xintercept=c(0, 40), col="#ededed") +
  labs(x = "Temperature (°C)", y = "Sensitivity" ~ "(" * Delta[47] / "°C, ppm)")
temp_sens_pl <- plot_temp + plot_sens + plot_layout(nrow=2, heights=c(3, 1))
temp_sens_pl
```



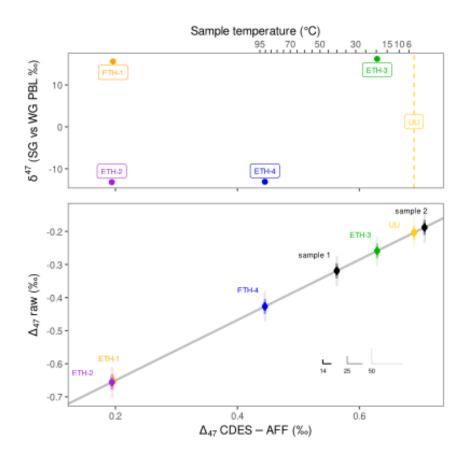
#### 8 create standard intro plot

The standards as a function of composition

```
lims <- c(.15, .71)
standards_plot <- ggplot(eth.info, aes(x = d47, y = D47.noacid, col = id, label = id))
geom_point(size = 2, show.legend = F) +
## geom_label(show.legend = F) +
ggrepel::geom_label_repel(size = 2.5, show.legend = FALSE) +
# add UU1 standard
geom_hline(yintercept = eth.info$D47.noacid[[5]], linetype = 2, col = eth.info$col[[annotate("label", x = 1.225, y = eth.info$D47.noacid[[5]], label = eth.info$id[[5]],
scale_colour_manual(values = eth.info$col) +
labs(colour = "", x = delta^{47}^"(SG vs WG PBL \u2030)",
    y = Delta[47] ~ "CDES" - "AFF (\u2030)") +</pre>
```

```
scale_y_continuous(sec.axis = sec_axis(^* sqrt((0.0449 * 1e+6)/(. + kaff - 0.167)) - 10.167)) - 10.167)
"Sample temperature (°C)", temp_breaks,
temp_labs)) +
  coord_flip(ylim=lims) +
  theme(axis.text.x.bottom=element_blank(), axis.title.x.bottom=element_blank())
  ## coord_cartesian(clip = "off") +
  ## theme(legend.pos = c(.15, .85))
   And the simulation input conditions illustrating the ETF.
stdevs <- c(14, 25, 50) / 1e3
xs < -.54 + c(0, .04, .08)
ys < -rep(-.6, 3)
standard_sample_data <- make_smp_info(c(0, 40)) %>%
  mutate(id=ifelse(temp == 40, "sample 1", "sample 2")) %>%
  ## mutate(id=paste(id, temp)) %>%
  bind_rows(eth.info)
base_plot <- standard_sample_data %>%
  ggplot(aes(x = D47.noacid, y = rawcat, col = id, label = id)) +
  # add etf
  geom_abline(intercept = kintercept, slope = kslope,
    linetype = 1, size = 1, col = "gray") +
  # 50 ppm uncertainty pointrange
  geom_linerange(aes(ymin = rawcat - 50 * kslope / 1e3,
    ymax = rawcat + 50 * kslope / 1e3),
    size = 1, linetype = 1, alpha = .1) +
  # 25 ppm uncertainty pointrange
  geom_linerange(aes(ymin = rawcat - 25 * kslope / 1e3,
    ymax = rawcat + 25 * kslope / 1e3),
    size = 1, linetype = 1, alpha = .4) +
  # 14 ppm uncertainty pointrange
  geom_linerange(aes(ymin = rawcat - 14 * kslope / 1e3,
    ymax = rawcat + 14 * kslope / 1e3),
    size = 1, linetype = 1) +
  geom_point(size=2) +
  # create a manual legend with the different input uncertainties
  annotate ("segment",
  x = xs, xend = xs,
   y = ys, yend = ys + stdevs * kslope,
```

```
alpha = c(1, .4, .1)) +
  annotate ("segment",
   x=xs, xend=xs + stdevs,
  y=ys, yend=ys,
  alpha=c(1, .4, .1)) +
  annotate("text",
  x = xs,
  y = ys - .02,
    label = c("14", "25", "50"), size = 2) +
  # add the input sample measurements
  ggrepel::geom_text_repel(force = 3, hjust = 1, nudge_y = .05, nudge_x = -.01, size=2
  # make it pretty, manual colour scale, samples are black
  scale_colour_manual(
      limits = standard_sample_data$id,
      values = standard_sample_data$col) +
  # nice axis labels
  labs(
    colour = "",
    x = Delta[47] \sim "CDES" - "AFF (\u2030)",
    y = Delta[47] \sim "raw (\u2030)"
 ) +
  scale_x_continuous(limits=lims) +
  theme(legend.pos = "none")
   In the text we combine them using patchwork, to create figure 1.
standards_pl <- standards_plot + base_plot + plot_layout(nrow=2, heights = c(.4, .6))
standards_pl
```



#### 9 micro benchmark

Calculate how long it takes for one simulation.

```
options(genplot = FALSE, verbose = FALSE)
tpersim <- microbenchmark::microbenchmark(sim_stds(out = "cis",
    stdtable = eth.info)) %>%
    summary() %>%
    pull(mean) / 100
tpersim
```

#### 10 example sims

We create some example simulations for fig. 2.

```
## set up small inputs dataframe
example_sims <- tibble(</pre>
    name=rep(c("Equal\nproportions", "Optimal\ndistribution", "Optimal\ndistribution + U
    stdfreqs=rep(list(c(1, 1, 1, 1, 0), c(1, 1, 9, 0, 0), c(1, 1, 0, 0, 9)), 2),
    smpt=c(rep(0, 3), rep(40, 3))) %>%
    ## run sims with inputs dataframe
    mutate(res = purrr::pmap(select(., -name), sim_stds, stdev=25, out="all", stdn=50, sim_stds, stdev=25, out=30, sim_stds, stdev=25, out=30, sim_stdev=25, out=30,
  # extract the default plots
  pl=purrr::map(res, plot_sim, graylines=F, point_alpha=.2, pointrange=T, labs=F, fixed=
  # add a row number for the next step
  exprow=1:n() %>% as.character())
# combine the smp and std outputs of each experiment, based on the row number
six_example_sims <- example_sims$res %>%
    map_dfr("smp", .id="exprow") %>%
    bind_rows(example_sims$res %>% map_dfr("std", .id="exprow")) %>%
    left_join(example_sims, by="exprow")
      We create the example plot
exmp_plot <- ggplot(six_example_sims, aes(x=D47.noacid, y=raw, col=id, fill=id)) +
    geom_smooth(aes(group=paste0(name, smpt)), method="lm", size=.1,
            fullrange=TRUE, data=filter(six_example_sims, id != "sample")) +
    geom_violin(alpha=.3, colour=NA, scale="count", width=.5, position=position_identity
    geom_point(shape=1, alpha=.2, size=.3) +
    facet_grid(rows=vars(name), cols=vars(paste(smpt, "°C"))) +
    ## coord_fixed(xlim=c(.1, .8)) +
    coord_cartesian(xlim=c(.14, .75)) +
    scale_colour_manual("ID",
            ## limits = c(out$cond$stdtable$id, out$cond$smpinfo$id),
            limits = c(example_sims$res[[1]]$cond$stdtable$id, example_sims$res[[1]]$cond$sm
            ## values = c(out$cond$stdtable$col, out$cond$smpinfo$col)) +
            values = c(example_sims$res[[1]]$cond$stdtable$col, example_sims$res[[1]]$cond$st
    scale_fill_manual("ID",
        limits = c(example_sims$res[[1]]$cond$stdtable$id, example_sims$res[[1]]$cond$smpin
        values = c(example_sims$res[[1]]$cond$stdtable$col, example_sims$res[[1]]$cond$smp;
    labs(x = Delta[47] \sim "CDES" - "AFF (\u2030)",
              y = Delta[47] \sim raw \sim "(\u2030)") +
    theme(legend.pos="top", legend.key.size=unit(3, "mm"), legend.text = element_text(si:
```

options(genplot=F, verbose=F)

