

# SOM4 - Reading and plotting Confocal data

Paixao, Gossa, Marreiros, Gneisinger, Hovers. 2025. Decoding Traces of Acheulean Percussive Technology

2026-01-25 11:38:44.001557

## Brief description of the script

This R markdown document reads, summarizes and plots data for: *Paixao et al. 2025. Decoding Traces of Acheulean Percussive Technology: Experimental Approach to understanding Use-Wear Traces in different raw materials. Submitted to JPA*

The document contains plots of the quantitative surface texture analysis, using Confocal microscopy.

This R project and respective scripts follow the procedures described by Marwick et al. 2017.

The authors would like to thank Ivan Calandra and Lisa Schunk for their help and contribution on several chunks of code included here in the script (pieces of code are also adapted from Calandra et al. 2019, Pedergrana et al. 2020a, 2020b).

To compile this markdown document do not delete or move files from their original folders. Please note that most of the tables and figures in this file do not match the numbering in the paper..

For any questions, comments and inputs, please contact:

Joao Marreiros, joao.marreiros@leiza.de

## Load data into R project

*Imported files are in: './analysis/raw\_data'*

*Figures are saved in: './analysis/plots'*

*Tables are saved in: './analysis/derived\_data'*

```
# Load required libraries
```

```
library(tidyverse)
```

```
## Warning: package 'tibble' was built under R version 4.3.3
```

```
## Warning: package 'tidyr' was built under R version 4.3.1
```

```
## Warning: package 'dplyr' was built under R version 4.3.1
```

```
## Warning: package 'lubridate' was built under R version 4.3.3
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.4      v readr      2.1.6
## v forcats   1.0.1      v stringr   1.6.0
## v ggplot2   4.0.1      v tibble    3.3.0
## v lubridate 1.9.4      v tidyr     1.3.1
## v purrr     1.2.0
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(utils)
library(knitr)
```

```
## Warning: package 'knitr' was built under R version 4.3.3
```

```
library(janitor)
```

```
## Warning: package 'janitor' was built under R version 4.3.3
```

```
##
## Attaching package: 'janitor'
##
## The following objects are masked from 'package:stats':
##
##   chisq.test, fisher.test
```

```
library(kableExtra)
```

```
## Warning: package 'kableExtra' was built under R version 4.3.1
```

```
##
## Attaching package: 'kableExtra'
##
## The following object is masked from 'package:dplyr':
##
##   group_rows
```

```
library(GGally)
library(doBy)
```

```
## Warning: package 'doBy' was built under R version 4.3.3
```

```
##
## Attaching package: 'doBy'
##
## The following object is masked from 'package:dplyr':
##
##   order_by
```

```
library(ggpubr)
library(tools)
library(ggthemes)
```

```
## Warning: package 'ggthemes' was built under R version 4.3.1
```

```
library(FactoMineR)
```

```
## Warning: package 'FactoMineR' was built under R version 4.3.1
```

```
library(factoextra)
```

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
```

```
# See your WD and update the following paths
```

```
# getwd()
```

```
# Load data from .csv
```

```
confocaldata <- read.delim("analysis/data/raw_data/surftext_db.csv", header = T, ";")
```

```
data_file <- list.files("analysis/data/raw_data/surftext_db", pattern = "\\*.csv$", full.names = TRUE)
```

```
md5_in <- md5sum(data_file)
```

```
info_in <- data.frame(file = basename(names(md5_in)), checksum = md5_in, row.names = NULL)
```

# Confocal micro surface texture data

## Import and summarize data

```
# compute descriptive statistics

nminmaxmeanmedsd <- function(x){
  y <- x[!is.na(x)]
  n_test <- length(y)
  min_test <- min(y)
  max_test <- max(y)
  mean_test <- mean(y)
  med_test <- median(y)
  sd_test <- sd(y)
  out <- c(n_test, min_test, max_test, mean_test, med_test, sd_test)
  names(out) <- c("n", "min", "max", "mean", "median", "sd")
  return(out)
}

num.var <- 23:ncol(confocaldata)

confostats <- summaryBy(.-Sample + Worked_material + Raw_material, data=confocaldata[c("Sample", "Worked_material", "Raw_material"), num.var])

write_csv(confostats, "analysis/data/derived_data/confocalstats_arch.csv")