```
strip.text.y = element_text(size = 8, angle = 90))
exmp_plot
   and the table for in the text:
tbl_exmp <-
  forplot_0 %>%
  bind_rows(forplot_40) %>%
  filter(expname %in% c("1:1:1:1:0", "1:1:9:0:0", "1:1:1:0:9")) %>%
    select(-stdfreqs, -exprow, -meanerr, -hascoldstandard) %>%
  group_by(expname, smpt, stdev) %>%
  ## nest() %>%
  summarize(err_mean = mean(smp, na.rm = TRUE) * 1e3,
    err_ci = qt((1 - .05), length(smp) - 1) * sd(smp, na.rm = TRUE) / sqrt(length(smp))
  mutate(err_temp = err_mean / 1e3 / tempcal_derivative(smpt) %>% abs,
 err_temp_ci = err_ci / 1e3 / tempcal_derivative(smpt) %>% abs,
   ) %>%
  arrange(stdev, -err_mean) %>%
  ## bind_cols(map_dfr(.$data, ~ mean(smp))) %>%
  pivot_wider(id_cols = c(expname, stdev), names_from = smpt, values_from = c(err_mean
  # change order of things
  mutate(Name = case_when(expname == "1:1:1:1:0" ~ "Equal proportions",
  expname == "1:1:9:0:0" ~ "Optimal distribution",
  expname == "1:1:1:0:9" ~ "Optimal distribution + UU1",
  TRUE ~ "wth")) %>%
  select(Name, expname, stdev, ends_with("_0"), ends_with("_40")) %>%
  as_tibble() %>%
  mutate(
    out_0_ppm = glue::glue("{round(err_mean_0, 2)} \\pm {round(err_ci_0, 2)}"),
    ## improv_0 = round(err_mean_0 / lead(err_mean_0), 2),
    ## improv_40 = round(err_mean_0 / lead(err_mean_0), 2),
    out_0_deg = glue::glue("{round(err_temp_0, 2)} \\pm {round(err_temp_ci_0, 2)}"),
    out_40_ppm=glue::glue("{round(err_mean_40, 2)} \\pm {round(err_ci_40, 2)}"),
    out_40_deg = glue::glue("{round(err_temp_40, 2)} \\pm {round(err_temp_ci_40, 2)}")
    ) %>%
  ## rename for latex output
  select(Name,
 'Standard distribution'=expname,
 '\\sigma' = stdev,
 '0\\us\\celsius (ppm)' = out_0_ppm,
 ## '\\times' = improv_0,
```

```
'40\\us\\celsius (ppm)'=out_40_ppm,
## '\\times' = improv_40,
## '0\\u
'0\\us\\celsius (\\celsius)'=out_0_deg,
'40\\us\\celsius (\\celsius)'=out_40_deg)
```

Name	Standard distribution	$\sigma$	0 °C (ppm)	$40^{\circ}\mathrm{C}\ (\mathrm{ppm})$	0 °C (°C)
Equal proportions	1:1:1:1:0	14	$9.37 \pm 0.12$	$7.21 \pm 0.09$	$2.13 \pm 0.03$
Optimal distribution	1:1:9:0:0	14	$6.8 \pm 0.09$	$5.58 \pm 0.07$	$1.54 \pm 0.02$
Optimal distribution + UU1	1:1:1:0:9	14	$6 \pm 0.07$	$5.63 \pm 0.07$	$1.36 \pm 0.02$
Equal proportions	1:1:1:1:0	25	$17.03 \pm 0.24$	$12.89 \pm 0.16$	$3.86 \pm 0.05$
Optimal distribution	1:1:9:0:0	25	$12.45 \pm 0.18$	$10.24 \pm 0.12$	$2.82 \pm 0.04$
Optimal distribution $+$ UU1	1:1:1:0:9	25	$10.86 \pm 0.12$	$10.11 \pm 0.1$	$2.46 \pm 0.03$
Equal proportions	1:1:1:1:0	50	$35.15 \pm 0.57$	$26.35 \pm 0.46$	$7.98 \pm 0.13$
Optimal distribution	1:1:9:0:0	50	$25.08 \pm 0.44$	$20.1 \pm 0.27$	$5.69 \pm 0.1$
Optimal distribution $+$ UU1	1:1:1:0:9	50	$22.09 \pm 0.33$	$20.03 \pm 0.28$	$5.01 \pm 0.08$

#### 11 stddis

To simulate the distribution of the standards we create a tibble of inputs for sim\_stds().

#### 11.1 create a list of all possible stddis combinations

```
# calculate relative abundances
  mutate(sums = rowSums(.),
    f1 = 'ETH-1' / sums,
    f2 = 'ETH-2' / sums,
    f3 = 'ETH-3' / sums,
    f4 = 'ETH-4' / sums,
    fu = UU1 / sums) \%
  # filter out the redundant ones
  distinct(f1, f2, f3, f4, fu, .keep_all = TRUE) %>%
  arrange('ETH-1', 'ETH-2', 'ETH-3', 'ETH-4', UU1)
# convert the proportions to a list we can put in our experimental matrix
props_list <- props %>%
  select(-c(sums, starts_with("f"))) %>%
  as.matrix() %>%
  split(seq(nrow(.)))
11.2
     expand the whole grid
stddis <- expand.grid(</pre>
    smpt = c(0, 40),
    stdfreqs = props_list, # this uses the list created previously
    stdev = c(50, 25, 14)) \%
  # we add an experiment name for plot labels and easy overview
  mutate(expname = map_chr(stdfreqs, paste, collapse = ":")) %>%
  # add cold standard logical for later filtering
  mutate(hascoldstandard = grepl("[[139]]", expname))
# repeat each experiment a hundred times
megastddis <- stddis[rep(stddis %>% nrow() %>% seq_len(), 100), ] %>%
  # and add a row character for later merging of results
  mutate(exprow = as.character(seq_len(n())))
```

#### 11.3 stddis runall

Note that we use the package future so that we can show a progress bar and use multiple cores. One could also use purrr::pmap\_dfr() which it's based on, here.