head(confostats, 10)
```

```
##      Sample Worked_material Raw_material Sq.n   Sq.min   Sq.max   Sq.mean
## 1  DWG1S3          bone      Basalt     9  5.063730 10.650124  8.228798
## 2  DWG1S3          no_use     Basalt     3  7.892292 11.631849  9.528152
## 3  DWG3S1          bone      Scoria     9 10.268990 83.562071 27.477733
## 4  DWG3S1          no_use     Scoria     3 14.756695 30.180572 21.990032
## 5  KWG1S1          bone      Glassy_ign  9  4.917584 14.705605  9.591037
## 6  KWG1S1          no_use     Glassy_ign  3  6.177752 12.641063 10.287329
## 7  KWG1S3_         bone      Glassy_ign  8  3.766167  9.883735  7.076967
## 8  KWG1S3_         no_use     Glassy_ign  3  3.577717  6.211463  5.090712
## 9  MW6G2S1         bone      Regular_ign  9  7.503876 21.214706 13.486495
## 10 MW6G2S1         no_use     Regular_ign  3  8.210394 12.331760  9.594486
##      Sq.median   Sq.sd Ssk.n   Ssk.min   Ssk.max   Ssk.mean Ssk.median
## 1  8.648178  1.939212   9 -0.02251418  0.7577052  0.28357010  0.27415163
## 2  9.060314  1.913172   3 -0.14897681  0.1818577  0.02267047  0.03513049
## 3 22.219930 22.258847   9  0.37276137  2.1128103  1.02777103  0.91055760
## 4 21.032828  7.756364   3  0.42534413  1.0881790  0.68413355  0.53887753
## 5  9.813930  3.267261   9 -0.38810870  1.4721812  0.67796861  0.57816045
## 6 12.043172  3.571531   3  0.01244696  0.1114047  0.06844835  0.08149343
## 7  8.078204  2.560580   8 -0.04934925  3.6317140  1.38958223  0.80416487
## 8  5.482956  1.359980   3  0.38477194  1.7332947  0.97502239  0.80700054
## 9 13.625286  3.783788   9 -0.19782916  0.6546132  0.34455717  0.30583174
## 10 8.241303  2.370600   3  1.06912028  1.3486471  1.18832836  1.14721768
##      Ssk.sd Sku.n  Sku.min  Sku.max Sku.mean Sku.median  Sku.sd Sp.n
## 1  0.25123325   9  2.645284  4.479417  3.494612   3.314367  0.5977422   9
```

## 2	0.16576885	3	2.757541	3.638224	3.271035	3.417338	0.4582075	3
## 3	0.55292390	9	2.519341	11.163102	6.626228	5.238011	3.0284775	9
## 4	0.35448836	3	3.100749	6.406618	4.667123	4.494004	1.6597199	3
## 5	0.62423622	9	3.262983	10.195028	6.539901	6.846613	2.6503588	9
## 6	0.05075221	3	2.607873	14.630243	6.764393	3.055065	6.8156937	3
## 7	1.43450733	8	3.223886	22.919378	9.667594	5.949720	8.0346084	8
## 8	0.68978395	3	5.404446	9.352493	7.393515	7.423604	1.9741956	3
## 9	0.28236034	9	2.838261	5.931765	4.039210	3.886268	1.0237456	9
## 10	0.14422683	3	5.035714	7.818396	6.313458	6.086265	1.4051838	3
##	Sp.min	Sp.max	Sp.mean	Sp.median	Sp.sd	Sv.n	Sv.min	Sv.max
## 1	17.18378	39.87515	25.68341	24.58514	6.667521	9	14.97265	34.40767
## 2	23.64134	34.89050	31.01266	34.50615	6.386646	3	25.56008	47.23796
## 3	40.30089	199.88299	106.25727	107.04106	47.219908	9	34.78964	143.30958
## 4	53.13381	101.14910	84.18791	98.28082	26.931854	3	36.56090	102.94767
## 5	17.78106	81.18747	48.89975	45.94380	19.788620	9	15.02573	59.00247
## 6	19.33636	72.59322	42.22180	34.73584	27.406262	3	17.25796	85.60943
## 7	17.26470	64.96520	38.06369	37.78137	18.931496	8	12.67903	32.49597
## 8	16.19630	33.12118	27.17700	32.21352	9.520390	3	13.03301	22.48445
## 9	22.75775	70.05513	52.00632	55.02275	14.510714	9	20.23667	58.05026
## 10	39.76457	62.86922	50.26186	48.15180	11.695957	3	23.72950	30.46131
##	Sv.mean	Sv.median	Sv.sd	Sz.n	Sz.min	Sz.max	Sz.mean	Sz.median
## 1	25.80182	27.13677	6.691041	9	32.72492	67.01192	51.48523	53.26250
## 2	33.59354	27.98257	11.878333	3	49.20142	81.74410	64.60620	62.87308
## 3	81.95437	80.99841	37.894035	9	75.09053	343.19257	188.21164	177.18170
## 4	69.63100	69.38444	33.194071	3	89.69471	204.09678	153.81891	167.66526
## 5	35.12224	33.15965	16.559462	9	32.80680	120.18495	84.02198	92.68390
## 6	50.06462	47.32647	34.257906	3	36.59432	158.20266	92.28643	82.06231
## 7	20.65125	21.44169	6.930091	8	30.35188	88.82715	58.71495	62.50386
## 8	16.52252	14.05009	5.188172	3	29.22931	55.60564	43.69952	46.26361
## 9	43.68278	51.62774	13.521014	9	42.99442	128.10540	95.68911	100.35808
## 10	27.34249	27.83666	3.393004	3	63.49407	90.70588	77.60435	78.61311
##	Sz.sd	Sa.n	Sa.min	Sa.max	Sa.mean	Sa.median	Sa.sd	
## 1	11.05407	9	4.021936	8.297740	6.396206	6.668945	1.5022141	
## 2	16.34042	3	6.303812	9.215386	7.590523	7.252370	1.4849502	
## 3	79.49558	9	7.600239	68.757057	20.352121	16.748056	18.8435653	
## 4	58.44441	3	11.828620	22.091625	16.325167	15.055256	5.2480304	
## 5	32.05592	9	3.880155	10.724066	6.981453	6.847019	2.4034272	
## 6	61.44548	3	4.900318	10.324936	7.611112	7.608081	2.7123100	
## 7	23.44636	8	2.922386	7.799800	4.699231	4.765441	1.6466088	
## 8	13.37381	3	2.551167	4.348197	3.563063	3.789825	0.9197255	
## 9	26.73483	9	6.049680	15.821603	10.214988	10.173646	2.7191844	
## 10	13.63392	3	5.962861	9.005989	7.071227	6.244832	1.6814735	
##	Smr...c...1.µm.below.highest.peak..n						Smr...c...1.µm.below.highest.peak..min	
## 1							9	0.129436652
## 2							3	0.136036764
## 3							9	0.003326797
## 4							3	0.010260262
## 5							9	0.003517030
## 6							3	0.003298784
## 7							8	0.026062347
## 8							3	0.019764275
## 9							9	0.002864635
## 10							3	0.014145319
##	Smr...c...1.µm.below.highest.peak..max						Smr...c...1.µm.below.highest.peak..mean	

## 1	0.25093810	0.21134175		
## 2	0.17077097	0.15171297		
## 3	0.09686346	0.03621447		
## 4	0.02731155	0.01631209		
## 5	0.11591567	0.02776300		
## 6	0.18030579	0.09874595		
## 7	0.16980952	0.09269989		
## 8	0.17915327	0.08656442		
## 9	0.19302423	0.05522177		
## 10	0.03707106	0.02264030		
##	Smr..c...1.µm.below.highest.peak..median			
## 1	0.23035046			
## 2	0.14833118			
## 3	0.01962398			
## 4	0.01136444			
## 5	0.01521358			
## 6	0.11263326			
## 7	0.08514802			
## 8	0.06077572			
## 9	0.02429787			
## 10	0.01670453			
##	Smr..c...1.µm.below.highest.peak..sd	Smc..p...10...n	Smc..p...10...min	
## 1	0.040494607	9	6.691335	
## 2	0.017612314	3	10.320923	
## 3	0.034977543	9	11.058562	
## 4	0.009541802	3	19.783437	
## 5	0.034533409	9	6.171449	
## 6	0.089316926	3	8.294266	
## 7	0.053584958	8	4.493045	
## 8	0.082764762	3	4.059458	
## 9	0.059919024	9	9.722711	
## 10	0.012562742	3	8.844100	
##	Smc..p...10...max	Smc..p...10...mean	Smc..p...10...median	Smc..p...10...sd
## 1	14.477477	10.938276	11.530379	2.6523625
## 2	14.746180	12.156261	11.401681	2.3071131
## 3	141.549348	35.649246	23.214725	40.8940784
## 4	40.863905	27.474072	21.774875	11.6386071
## 5	15.041786	10.344946	10.055254	3.4315575
## 6	16.366388	11.981739	11.284563	4.0809719
## 7	13.007607	6.672026	4.940842	3.1507142
## 8	5.908241	5.163481	5.522745	0.9753476
## 9	28.779154	16.539688	14.991565	5.3545996
## 10	15.447508	11.661309	10.692319	3.4066782
##	Sxp..p...50...q...97.5...n	Sxp..p...50...q...97.5...min		
## 1	9	8.126920		
## 2	3	14.932044		
## 3	9	14.583036		
## 4	3	24.598504		
## 5	9	9.402009		
## 6	3	12.281577		
## 7	8	6.770593		
## 8	3	6.964112		
## 9	9	14.431524		
## 10	3	10.444185		

##	Sxp..p...50...q...97.5...max	Sxp..p...50...q...97.5...mean		
## 1	19.80979	14.077415		
## 2	20.19767	17.349271		
## 3	80.58646	38.333654		
## 4	52.43185	37.206342		
## 5	26.08240	17.380564		
## 6	24.75996	17.460145		
## 7	16.61239	11.476646		
## 8	12.32030	8.998026		
## 9	38.99502	24.784749		
## 10	16.56297	13.559731		
##	Sxp..p...50...q...97.5...median	Sxp..p...50...q...97.5...sd	Sal..s...0.2..n	
## 1	14.060023	3.964723	9	
## 2	16.918101	2.659160	3	
## 3	35.912080	21.638165	9	
## 4	34.588675	14.100103	3	
## 5	16.220668	5.892286	9	
## 6	15.338894	6.504022	3	
## 7	11.762347	3.591365	8	
## 8	7.709669	2.901219	3	
## 9	24.298570	7.099866	9	
## 10	13.672034	3.060939	3	
##	Sal..s...0.2..min	Sal..s...0.2..max	Sal..s...0.2..mean	Sal..s...0.2..median
## 1	52.80412	89.81858	72.55206	67.45247
## 2	61.91879	86.93064	71.14821	64.59520
## 3	46.39301	105.05251	63.17700	60.01913
## 4	48.17865	71.49917	59.97350	60.24269
## 5	42.20291	73.66658	60.09439	60.29915
## 6	37.49694	89.63585	66.61831	72.72213
## 7	32.42020	79.58032	55.62719	60.45598
## 8	37.19822	48.70143	44.04602	46.23842
## 9	46.28510	84.34683	57.68914	53.14922
## 10	38.91076	54.21957	47.81911	50.32702
##	Sal..s...0.2..sd	Std..Reference.angle...0...n	Std..Reference.angle...0...min	
## 1	12.775287	9	3.253287	
## 2	13.733340	3	2.757417	
## 3	16.723446	9	16.500067	
## 4	11.662592	3	3.753496	
## 5	9.121335	9	73.251527	
## 6	26.599981	3	20.000148	
## 7	16.115243	8	16.493768	
## 8	6.056890	3	39.252011	
## 9	12.802860	9	58.002030	
## 10	7.956576	3	7.759671	
##	Std..Reference.angle...0...max	Std..Reference.angle...0...mean		
## 1	174.24867	60.27726		
## 2	86.24629	47.66927		
## 3	167.25349	63.13998		
## 4	98.74818	49.99662		
## 5	173.99388	114.08134		
## 6	73.25047	47.83506		
## 7	100.00630	76.71804		
## 8	93.51012	73.17181		
## 9	137.74483	111.97084		

## 10			163.00255			81.25326	
##	Std..Reference.angle...0...median	Std..Reference.angle...0...sd	Sdq.n				
## 1			22.50332			63.76333	9
## 2			54.00410			42.10339	3
## 3			50.50421			49.60585	9
## 4			47.48818			47.54699	3
## 5			98.49042			36.70735	9
## 6			50.25455			26.70748	3
## 7			79.99869			26.26219	8
## 8			86.75329			29.56904	3
## 9			120.50400			27.26386	9
## 10			72.99756			77.95002	3
##	Sdq.min	Sdq.max	Sdq.mean	Sdq.median	Sdq.sd	Sdr.n	Sdr.min
## 1	0.5884395	0.9309811	0.7006765	0.7016803	0.09898725	9	13.713395
## 2	0.6757145	0.9302041	0.8183018	0.8489867	0.12999003	3	17.018494
## 3	1.0971752	4.9937168	2.6542813	2.2442055	1.24940342	9	30.359955
## 4	1.3396451	2.4753302	2.0027938	2.1934061	0.59135014	3	26.424201
## 5	0.5672499	1.3066440	0.9607505	0.9870976	0.23254134	9	11.363945
## 6	0.6395680	1.3057172	0.9853017	1.0106198	0.33379552	3	14.090017
## 7	0.4935876	1.3520065	0.8000794	0.6770484	0.35102302	8	9.142846
## 8	0.5035897	0.7846100	0.6550251	0.6768755	0.14177866	3	9.750539
## 9	0.6536768	2.0432940	1.3418718	1.2713830	0.41681690	9	15.945458
## 10	0.9446065	1.4581162	1.1680681	1.1014818	0.26315082	3	23.386377
##	Sdr.max	Sdr.mean	Sdr.median	Sdr.sd	Vm..p...10...n	Vm..p...10...min	
## 1	23.84170	17.68615	17.97856	2.955713	9	0.3201235	
## 2	27.12620	22.19458	22.43903	5.058284	3	0.3489735	
## 3	104.27514	53.93267	45.30116	26.073466	9	0.8999336	
## 4	61.04453	44.88199	47.17725	17.423919	3	0.8142877	
## 5	26.97951	20.42928	21.22632	5.333596	9	0.2836160	
## 6	28.43362	23.38228	27.62321	8.057532	3	0.3003648	
## 7	25.44176	15.97144	15.02754	6.137314	8	0.2254618	
## 8	18.28971	13.78189	13.30541	4.289479	3	0.2927578	
## 9	54.82298	33.40845	32.01672	11.438759	9	0.3687425	
## 10	39.18970	30.68873	29.49013	7.969549	3	0.6872419	
##	Vm..p...10...max	Vm..p...10...mean	Vm..p...10...median	Vm..p...10...sd			
## 1	0.7716772		0.4867815	0.4420677		0.13705388	
## 2	0.5319171		0.4362017	0.4277146		0.09176663	
## 3	3.2005335		2.0836236	2.2646496		0.76888228	
## 