```
## I turn off plotting and info messages
options(genplot = FALSE, verbose = FALSE)
```

```
## keep track of how long it takes
message(nrow(megastddis), " simulations started at ", Sys.time())
## very rough expected finish time (it's usually faster)
message("expected to take until ", Sys.time() + tpersim * nrow(megastddis) / 4)
## track actual duration with tictoc
tic("stddis total time")
## run sim_stds with parameters from mgstddis and global parameters after it
stddis_cnf <- furrr::future_pmap_dfr(</pre>
       select(megastddis, smpt, stdfreqs, stdev),
       sim_stds, stdtable = eth.info,
       out = "cis", stdn = 50, smpn = 50,
       .id = "exprow", # append a row name id
       .progress=TRUE  # show a progress bar
     ) %>%
  filter(id == "smp") %>%
                                                   # filter output
  select(exprow, id, cv) %>%
                                                   # select output
  spread(id, cv) %>%
                                                   # make it wide format
  right_join(megastddis, by="exprow")
                                                   # join it with experimental df
toc()
message("simulations ran until ", Sys.time())
   Save the results so that we don't have to run the simulations every time.
saveRDS(stddis_cnf, "stddis_cnf_2019-06-17.rds")
   Restore the results from previous simulations.
stddis_cnf <- readRDS("stddis_cnf_2019-06-17.rds")</pre>
      arrange results for plot
We re-organize the dataframes and make a selection of the best results for 0
and 40 degrees.
```

# the mean error of equal proportions at 0 degrees

filter(smpt == 0, expname == "1:1:1:1:0") %>%

normerr\_0 <- stddis\_cnf %>%

pull(meanerr)

summarize(meanerr=mean(smp)) %>%

```
idealerr_0 <- stddis_cnf %>%
  filter(smpt == 0, expname == "1:1:9:0:0") %>%
  summarise(meanerr = mean(smp)) %>%
  pull(meanerr)
normerr_40 <- stddis_cnf %>%
  filter(smpt == 40, expname == "1:1:1:1:0") %>%
  summarize(meanerr=mean(smp)) %>%
  pull(meanerr)
idealerr_40 <- stddis_cnf %>%
  filter(smpt == 40, expname == "1:1:9:0:0") %>%
  summarise(meanerr = mean(smp)) %>%
  pull(meanerr)
forplot_0 <-
  stddis_cnf %>%
  filter(smpt == 0) %>%
  # append mean error per treatment to original dataframe
  group_by(expname) %>%
  summarise(meanerr = mean(smp, na.rm=TRUE)) %>%
  left_join(filter(stddis_cnf, smpt == 0), by = "expname") %>%
  # if there is no cold standard, it should be at least as good as normerr
  filter((!hascoldstandard & (meanerr <= normerr_0)) |</pre>
  # if there is a cold standard, it should be at least as good as idealerr
   (hascoldstandard & (meanerr <= idealerr_0))) %>%
  arrange(meanerr, expname) %>%
  ungroup() %>%
  # this is a hack to order the labels of a factor in a plot
  mutate(expname = factor(expname, unique(expname)))
forplot_40 <-
  stddis_cnf %>%
  filter(smpt == 40) %>%
  # append mean error per treatment to original dataframe
  group_by(expname) %>% # note that we do not take into account sample temp...
  summarise(meanerr = mean(smp, na.rm=TRUE)) %>%
  left_join(filter(stddis_cnf, smpt == 40), by = "expname") %>%
  # if there is no cold standard, it should be at least as good as normerr
```

```
filter((!hascoldstandard & (meanerr <= normerr_40)) |
# if there is a cold standard, it should be at least as good as idealerr
  (hascoldstandard & (meanerr <= idealerr_40))) %>%
arrange(meanerr, expname) %>%
mutate(expname = factor(expname, unique(expname)))
```

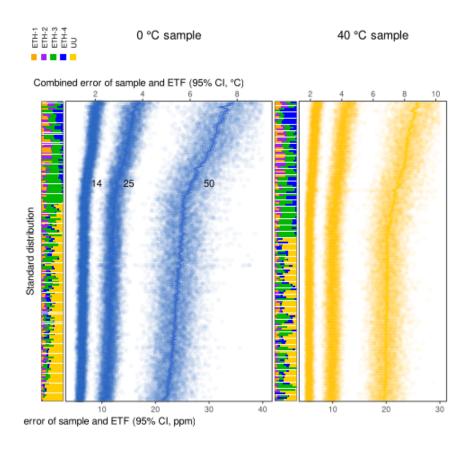
#### 11.5 create barcharts for proportion axes

```
forplot_0_x <- forplot_0 %>%
  distinct(expname, .keep_all = TRUE) %>%
  unnest(cols=stdfreqs) %>%
  mutate(std = rep(c(pasteO("ETH-", 1:4), "UU"), n()/5))
x_0 <- forplot_0_x %>%
  ggplot(aes(x = expname, y = stdfreqs, fill = std)) +
  geom_col(position="fill") +
  ## geom_text(aes(label = lab), y = .5, size = 2,
  ##
               data = mutate(forplot_0_x,
                             lab = ifelse(expname %in% c("1:1:1:1:0", "1:1:9:0:0", "1:
  ##
  scale_fill_manual(values=eth.info$col,
    guide = guide_legend(label.position = "top",
 direction = "horizontal",
 label.theme = element_text(size = 8, angle = 90, hjust = 0))) +
  scale_y_reverse(position = "right", expand = c(0, 0)) +
  labs(x = "Standard distribution", y = "ETH-1:2:3:4:UU1", fill = "") +
  ## labs(x = "Standard distribution", y = "ETH-<span style='color:orange'>1</span>:<s
  coord_flip() +
  theme(axis.text = element_blank(),
axis.ticks = element_blank(),
plot.margin = margin(r = 0),
axis.title.x = element_blank(),
legend.pos = "top",
legend.justification = 12,
legend.spacing.x = unit(1, "mm"),
## legend.margin = margin(2, 0, 2, 0),
legend.key.size = unit(2, "mm"))
forplot_40_x <- forplot_40 %>%
  distinct(expname, .keep_all = TRUE) %>%
  unnest(cols=stdfreqs) %>%
```

```
mutate(std = rep(c(paste0("ETH-", 1:4), "UU"), n()/5))
x_40 <- x_0 %+% forplot_40_x +
  theme(axis.title.x = element_blank(),
axis.title.y = element_blank(),
legend.pos = "none")
11.6
      stddis pl
sub_vjust <- -7</pre>
sds < -tibble(expname = "1:0:9:1:0", sd = c(14, 25, 50), smp = c(.006, .012, .027) + .000
# create a plot
stddis_pl_0 <- forplot_0 %>%
  ggplot(aes(x = expname, y = smp * 1e3, # error in ppm)
    colour = factor(smpt), fill = factor(smpt))) +
  # theming
  labs(x = "Standard distribution",
    y = "Combined error of sample and ETF (95% CI, ppm)",
    fill = ktit_smpid,
    colour = ktit_smpid,
    shape = "Total standards") +
  scale_fill_manual(values = kcols[[1]], labels = "0 °C") +
  scale_colour_manual(values = kcols[[1]], labels = "0 °C") +
  scale_y_continuous(breaks = seq(0, 65, 10), lim = c(5, 40),
    sec.axis = sec_axis("Combined error of sample and ETF (95% CI, °C)",
      trans = \tilde{} . / abs(tempcal_derivative(0) * 1e3), breaks = seq(0, 20, 2))) +
    plot.title = element_text(hjust = 0.5, vjust = -10),
    plot.margin = margin(1 = 0),
    ## plot.subtitle = element_text(size = 8, hjust = -.1, vjust = -8),
    axis.title.x.top = element_text(hjust = 7),
    axis.title.x.bottom = element_text(hjust = 7),
    axis.text.y = element_blank(),
    axis.ticks.y = element_blank(),
    axis.title.y = element_blank(),
    strip.text = element_blank(),
    legend.position = "none") +
  # manual legend for input stdevs (sigma)
  # the actual data
```