4	2.2955337		1.7731366	2.2095883		0.83149869	
## 5	1.4529470		0.7806336	0.7299465		0.39975484	
## 6	1.1006135		0.6477983	0.5424166		0.41040036	
## 7	1.4390420		0.7262197	0.5971844		0.47004765	
## 8	0.6136174		0.4874212	0.5558884		0.17103664	
## 9	1.5783698		0.9547482	0.8878558		0.38241026	
## 10	1.1523588		0.8508590	0.7129763		0.26142328	
##	Vv..p...10...n	Vv..p...10...min	Vv..p...10...max	Vv..p...10...mean			
## 1	9	7.011406	15.065817	11.425060			
## 2	3	10.669988	15.278133	12.592513			
## 3	9	11.958407	143.821705	37.733038			
## 4	3	20.597814	43.159485	29.247207			
## 5	9	6.455019	16.493383	11.125575			
## 6	3	8.594619	16.908842	12.629500			
## 7	8	4.837232	13.619321	7.398182			
## 8	3	4.352180	6.464198	5.650918			

## 9	9	10.091505	30.357493	17.494477
## 10	3	9.557139	16.599846	12.512158
##	Vv..p...10...median	Vv..p...10...sd	Vmp..p...10...n	Vmp..p...10...min
## 1	12.141650	2.754885	9	0.3201235
## 2	11.829417	2.396974	3	0.3489735
## 3	25.821679	41.100204	9	0.8999336
## 4	23.984323	12.166787	3	0.8142877
## 5	10.871803	3.680743	9	0.2836160
## 6	12.385040	4.162499	3	0.3003648
## 7	5.883944	3.118801	8	0.2254618
## 8	6.136375	1.136621	3	0.2927578
## 9	16.419288	5.652112	9	0.3687425
## 10	11.379487	3.655426	3	0.6872419
##	Vmp..p...10...max	Vmp..p...10...mean	Vmp..p...10...median	Vmp..p...10...sd
## 1	0.7716772	0.4867815	0.4420677	0.13705388
## 2	0.5319171	0.4362017	0.4277146	0.09176663
## 3	3.2005335	2.0836236	2.2646496	0.76888228
## 4	2.2955337	1.7731366	2.2095883	0.83149869
## 5	1.4529470	0.7806336	0.7299465	0.39975484
## 6	1.1006135	0.6477983	0.5424166	0.41040036
## 7	1.4390420	0.7262197	0.5971844	0.47004765
## 8	0.6136174	0.4874212	0.5558884	0.17103664
## 9	1.5783698	0.9547482	0.8878558	0.38241026
## 10	1.1523588	0.8508590	0.7129763	0.26142328
##	Vmc..p...10...q...80...n	Vmc..p...10...q...80...min		
## 1	9	4.450422		
## 2	3	7.467709		
## 3	9	7.976398		
## 4	3	13.587119		
## 5	9	4.397925		
## 6	3	5.169091		
## 7	8	3.102471		
## 8	3	2.505969		
## 9	9	6.893585		
## 10	3	6.229499		
##	Vmc..p...10...q...80...max	Vmc..p...10...q...80...mean		
## 1	9.592246	7.096468		
## 2	10.260280	8.678579		
## 3	63.894370	19.246766		
## 4	20.738619	16.329152		
## 5	11.618668	7.257442		
## 6	11.552078	8.068189		
## 7	8.515048	4.572191		
## 8	4.488539	3.536744		
## 9	16.386982	11.232513		
## 10	9.091423	7.238928		
##	Vmc..p...10...q...80...median	Vmc..p...10...q...80...sd		
## 1	7.003398	1.8211751		
## 2	8.307747	1.4327421		
## 3	14.478823	17.4102465		
## 4	14.661717	3.8563247		
## 5	6.779995	2.5811942		
## 6	7.483400	3.2314262		
## 7	3.820601	1.8489941		

## 8	3.615725	0.9936424		
## 9	10.824962	2.6821851		
## 10	6.395861	1.6064626		
##	Vvc..p...10...q...80...n	Vvc..p...10...q...80...min		
## 1	9	6.578057		
## 2	3	9.829754		
## 3	9	10.950992		
## 4	3	19.388028		
## 5	9	5.947101		
## 6	3	7.878771		
## 7	8	4.306807		
## 8	3	3.885029		
## 9	9	9.332079		
## 10	3	8.780407		
##	Vvc..p...10...q...80...max	Vvc..p...10...q...80...mean		
## 1	14.144531	10.612759		
## 2	14.017012	11.581646		
## 3	138.758852	35.409369		
## 4	40.128179	27.149625		
## 5	15.242880	10.082201		
## 6	15.519505	11.519225		
## 7	12.731895	6.697255		
## 8	5.705291	5.088036		
## 9	28.170450	16.082099		
## 10	15.689234	11.767761		
##	Vvc..p...10...q...80...median	Vvc..p...10...q...80...sd	Vvv..p...80...n	
## 1	11.160474	2.577457	9	
## 2	10.898172	2.175692	3	
## 3	23.018057	39.961589	9	
## 4	21.932669	11.311540	3	
## 5	9.691994	3.370331	9	
## 6	11.159399	3.833055	3	
## 7	5.146900	2.962287	8	
## 8	5.673789	1.041954	3	
## 9	14.489728	5.314823	9	
## 10	10.833641	3.547874	3	
##	Vvv..p...80...min	Vvv..p...80...max	Vvv..p...80...mean	Vvv..p...80...median
## 1	0.4333487	1.1209188	0.8123011	0.9146472
## 2	0.8402340	1.2611208	1.0108667	0.9312452
## 3	0.9118226	5.0628535	2.3236692	1.7662483
## 4	1.2097860	3.0313052	2.0975818	2.0516542
## 5	0.5079180	1.5227824	1.0433743	1.1798090
## 6	0.7158476	1.3893372	1.1102753	1.2256412
## 7	0.3750757	0.9281178	0.7009262	0.7370450
## 8	0.4625862	0.7589066	0.5628813	0.4671511
## 9	0.7594260	2.1870439	1.4123781	1.3787802
## 10	0.5458456	0.9106123	0.7443969	0.7767329
##	Vvv..p...80...sd	Maximum.depth.of.furrows.n	Maximum.depth.of.furrows.min	
## 1	0.2172396	9	16.75276	
## 2	0.2214523	3	26.26104	
## 3	1.4025854	9	31.68333	
## 4	0.9116277	3	43.47218	
## 5	0.3409910	9	18.23597	
## 6	0.3512535	3	18.99557	

## 7	0.2143057	8	15.77026
## 8	0.1697782	3	17.81542
## 9	0.4705498	9	20.92165
## 10	0.1845207	3	28.94002
##	Maximum.depth.of.furrows.max	Maximum.depth.of.furrows.mean	
## 1	28.83774	21.54506	
## 2	31.86601	29.10138	
## 3	157.73925	91.79489	
## 4	94.90850	68.66226	
## 5	43.65327	30.76170	
## 6	52.50780	38.57436	
## 7	43.72077	25.72067	
## 8	21.53994	20.02998	
## 9	81.78266	48.70766	
## 10	46.65155	36.33151	
##	Maximum.depth.of.furrows.median	Maximum.depth.of.furrows.sd	
## 1	21.92627	3.686675	
## 2	29.17709	2.803250	
## 3	73.12573	44.307745	
## 4	67.60612	25.734420	
## 5	31.80954	8.295442	
## 6	44.21971	17.454793	
## 7	26.24559	9.492498	
## 8	20.73457	1.959682	
## 9	47.11639	19.231349	
## 10	33.40297	9.211778	
##	Mean.depth.of.furrows.n	Mean.depth.of.furrows.min	Mean.depth.of.furrows.max
## 1	9	6.002999	9.051419
## 2	3	7.461272	8.622203
## 3	9	11.185208	46.111107
## 4	3	15.187073	29.205054
## 5	9	5.064537	12.064219
## 6	3	5.696616	10.593519
## 7	8	4.347303	9.174991
## 8	3	4.664563	6.692360
## 9	9	7.641667	23.177453
## 10	3	9.069816	17.681250
##	Mean.depth.of.furrows.mean	Mean.depth.of.furrows.median	
## 1	7.843046	8.187588	
## 2	7.852543	7.474154	
## 3	26.426294	26.784884	
## 4	23.349417	25.656125	
## 5	8.885741	7.902523	
## 6	8.656570	9.679574	
## 7	6.845322	6.933926	
## 8	5.459767	5.022376	
## 9	13.799330	12.461877	
## 10	12.941694	12.074016	
##	Mean.depth.of.furrows.sd	Mean.density.of.furrows.n	
## 1	0.9466227	9	
## 2	0.6665760	3	
## 3	13.6833953	9	
## 4	7.2881153	3	
## 5	2.5903413	9	

## 6	2.6038085	3	
## 7	1.7637038	8	
## 8	1.0823459	3	
## 9	4.2665737	9	
## 10	4.3707951	3	
##	Mean.density.of.furrows.min	Mean.density.of.furrows.max	
## 1	2298.821	2605.148	
## 2	2366.364	2522.769	
## 3	1791.625	2919.208	
## 4	2297.497	2494.644	
## 5	2643.277	2901.836	
## 6	2564.022	2924.246	
## 7	2680.364	2971.561	
## 8	2669.025	2869.392	
## 9	2386.039	2750.615	
## 10	2378.525	2699.401	
##	Mean.density.of.furrows.mean	Mean.density.of.furrows.median	
## 1	2432.042	2411.718	
## 2	2449.421	2459.131	
## 3	2439.866	2475.756	
## 4	2389.146	2375.297	
## 5	2781.130	2792.991	
## 6	2718.721	2667.893	
## 7	2836.581	2842.739	
## 8	2783.521	2812.145	
## 9	2495.761	2453.730	
## 10	2519.780	2481.414	
##	Mean.density.of.furrows.sd	First.direction.n	First.direction.min
## 1	90.81019	9	0.013088863
## 2	78.65352	3	0.017821948
## 3	348.19673	9	0.007610828
## 4	99.30027	3	0.007410580
## 5	90.04938	9	0.022110916
## 6	185.41275	3	0.009692957
## 7	89.66514	8	0.019557202
## 8	103.20532	3	44.976009260
## 9	116.42962	9	0.009121793
## 10	163.84231	3	89.994521960
##	First.direction.max	First.direction.mean	First.direction.median
## 1	179.99750	99.99916	90.00905
## 2	89.99993	48.75543	56.24853
## 3	135.00784	62.94580	89.98562
## 4	179.99790	75.00878	45.02103
## 5	134.97617	80.00343	90.00258
## 6	90.01321	45.01000	45.00709
## 7	135.00618	84.37119	89.99210
## 8	90.01678	74.99408	89.98945
## 9	135.02644	95.89072	116.49217
## 10	179.99868	120.00121	90.01043
##	First.direction.sd	Second.direction.n	Second.direction.min
## 1	62.74469	9	0.002585364
## 2	45.45663	3	0.006468474
## 3	47.03073	9	0.002699942
## 4	93.66746	3	90.002718820

## 5	37.49037	9	44.985288150
## 6	45.00176	3	26.457376620
## 7	44.58743	8	0.001381717
## 8	25.99642	3	89.979640140
## 9	46.49592	9	0.002614089
## 10	51.95933	3	63.531150460
##	Second.direction.max	Second.direction.mean	Second.direction.median
## 1	179.99306	119.99597	135.00156
## 2	179.99690	104.99997	134.99653
## 3	179.99947	114.11516	153.50994
## 4	134.99764	119.99888	134.99629
## 5	161.50171	95.00381	90.01268
## 6	89.99463	59.99360	63.52880
## 7	179.98498	100.33798	125.73907
## 8	135.00017	119.99310	134.99950
## 9	135.01137	98.74237	116.44870
## 10	179.98588	126.16501	134.97800
##	Second.direction.sd	Third.direction.n	Third.direction.min
## 1	74.61905	9	2.648285e+01
## 2	93.66955	3	4.499814e+01
## 3	70.07597	9	1.849645e+01
## 4	25.97744	3	3.375555e+01
## 5	47.08987	9	2.750789e-02
## 6	31.91581	3	8.351778e-03
## 7	66.92108	8	4.500393e+01
## 8	25.99242	3	1.164543e+02
## 9	44.75928	9	7.203658e-03
## 10	58.72544	3	1.137504e-02
##	Third.direction.max	Third.direction.mean	Third.direction.median
## 1	134.99924	65.88912	45.02250
## 2	63.51498	51.17054	44.99851
## 3	179.99760	99.99419	90.00803
## 4	135.00765	86.25359	89.99757
## 5	171.76576	84.08464	71.48723
## 6	179.98774	81.17287	63.52252
## 7	135.02127	108.77886	125.73425
## 8	179.96919	143.81446	135.01991
## 9	108.53240	47.50569	45.00770
## 10	45.01071	30.00515	44.99337
##	Third.direction.sd	Texture.isotropy.n	Texture.isotropy.min
## 1	36.96718	9	47.13495
## 2	10.69059	3	70.77438
## 3	61.76833	9	26.13837
## 4	50.72978	3	69.61484
## 5	58.77415	9	11.64210
## 6	91.27867	3	38.21673
## 7	34.08861	8	26.78453
## 8	32.65798	3	70.47470
## 9	43.22885	9	30.58442
## 10	25.97537	3	85.89070
##	Texture.isotropy.max	Texture.isotropy.mean	Texture.isotropy.median
## 1	84.73920	72.46804	75.69371
## 2	75.44572	72.89308	72.45914
## 3	92.30589	71.54840	77.46036

## 4	79.61672	75.63799	77.68242
## 5	86.22843	58.60758	60.14503
## 6	84.87242	68.72778	83.09420
## 7	86.68026	67.59308	79.56700
## 8	83.74952	76.51237	75.31288
## 9	86.74423	63.83621	69.65782
## 10	92.25443	88.80248	88.26233
##	Texture.isotropy.sd Length.scale.anisotropy..Sfrax...epLsar..n		
## 1	11.551649		9
## 2	2.365707		3
## 3	19.306849		9
## 4	5.305107		3
## 5	23.702520		9
## 6	26.438298		3
## 7	24.419322		8
## 8	6.718207		3
## 9	20.271386		9
## 10	3.216065		3
##	Length.scale.anisotropy..Sfrax...epLsar..min		
## 1		0.0003036412	
## 2		0.0004903408	
## 3		0.0002322810	
## 4		0.0006145745	
## 5		0.0005041829	
## 6		0.0002173764	
## 7		0.0007216012	
## 8		0.0013732855	
## 9		0.0006712386	
## 10		0.0004264319	
##	Length.scale.anisotropy..Sfrax...epLsar..max		
## 1		0.001792291	
## 2		0.001118900	
## 3		0.006755924	
## 4		0.003863928	
## 5		0.005463414	
## 6		0.003835056	
## 7		0.004121158	
## 8		0.003495790	
## 9		0.003949583	
## 10		0.002831267	
##	Length.scale.anisotropy..Sfrax...epLsar..mean		
## 1		0.0010222791	
## 2		0.0008424768	
## 3		0.0031057825	
## 4		0.0021953597	
## 5		0.0026060333	
## 6		0.0014276674	
## 7		0.0024981068	
## 8		0.0021015144	
## 9		0.0015971049	
## 10		0.0018796506	
##	Length.scale.anisotropy..Sfrax...epLsar..median		
## 1		0.0009220201	
## 2		0.0009181891	