```
geom_point(alpha = .03) +
  ## annotate("text", x = -Inf, y = Inf, label = paste0("\u3c3 = ", c(14, 25, 50))) +
  geom_text(aes(label = sd), data = sds, colour = "black") +
  coord_flip(clip = "off") +
  stat_summary(aes(group=stdev), geom="ribbon", fun.data=mean_cl_normal, fun.args=list
  stat_summary(aes(group=stdev), geom="line", fun.data=mean_cl_normal) +
  labs(title="0 °C sample") #, subtitle = "ETH-1:2:3:4:UU1")
# add the average lines +- 95% CIs for all the sample temperatures
stddis_pl_40 <- forplot_40 %>%
  filter(smpt == 40) %>%
  ggplot(aes(x = expname, y = smp * 1e3, # error in ppm
    colour = factor(smpt), fill = factor(smpt))) +
  # theming
  labs(x = "Standard distribution",
   y = "Combined error of sample and ETF (95% CI, ppm)",
   fill = ktit_smpid,
   colour = ktit_smpid,
    shape = "Total standards") +
  scale_fill_manual(values = kcols[[2]], labels = "40 °C") +
  scale_colour_manual(values = kcols[[2]], labels = "40 °C") +
  scale_y_continuous(breaks = seq(0, 65, 10), lim = c(5, 30),
    sec.axis = sec_axis("Combined error of sample and ETF (95% CI, °C)",
      trans = \tilde{} / abs(tempcal_derivative(40) * 1e3), breaks = seq(0, 20, 2))) +
  theme(
   plot.title = element_text(hjust = 0.5, vjust = -10),
   plot.margin = margin(1 = 0),
   plot.subtitle = element_blank(),
   axis.title = element_blank(),
   axis.text.y = element_blank(),
    axis.ticks.y = element_blank(),
   axis.title.y = element_blank(),
   strip.text = element_blank(),
   legend.position = "none") +
  labs(title="40 °C sample") +
  coord_flip(clip = "off") +
  # the actual data
  geom_point(alpha = .03) +
  stat_summary(aes(group=stdev), geom="ribbon", fun.data=mean_cl_normal,
       fun.args=list(conf.int=0.95), alpha=.4, colour=NA) +
```

```
stat_summary(aes(group=stdev), geom="line", fun.data=mean_cl_normal)
## stddis_pl <- stddis_pl_0 + stddis_pl_40 + plot_layout(widths=c(35 / 25 , 1))
stddis_pl <- x_0 + stddis_pl_0 + x_40 + stddis_pl_40 + plot_layout(widths = c(.15, 35/stddis_pl_40))</pre>
```



#### 11.7 prep text

This generates some dataframes so we can easily extract relevant averages for use in-text.

```
stddis_exmp_0 <- forplot_0 %>%
filter(smpt == 0) %>%
group_by(expname) %>%
summarize(err=mean(smp, na.rm=TRUE), cnf = confidence(smp, n(), alpha=0.05)) %>%
```

```
arrange(err)

stddis_exmp_40 <- forplot_40 %>%
  filter(smpt == 40, stdev == 25) %>%
  group_by(expname) %>%
  summarize(err=mean(smp,na.rm=TRUE), cnf=confidence(smp, n(), alpha= 0.05)) %>%
  arrange(err)
```

#### 12 stdvssmp

Here we simulate the standards versus samples based on three input distributions,

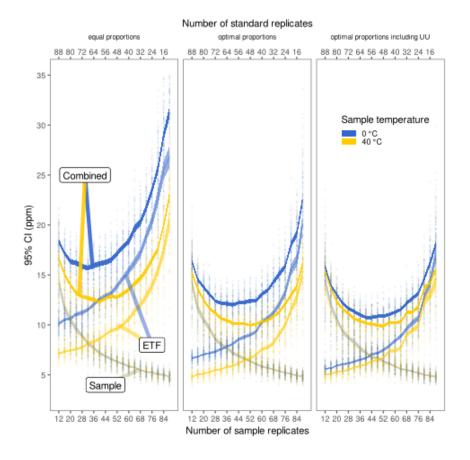
#### 12.1 setup

```
stdvssmp <- expand.grid(</pre>
  smpt = c(0, 40), stdn = as.integer(seq(12, 88, 4)),
  dist = c("equal proportions", "optimal proportions", "optimal proportions including "
  stdev = c(50, 25, 14)) \%
  as_tibble() %>%
  mutate(smpn = as.integer(100 - stdn),
 n = as.integer(smpn + stdn),
 smpf = smpn / n,
 stdf = stdn / n,
 stdfreqs = case_when(dist == "equal proportions" ~ list(c(1, 1, 1, 1, 0)),
      dist == "optimal proportions" ~ list(c(1, 1, 9, 0, 0)),
      dist == "optimal proportions including UU1" \sim list(c(1, 1, 1, 0, 9))))
# repeat each experiment a hundred times
megastdvssmp <- stdvssmp[rep(stdvssmp %>% nrow() %>% seq_len(), 100), ] %>%
  mutate(exprow = as.character(seq_len(n())))  # row number
12.2
     run sims
# again, I turn off plotting and info messages
options(genplot = FALSE, verbose = FALSE)
# keep track of time
message(nrow(megastdvssmp), " simulations started at ", Sys.time())
message("expected to take until ", Sys.time() + tpersim * nrow(megastdvssmp) / 4)
```

```
# run duration
tic("overview")
# run sim_stds with parameters from megastdvssmp and global parameters after it
stdvssmp_cnf <- furrr::future_pmap_dfr(</pre>
 select(megastdvssmp,
smpt, stdn, stdev, smpn, stdfreqs),
 sim_stds,
 stdtable = eth.info, out = "cis",
 .id = "exprow",
                          # append a row name id
 .progress=TRUE
 ) %>%
  filter(id %in% c("etf", "sample", "smp")) %>%
                                                   # filter output
  select(exprow, id, cv) %>%
                                                   # select output
  spread(id, cv) %>%
                                                   # make it wide format
  right_join(megastdvssmp)
                                                   # join it with experimental df
toc()
message("simulations ran until ", Sys.time())
12.3
     save results
saveRDS(stdvssmp_cnf, "stdvssmp_cnf_2019-06-13.rds")
stdvssmp_cnf <- readRDS("stdvssmp_cnf_2019-06-13.rds")</pre>
12.4
     tidy it up
tidy_stdvssmp_results <- stdvssmp_cnf %>%
  gather(errortype, error, sample, etf, smp)
12.5
      stdvssmp pl
## create label annotation because the legend with opacity was unclear
smpn=c(76, 44, 30)
error=c(8, 4, 25)
labels = c("ETF", "Sample", "Combined")
leg <- tibble(</pre>
  ## x=c(58, 67, 28,
         53, 62, 63),
  ## y=c(14.5, 5.5, 16,
```