```

## 3 0.0022652921
## 4 0.0021075769
## 5 0.0026682883
## 6 0.0002305694
## 7 0.0027324841
## 8 0.0014354676
## 9 0.0014998304
## 10 0.0023812526
## Length.scale.anisotropy..Sfrax...epLsar..sd
## 1 0.0005312633
## 2 0.0003210468
## 3 0.0022981164
## 4 0.0016264542
## 5 0.0014524109
## 6 0.0020848704
## 7 0.0011636999
## 8 0.0012078783
## 9 0.0010293285
## 10 0.0012784803
## Length.scale.anisotropy..NewEplsar..n
## 1 9
## 2 3
## 3 9
## 4 3
## 5 9
## 6 3
## 7 8
## 8 3
## 9 9
## 10 3
## Length.scale.anisotropy..NewEplsar..min
## 1 0.01768289
## 2 0.01778222
## 3 0.01755183
## 4 0.01756227
## 5 0.01713069
## 6 0.01746231
## 7 0.01687653
## 8 0.01719091
## 9 0.01701070
## 10 0.01777294
## Length.scale.anisotropy..NewEplsar..max
## 1 0.01834995
## 2 0.01815137
## 3 0.01910737
## 4 0.01805707
## 5 0.01795282
## 6 0.01772013
## 7 0.01777504
## 8 0.01749688
## 9 0.01801316
## 10 0.01783933
## Length.scale.anisotropy..NewEplsar..mean
## 1 0.01803226

```

```

## 2          0.01798564
## 3          0.01811044
## 4          0.01775986
## 5          0.01754005
## 6          0.01759671
## 7          0.01731615
## 8          0.01734158
## 9          0.01744137
## 10         0.01780281
## Length.scale.anisotropy..NewEplsar..median
## 1          0.01807702
## 2          0.01802332
## 3          0.01811201
## 4          0.01766025
## 5          0.01756788
## 6          0.01760767
## 7          0.01729995
## 8          0.01733696
## 9          0.01746496
## 10         0.01779615
## Length.scale.anisotropy..NewEplsar..sd Fractal.complexity..Asfc..n
## 1          1.930067e-04          9
## 2          1.874365e-04          3
## 3          5.325728e-04          9
## 4          2.620097e-04          3
## 5          2.960458e-04          9
## 6          1.292601e-04          3
## 7          2.785530e-04          8
## 8          1.530368e-04          3
## 9          2.718568e-04          9
## 10         3.369016e-05          3
## Fractal.complexity..Asfc..min Fractal.complexity..Asfc..max
## 1          17.90917          32.86133
## 2          21.83468          36.96163
## 3          31.53697          144.81210
## 4          42.84526          73.08920
## 5          15.14748          28.65736
## 6          18.86207          30.70305
## 7          12.26438          34.95146
## 8          12.24780          24.30124
## 9          17.52856          55.94547
## 10         27.36583          48.41544
## Fractal.complexity..Asfc..mean Fractal.complexity..Asfc..median
## 1          23.89955          22.93562
## 2          29.17703          28.73478
## 3          76.20627          70.56423
## 4          62.94505          72.90068
## 5          22.18242          21.91921
## 6          26.36417          29.52741
## 7          20.52695          16.77779
## 8          17.64458          16.38469
## 9          39.47646          43.20929
## 10         38.96393          41.11050
## Fractal.complexity..Asfc..sd Scale.of.max.complexity..Smfc..n

```

## 1	4.213738	9	
## 2	7.573168	3	
## 3	39.032757	9	
## 4	17.407177	3	
## 5	4.584811	9	
## 6	6.523553	3	
## 7	9.164681	8	
## 8	6.124693	3	
## 9	13.388156	9	
## 10	10.687724	3	
##	Scale.of.max.complexity..Smfc..min	Scale.of.max.complexity..Smfc..max	
## 1	19.521476	46.25603	
## 2	14.265081	39.54113	
## 3	415.754830	9576.82518	
## 4	842.119572	2953.43532	
## 5	28.894201	2334.25686	
## 6	21.114085	175.46142	
## 7	5.566202	985.12871	
## 8	15.428860	24.69969	
## 9	24.699688	985.12871	
## 10	162.226585	415.75483	
##	Scale.of.max.complexity..Smfc..mean	Scale.of.max.complexity..Smfc..median	
## 1	30.87067	31.25146	
## 2	29.20241	33.80103	
## 3	3044.12464	2730.66144	
## 4	1833.76154	1705.72973	
## 5	386.50461	118.54510	
## 6	92.22228	80.09134	
## 7	168.05218	28.31882	
## 8	19.88334	19.52148	
## 9	465.86737	486.35851	
## 10	293.92973	303.80777	
##	Scale.of.max.complexity..Smfc..sd	HAsfc9..HAsfc9..n	HAsfc9..HAsfc9..min
## 1	7.067708	9	0.1316330
## 2	13.250666	3	0.1255464
## 3	2965.415927	9	0.4455791
## 4	1061.464860	3	0.5851555
## 5	742.721750	9	0.2710966
## 6	77.885458	3	0.2331469
## 7	337.614585	8	0.2132304
## 8	4.645996	3	0.2399171
## 9	305.964329	9	0.1307847
## 10	127.052448	3	0.4507234
##	HAsfc9..HAsfc9..max	HAsfc9..HAsfc9..mean	HAsfc9..HAsfc9..median
## 1	0.2981468	0.1855606	0.1487189
## 2	0.2493303	0.2006154	0.2269694
## 3	13.3993886	2.6339071	1.1790757
## 4	1.6127738	0.9869700	0.7629809
## 5	6.1967259	1.1809769	0.5654418
## 6	0.9208822	0.4626529	0.2339297
## 7	5.2171286	0.9908077	0.3453896
## 8	0.5555807	0.3814661	0.3489006
## 9	1.9469006	0.8496816	0.7590458
## 10	0.7342881	0.6092278	0.6426717

```

##      HAsfc9..HAsfc9..sd HAsfc81..HAsfc81..n HAsfc81..HAsfc81..min
## 1          0.06507991          9          0.2502991
## 2          0.06596603          3          0.2922487
## 3          4.12872799          9          -Inf
## 4          0.54920687          3          3.2992331
## 5          1.89658947          9          0.5763855
## 6          0.39683837          3          0.4197154
## 7          1.71734602          8          0.3835358
## 8          0.16033175          3          0.5153154
## 9          0.66391685          9          0.4168093
## 10         0.14471043          3          0.7749490
##      HAsfc81..HAsfc81..max HAsfc81..HAsfc81..mean HAsfc81..HAsfc81..median
## 1          0.4765597          0.3146768          0.2906821
## 2          0.3858270          0.3305777          0.3136574
## 3          29.9869968          -Inf          3.0279886
## 4          7.1479098          5.1953535          5.1389175
## 5          3.8599229          1.5651485          1.0658350
## 6          1.6018539          0.8442619          0.5112164
## 7          13.4263962          2.5956076          0.6508434
## 8          0.6441176          0.5735222          0.5611335
## 9          3.2954504          1.5500254          1.4498423
## 10         1.2744163          1.0568059          1.1210524
##      HAsfc81..HAsfc81..sd
## 1          0.08511859
## 2          0.04903006
## 3          NaN
## 4          1.92495890
## 5          1.19102556
## 6          0.65768710
## 7          4.49255308
## 8          0.06528869
## 9          0.93993842
## 10         0.25585662

```

Plot all paramaters organised by Raw material

```

# Loop for plotting all surface texture parameters

for (i in num.var) {

  cat("[", i, " ] ", names(confocaldata)[i], "\n", sep = "")

  p <- ggplot(
    confocaldata,
    aes(
      x = Raw_material,
      y = .data[[names(confocaldata)[i]]]
    )
  ) +
  # Neutral boxplot (no categorical encoding)
  geom_boxplot(

```

```

colour = "grey30",
fill   = "grey90",
width  = 0.6,
outlier.shape = NA
) +

# Overlay points with SAME encoding as scatter plots
geom_jitter(
  aes(
    colour = Raw_material,
    shape  = Worked_material
  ),
  width  = 0.15,
  size   = 1.5,
  stroke = 1,
  alpha  = 0.85
) +

theme_classic() +
labs(
  x    = "Raw material",
  y    = gsub("\\.", " ", names(confocaldata)[i]),
  colour = "Raw material",
  shape  = "Worked material"
) +

# Colour-blind-safe palette
scale_colour_brewer(palette = "Dark2") +

# Shape mapping (same as scatter plots)
scale_shape_manual(
  values = c(
    bone    = 1, # open circle
    no_use  = 17 # filled triangle
  )
)

file_out <- paste0(
  file_path_sans_ext(info_in[["file"]]),
  "_plot_",
  names(confocaldata)[i],
  ".pdf"
)

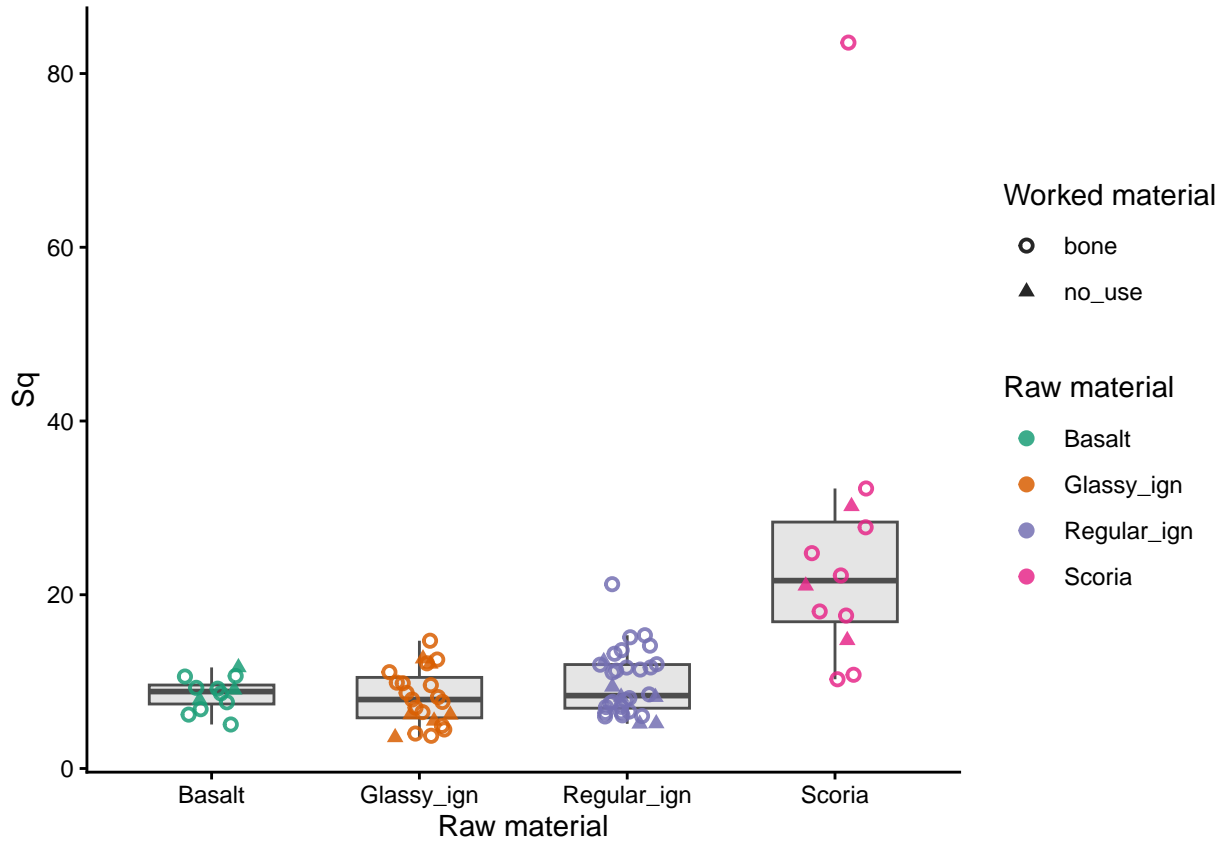
print(p)

ggsave(
  filename = file_out,
  plot     = p,
  path     = "analysis/plots",
  device   = "pdf",
  width    = 26,
  height   = 21,