```
##
         10, 5.5, 12.5),
 x=rep(smpn, 2), y=rep(error, 2),
  xend=c(58, 67, 35,
53, 67, 26),
  yend=c(15, 5.5, 16,
 10, 5.5, 13),
  smpt=c(rep(0, 3), rep(40, 3)),
  stdev = 25,
  ## lab=paste(labels, "uncertainty at", smpt, "°C"),
  lab=rep(labels, 2),
  errortype = rep(c("etf", "sample", "smp"), 2),
  dist=rep("equal proportions", 6))
      ## smpt=0)
stdvssmp_pl <- ggplot(tidy_stdvssmp_results %>% filter(stdev == 25),
  aes(x = smpn, y = error * 1e3, fill = as.factor(smpt),
    colour = as.factor(smpt), alpha = as.factor(errortype))) +
  # draw the points for all simulations, but make them very vague
  geom_point(alpha = .05, size = .3) +
  # draw an error range through the different experiments
  stat_summary(geom = "ribbon", colour = NA, # no border
   fun.data = mean_cl_normal,
   fun.args = list(conf.int = .95, na.rm = TRUE)) +
  # draw an average through the different experiments
  stat_summary(geom = "line", fun.data = mean_cl_normal) +
  # three standard distributions on the x-facets, 3 standard deviations on y
  ## we add some ugly labels here because they are more clear than a legend in this car
  geom_segment(aes(x=x, xend=xend, y=y, yend=yend), data=leg, size=2, show.legend=F) +
  geom_label(aes(x=x, y=y, label=lab), data=leg, colour="black", fill="white", alpha=1
  facet_grid(#rows = vars(stdev),
   ## rows = vars(stdev),
   cols = vars(dist),
   as.table = FALSE,
   ## shrink = TRUE,
   ## scales = "free_y",
   ## space = "free_y"
   ) +
  # x-axes
  scale_x_continuous("Number of sample replicates", lim = c(10, 90),
   breaks = seq(12, 88, 8),
```

```
sec.axis = sec_axis(~ 100 - ., name = "Number of standard replicates",
     breaks = seq(88, 12, -8))) +
 scale_y_continuous("95% CI (ppm)",
    ## trans="log10",
    limits = c(3, 35),
    breaks = c(seq(0, 50, 5), seq(60, 100, 10)),
    ## sec.axis = sec_axis(~. / abs(tempcal_derivative(0) * 1e3), name = "Approximate
## breaks=c(seq(0, 20, 1), seq(25, 40, 5)))
    ) +
## coord_trans(y = "log10") +
# colours
 scale_colour_manual(ktit_smpid, labels = klab_smpid, values = kcols) +
 scale_fill_manual(ktit_smpid, labels = klab_smpid, values = kcols) +
 scale_alpha_manual("Source of Error",
  labels = c("ETF", "Sample", "Combined"),
  values = c(.5, .2, .9), guide=FALSE) +
# theming
## annotation_logticks(sides="l") +
 theme(legend.position = c(.85, .8), strip.text=element_text(size=8), strip.placement=
stdvssmp_pl
```



## 12.6 supplementary figure plot with all input standard deviations

we use a new package that allows different scales on different facets

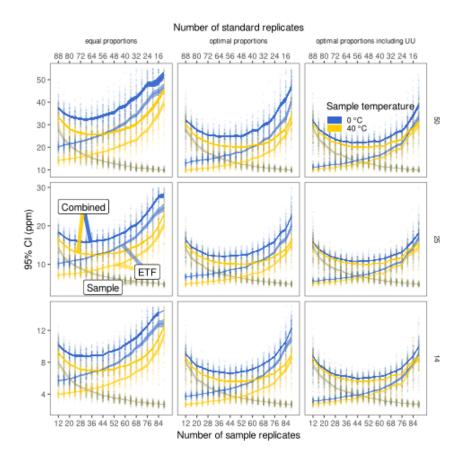
devtools::install\_github("zeehio/facetscales")

```
stdvssmp_pl_all <- ggplot(tidy_stdvssmp_results,
  aes(x = smpn, y = error * 1e3, fill = as.factor(smpt),
    colour = as.factor(smpt), alpha = as.factor(errortype))) +
  # draw the points for all simulations, but make them very vague
  geom_point(alpha = .05, size = .3) +
  # draw an error range through the different experiments
  stat_summary(geom = "ribbon", colour = NA, # no border
  fun.data = mean_cl_normal,</pre>
```

fun.args = list(conf.int = .95, na.rm = TRUE)) +

```
# draw an average through the different experiments
stat_summary(geom = "line", fun.data = mean_cl_normal) +
# three standard distributions on the x-facets, 3 standard deviations on y
## we add some ugly labels here because they are more clear than a legend in this car
geom_segment(aes(x=x, xend=xend, y=y, yend=yend), data=leg, size=2, show.legend=F) +
geom_label(aes(x=x, y=y, label=lab), data=leg, colour="black", fill="white", alpha=1
facetscales::facet_grid_sc(
 rows = vars(stdev),
 cols = vars(dist),
 as.table = FALSE,
 shrink = TRUE,
 scales = list(
    y = list('14' = scale_y_continuous("95% CI (ppm)", lim = c(2, 15)),
     '25' = scale_y_continuous("95% CI (ppm)", lim = c(3, 30)),
     '50' = scale_y_continuous("95% CI (ppm)", lim = c(9, 55)))
 ),
 ## space = "free_y"
 ) +
# x-axes
scale_x_continuous("Number of sample replicates", lim = c(10, 90),
 breaks = seq(12, 88, 8),
 sec.axis = sec_axis(~ 100 - ., name = "Number of standard replicates",
    breaks = seq(88, 12, -8)) +
## scale_y_continuous(,
##
                      ## trans="log10",
##
                      ## limits = c(NA, 50),
                      breaks = c(seq(0, 50, 5), seq(60, 100, 10)),
##
##
                      ## sec.axis = sec_axis(~. / abs(tempcal_derivative(0) * 1e3), n
                                          ## breaks=c(seq(0, 20, 1), seq(25, 40, 5))
##
##
                      ) +
## coord_trans(y = "log10") +
# colours
scale_colour_manual(ktit_smpid, labels = klab_smpid, values = kcols) +
scale_fill_manual(ktit_smpid, labels = klab_smpid, values = kcols) +
scale_alpha_manual("Source of Error",
  labels = c("ETF", "Sample", "Combined"),
  values = c(.5, .2, .9), guide=FALSE) +
# theming
## annotation_logticks(sides="l") +
theme(legend.position = c(.85, .85), strip.text=element_text(size=8), strip.placemen
```

#### stdvssmp\_pl\_all



#### 13 prop-eth3

In the discussion we create a new set of simulations.

#### 13.1 prop-eth3 for continuous sample range

```
new_smp_info2 <- tibble(smpid = "smp", smp_D47 = seq(0.18, 0.9, 0.0025)) %>%
  mutate(smp_D47.noacid = smp_D47 - kaff,
  rawcat = smp_D47.noacid * kslope + kintercept,
  smpt = revcal(smp_D47, ignorecnf = TRUE))
prop_eth3 <- seq(.02, .98, length.out = 500)</pre>
```

### 13.2 prop-eth3 expand experimental matrices and run simulations

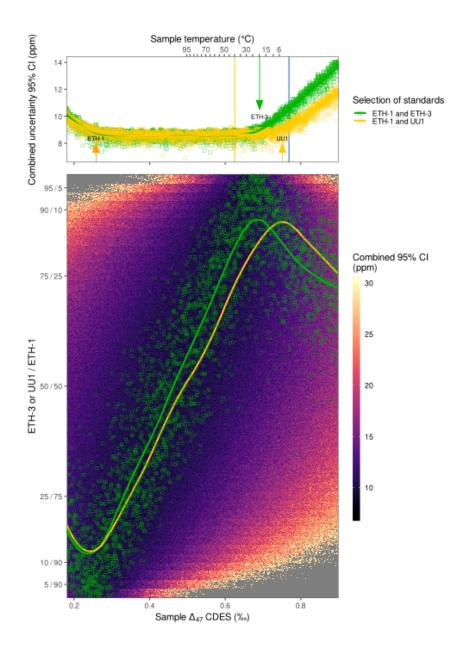
```
mat <- expand.grid(smp_D47 = new_smp_info2$smp_D47, prop_eth3 = prop_eth3) %%</pre>
  left_join(new_smp_info2, by = "smp_D47") %>%
  mutate(prop_left = 1 - prop_eth3,
 exprow = as.character(seq_along(1:n()))) %>%
  mutate(stdfreqs = select(., prop_eth3, prop_left) %>% as.matrix() %>% split(seq(nrow
  select(-prop_left)
13.2.1 run many sims
smp_out <- furrr::future_pmap_dfr(</pre>
    select(mat, stdfreqs, smpt),
    sim_stds,
    ## here we subset the standards to ETH-3 and ETH-1, in that order
    stdtable = make_std_table()[c(3, 1), ],
    stdev = 25, out = "cis", .id = "exprow", .progress = TRUE) %>%
  filter(id %in% c("etf", "sample", "smp")) %>%  # filter output
  select(exprow, id, cv) %>%
                                                  # select output
  spread(id, cv) %>%
                                                  # make it wide format
  right_join(mat, by = "exprow") %>% # join it with experimental df
  mutate(exp="ETH-1 and ETH-3")
13.2.2 save results
saveRDS(smp_out, "smp_out_new_2019-06-12.rds")
smp_out <- readRDS("smp_out_new_2019-06-12.rds")</pre>
13.3
     smp out uu
smp_out_uu <-</pre>
  furrr::future_pmap_dfr(
   select(mat, stdfreqs, smpt),
   sim_stds,
   # here we subset the standards to ETH-3 and ETH-1, in that order
   stdtable = make_std_table()[c(5, 1), ],
   stdev = 25, out = "cis", .id = "exprow", .progress = TRUE) %>%
```

select(exprow, id, cv) %>%

filter(id %in% c("etf", "sample", "smp")) %>% # filter output

# select output

```
spread(id, cv) %>%
                                                 # make it wide format
  right_join(mat, by = "exprow") %>% # join it with experimental df
  mutate(exp="ETH-1 and UU1")
saveRDS(smp_out_uu, "smp_out_uu_2019-06-12.rds")
smp_out_uu <- readRDS("smp_out_uu_2019-06-12.rds")</pre>
13.4 best dat
best_range <- 1:10
best_dat <- bind_rows(smp_out, smp_out_uu) %>%
  group_by(exp, smp_D47) %>%
  arrange(smp) %>%
  slice(best_range)
13.5
      smp out comb
combine the sims for one plot with faceting
smp_out_comb <- smp_out %>%
  bind_rows(smp_out_uu) %>%
  select(exprow, smp_D47, prop_eth3, exp, smp) %>%
  spread("exp", "smp") %>%
  mutate(diff='ETH-1 and UU1' - 'ETH-1 and ETH-3')
13.6 prop eth3 pl
13.6.1 plot best
13.6.2 plot prop
13.6.3 combine and print
prop_eth3_pl <- plot_best + plot_prop + plot_layout(nrow=2, heights=c(.2, .8))</pre>
prop_eth3_pl
```



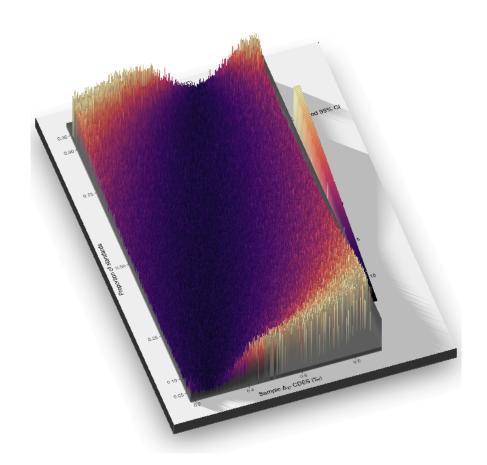
#### 13.6.4 3d rayshader plot

This one is not included in the manuscript or the supplementary information pdf, but I highly recommend creating one to play around with it!

# new rayshader 3d option

```
# remotes::install_github("tylermorganwall/rayshader")
library(rayshader)
ray <- smp_out %>%
  ggplot(aes(x = smp_D47, y = prop_eth3, fill = smp * 1e3)) +
  geom_raster() +
  ## geom_smooth(aes(col = exp), se = F, method = "loess", span = .3, size = 1,
               data = best_dat) +
  ## geom_point(aes(col = exp), shape = 0, size = 1, alpha = .5,
                data = filter(best_dat, exp == "ETH-1 and ETH-3")) +
  ##
  labs(x = Sample^Delt[[47]]"CDES (\u2030)",
       y = "Proportion of standards",
       fill = "Combined 95% CI \setminus n(ppm)") +
  viridis::scale_fill_viridis(
     ## the rescaler works nicely, but messes up the legend a bit
     ## rescaler = function(x, to = c(0, 1), from = NULL, newmax=30) {
          ifelse(x < newmax,
     ##
     ##
                 scales::rescale(x, to = to, from = c(min(x, na.rm = TRUE), newmax)),
     ## I'll go back to simple clipping again
     limits = c(NA, 30),
     expand=c(0, 0),
     ## oob = function(x) \{x\},
     option = "magma",
     breaks = err_breaks, #labels = err_ticks
  scale_y_continuous(expand = c(0, 0), breaks = c(.05, .1, .25, .5, .75, .9, .95)) +
  scale_x_continuous(expand = c(0, 0), lim=c(.18, .9),
  sec.axis = sec_axis( sqrt((0.0449 * 1e+6)/(. - 0.167)) - 273.15,
  "Sample temperature (°C)", temp_breaks, temp_labs)) +
  theme(legend.key.width = unit(.3, "cm"),
legend.key.height = unit(2, "cm"),
strip.placement="outside")
plot_gg(ray, multicore=TRUE, width=5, height=7, scale=350, raytrace=TRUE,
sunangle=40)
## render_movie("rayshader_movie_plot.mp4")
render_snapshot("imgs/rayshader_snapshot.png")
```

See the screenshot for a preview



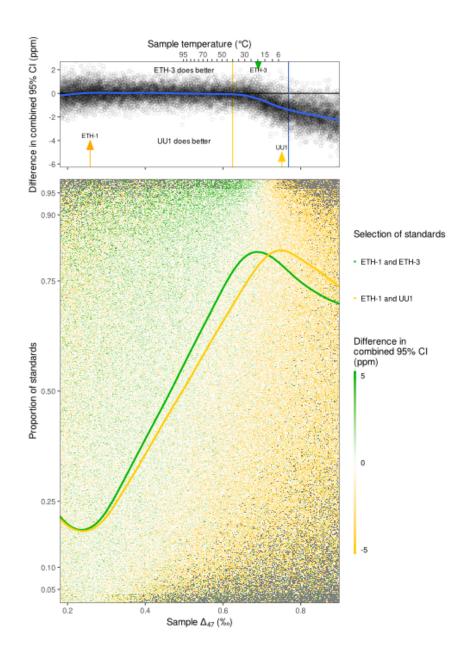
#### 13.7 best prop diff pl

The difference plot for the supplementary information.

```
best_dat_100 <- bind_rows(smp_out, smp_out_uu) %>%
  group_by(exp, smp_D47) %>%
  arrange(smp) %>%
  slice(1:100)

best_dat_comb <- best_dat_100 %>%
  ungroup() %>%
  select(smp_D47, prop_eth3, exp, smp) %>%
  spread("exp", "smp") %>%
```

```
mutate(diff='ETH-1 and UU1' - 'ETH-1 and ETH-3')
plot_best_comb <- best_dat_comb %>%
  ggplot(aes(x = smp_D47, y = diff * 1e3)) +
  geom_hline(yintercept=0) +
  annotate ("text", x = c(.5, .5), y = c(2, -4), label = c("ETH-3 does better", "UU1 does
  geom_vline(xintercept=smpinfo$D47, col=kcols[1:2]) +
  ## show ALL the points?
  ## geom_point(shape = 16, alpha = .1, data=bind_rows(smp_out,smp_out_uu)) +
  geom_point(shape = 1, alpha = .1) +
  geom_smooth(method="loess", span = .3) +
  labs(col="Selection of standards",
       x = Sample \sim Delta[47] \sim "(\u2030)",
       y = "Difference in combined 95% CI (ppm)") +
  scale_colour_manual(values=eth.info$col[c(3, 5)]) +
  scale_x_continuous(expand=c(0, 0), lim=c(.18, .9),
     sec.axis = sec_axis(~sqrt((0.0449 * 1e+6)/(. - 0.167)) - 273.15,
 "Sample temperature (°C)", temp_breaks, temp_labs)) +
  scale_y\_continuous(expand = c(0, 0)) +
  ## add arrows to ETH-1 and ETH-3 positions
  annotate ("segment",
   arrow = arrow(angle = 20, length = unit(.4, "cm"), type = "closed"),
  x = eth.info$D47[c(1, 3, 5)], y = c(-Inf, Inf, -Inf),
  xend = eth.info$D47[c(1, 3, 5)], yend = c(-4, 2, -5),
   colour = eth.info$col[c(1, 3, 5)],
   alpha = 1,
   size = .4) +
  annotate("text", x = eth.info$D47[c(1, 3, 5)], # + c(0.03, -.02),
   y = c(-4, 2, -5), label = eth.info(id[c(1, 3, 5)],
   vjust= c(-.5, .5, -.5), size = 2.5) +
  theme(legend.pos=c(.2, .7),
## shared axis with bottom panel
axis.title.x.bottom=element_blank(), axis.text.x.bottom=element_blank())
plot_prop_diff <- smp_out_comb %>%
  ggplot(aes(x = smp_D47, y = prop_eth3, fill = diff * 1e3)) +
  geom_raster() +
  geom_smooth(aes(col = exp, fill=smp*1e3), se = F, method = "loess", span = .3, size :
      data = best_dat_100) +
  ## geom_point(aes(col=exp), data=best_dat_100, shape = 1) +
```



#### 13.8 calculate some summary statistics for use in-text

# these are some helper functions to calculate the values we put in the text # get the confidence value at alpha value alpha from a t-distribution confidence <- function(x, n, alpha=.05) {

```
qt(1 - alpha / 2, df=n - 1) * sd(x) / sqrt(n)
}
# convert x from permil to ppm and round down to dig digits.
ppmround <- function(x, dig=2) {</pre>
  round(x * 1e3, digits=dig)
}
interp_eth3 <- best_dat %>%
  ungroup() %>%
  filter(exp=="ETH-1 and ETH-3", smp_D47 \ge eth.info$D47[[1]], smp_D47 \le eth.info$D47
  summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
extrap_eth3 <- best_dat %>%
  ungroup() %>%
  filter(exp=="ETH-1 and ETH-3", smp_D47 >= eth.info$D47[[5]]) %>%
  summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
interp_uu <- best_dat %>%
  ungroup() %>%
  filter(exp=="ETH-1 and UU1", smp_D47 \ge eth.info$D47[[1]], smp_D47 \le eth.info$D47[[1]]
  summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
extrap_uu <- best_dat %>%
  ungroup() %>%
  filter(exp=="ETH-1 and UU1", smp_D47 >= eth.info$D47[[5]]) %>%
  summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
eth3_to_uu_eth3 <- best_dat %>%
  ungroup() %>%
  filter(exp=="ETH-1 and ETH-3", smp_D47 \ge eth.info$D47[[3]], smp_D47 \le eth.info$D47
  summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
eth3_to_uu_uu <- best_dat %>%
  ungroup() %>%
```

```
filter(exp == "ETH-1 and UU1", smp_D47 >= eth.info$D47[[3]], smp_D47 <= eth.info$D47
  summarize(mean=mean(smp, na.rm = TRUE),
    ci=confidence(smp, n=n()))
eth3_to_uu_tempsens <- seq(eth.info$D47[[3]], eth.info$D47[[5]], .01) %>%
  tempcal_derivative() %>%
  mean()
uu_to_0_eth3 <- best_dat %>%
  ungroup() %>%
  filter(exp=="ETH-1 and ETH-3", smp_D47 \ge eth.info$D47[[5]], smp_D47 \le smp_D47
  summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
uu_to_0_uu <- best_dat %>%
  ungroup() %>%
  filter(exp=="ETH-1 and UU1", smp_D47 >= eth.info$D47[[5]], smp_D47 <= smpinfo$D47[[1]
  summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
uu_to_0_tempsens <- seq(eth.info$D47[[5]], smpinfo$D47[[1]], .01) %%
  tempcal_derivative() %>%
  mean()
```

# 14 prop-eth3 with a very very very cold and hot standard?

The reviewers requested another set of simulations.

#### 14.1 prop-eth3 for continuous sample range

```
same as before
```

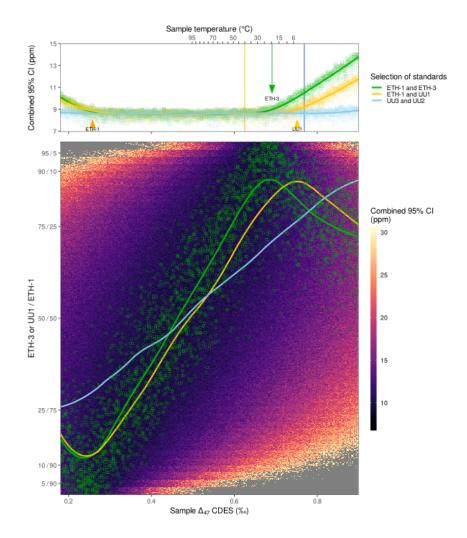
```
new_smp_info2 <- tibble(smpid = "smp", smp_D47 = seq(0.18, 0.9, 0.0025)) %>%
  mutate(smp_D47.noacid = smp_D47 - kaff,
  rawcat = smp_D47.noacid * kslope + kintercept,
  smpt = revcal(smp_D47, ignorecnf = TRUE))
prop_eth3 <- seq(.02, .98, length.out = 500)</pre>
```

### 14.2 prop-eth3 expand experimental matrices and run simulations

```
same as before
mat <- expand.grid(smp_D47 = new_smp_info2$smp_D47, prop_eth3 = prop_eth3) %%</pre>
  left_join(new_smp_info2, by = "smp_D47") %>%
  mutate(prop_left = 1 - prop_eth3,
 exprow = as.character(seq_along(1:n()))) %>%
  mutate(stdfreqs = select(., prop_eth3, prop_left) %>% as.matrix() %>% split(seq(nrow
  select(-prop_left)
14.3 run many sims
brrr <- make_std_table(id = c("UU2", "UU3"), col = c("darkblue", "red"),</pre>
       D47_{std} = c(0.9252, 0.0266)) \# based on Wang 2004
brrr_smp_out <- furrr::future_pmap_dfr(</pre>
    select(mat, stdfreqs, smpt),
    sim_stds,
    ## here we subset the standards to ETH-3 and ETH-1, in that order
    stdtable = brrr,
    stdev = 25, out = "cis", .id = "exprow", .progress = TRUE) %>%
  filter(id %in% c("etf","sample", "smp")) %>%  # filter output
  select(exprow, id, cv) %>%
                                                    # select output
  spread(id, cv) %>%
                                                    # make it wide format
  right_join(mat, by = "exprow") %>% # join it with experimental df
  mutate(exp="UU3 and UU2")
14.3.1 save results
saveRDS(brrr_smp_out, "brrr_smp_out_2019-08-20.rds")
   the 0.8~\mathrm{U2}~\mathrm{standard} + \mathrm{ETH}\text{-}1
brrr_smp_out <- readRDS("brrr_smp_out_2019-08-19.rds")</pre>
   The heated/eq. gas equivalents
```

brrr\_smp\_out <- readRDS("brrr\_smp\_out\_2019-08-20.rds")</pre>

```
14.4 \quad brrr\_best\_dat
best_range <- 1:10
brrr_best_dat <- bind_rows(smp_out, smp_out_uu, brrr_smp_out) %>%
  group_by(exp, smp_D47) %>%
  arrange(smp) %>%
  slice(best_range)
      \operatorname{smp\_out\_comb}
14.5
combine the sims for one plot with faceting
brrr_smp_out_comb <- bind_rows(smp_out, brrr_smp_out, smp_out_uu) %>%
  select(exprow, smp_D47, prop_eth3, exp, smp) %>%
  spread("exp", "smp") %>%
  mutate(diff='UU3 and UU2' - 'ETH-1 and ETH-3')
14.6
      prop eth3 pl
14.6.1 plot best
14.6.2 plot prop
14.6.3 combine and print
brrr_prop_eth3_pl <- brrr_plot_best + brrr_plot_prop + plot_layout(nrow=2, heights=c(...))</pre>
brrr_prop_eth3_pl
```



#### 14.7 calculate some summary statistics for use in-text

```
eth3_to_u2_eth3 <- brrr_best_dat %>%
  ungroup() %>%
  filter(exp=="ETH-1 and ETH-3", smp_D47 >= eth.info$D47[[3]], smp_D47 <= brrr$D47[[1]]
  summarize(mean=mean(smp),
      ci=confidence(smp, n=n()))

eth3_to_u2_u2 <- brrr_best_dat %>%
```

```
ungroup() %>% filter(exp=="UU3 and UU2", smp_D47 >= eth.info$D47[[3]], smp_D47 <= brrr$D47[[1]]) %:
```

```
summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
eth3_to_u2_tempsens <- seq(eth.info$D47[[3]], brrr$D47[[1]], .001) %>%
  tempcal_derivative() %>%
  mean(na.rm=FALSE)
u2_to_0_eth3 <- best_dat %>%
  ungroup() %>%
  filter(exp=="ETH-1 and ETH-3", smp_D47 >= smpinfo$D47[[1]], smp_D47 <= brrr$D47[[1]]
  summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
u2_to_0_u2 <- brrr_best_dat %>%
  ungroup() %>%
  filter(exp=="UU3 and UU2", smp_D47 >= smpinfo$D47[[1]], smp_D47 <= brrr$D47[[1]]) %>5
  summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
u2_to_0_tempsens <- seq(smpinfo$D47[[1]], brrr$D47[[1]], .01) %>%
  tempcal_derivative() %>%
  mean()
eth3_to_0_eth3 <- brrr_best_dat %>%
  ungroup() %>%
  filter(exp=="ETH-1 and ETH-3", smp_D47 \ge eth.info$D47[[3]], smp_D47 \le smp_info$D47[
  summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
eth3_to_0_u2 <- brrr_best_dat %>%
  ungroup() %>%
  filter(exp=="UU3 and UU2", smp_D47 >= eth.info$D47[[3]], smp_D47 <= smpinfo$D47[[1]]
  summarize(mean=mean(smp),
    ci=confidence(smp, n=n()))
eth3_to_0_tempsens <- seq(eth.info$D47[[3]], smpinfo$D47[[1]], .01) %>%
  tempcal_derivative() %>%
  mean()
```

# 14.8 quick check on how much it matters if we add two hypothetical very large-range standards

```
(eth3_to_0_eth3$mean - eth3_to_0_u2$mean) * 1000
    ppm difference, which equates to

# in permil  # in permil  # in permil / degreeC
(eth3_to_0_eth3$mean - eth3_to_0_u2$mean) / eth3_to_0_tempsens
    so... 3 °C improvement for samples between -29.8000392923437
    maybe it's even more at the more extreme end?
(eth3_to_0_eth3$mean - eth3_to_0_u2$mean) / eth3_to_0_tempsens
    degrees improvement for samples between ETH-3 (~20) and 0 °C.
    or in terms of improvement:
((eth3_to_0_eth3$mean / eth3_to_0_u2$mean) - 1) * 100
```