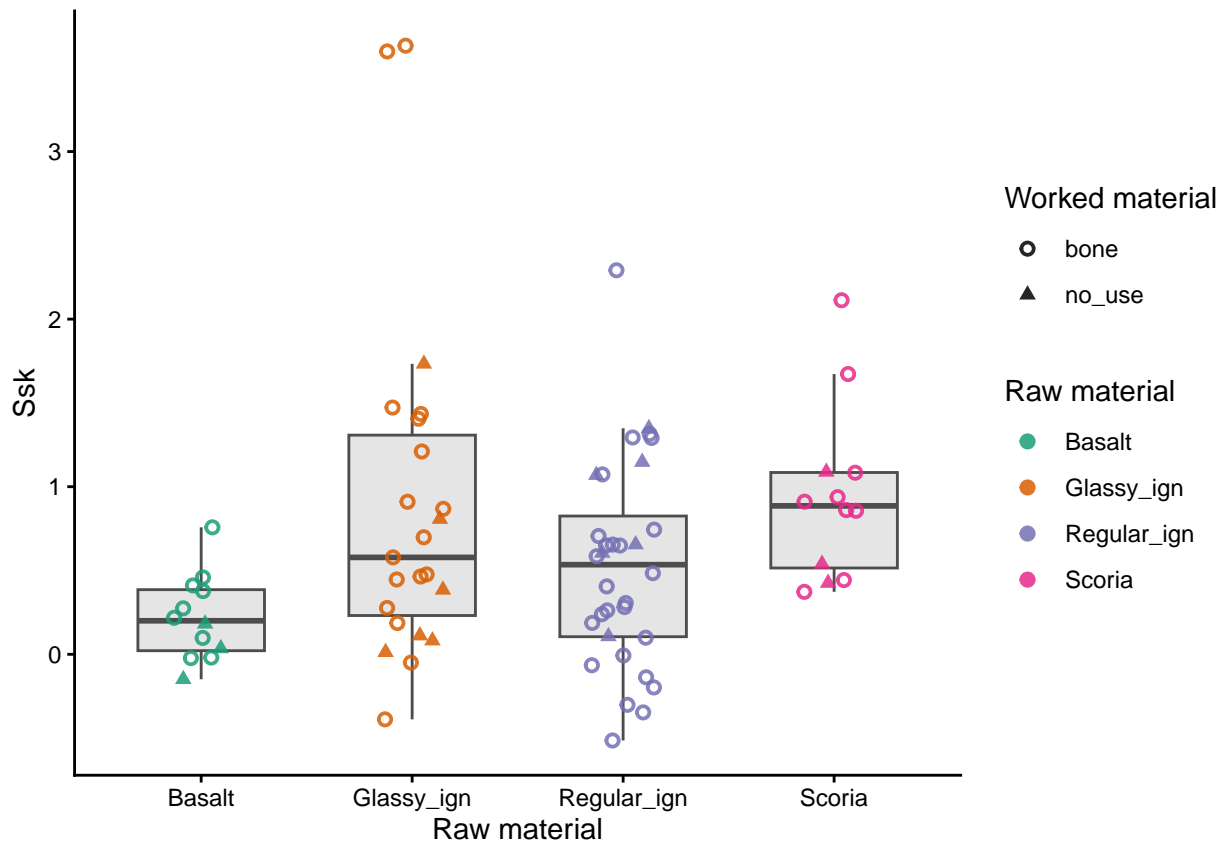
```

```
units = "cm"  
)  
}
```

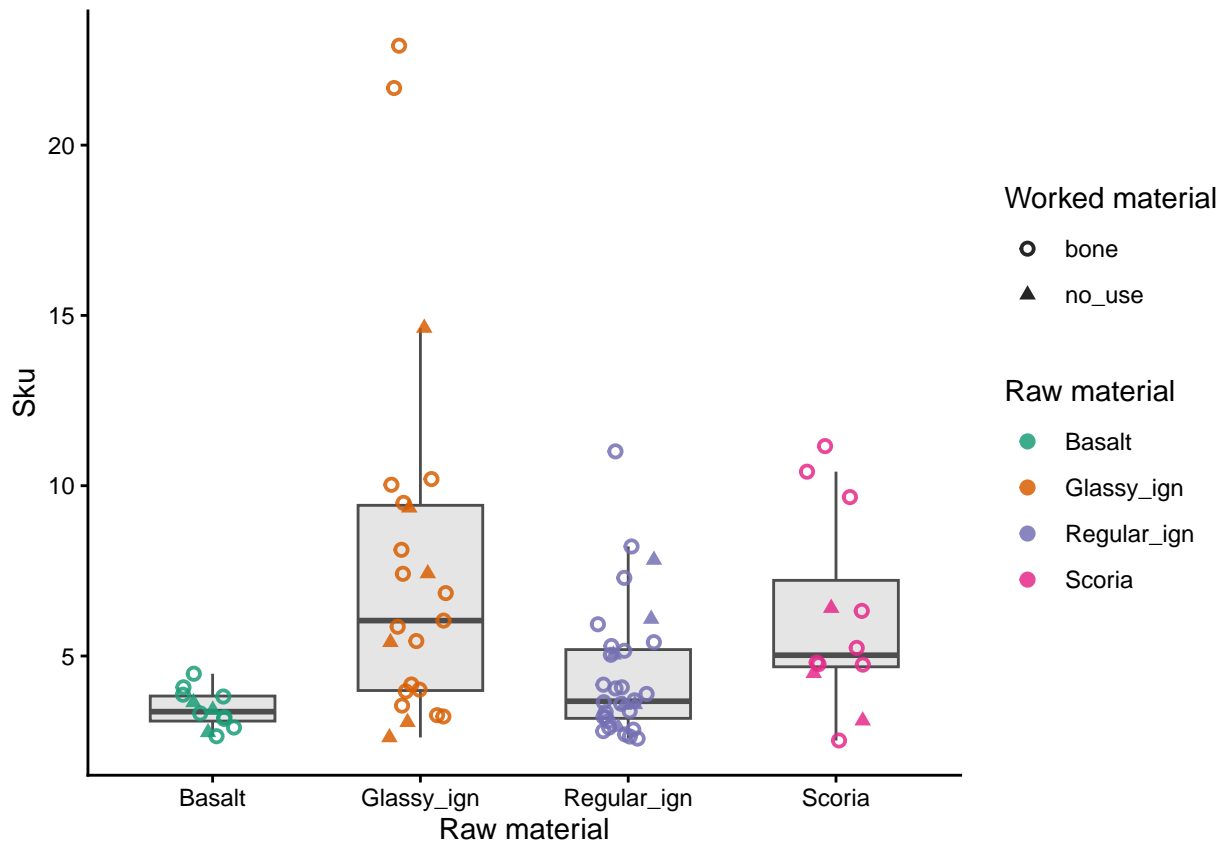
## [23] Sq



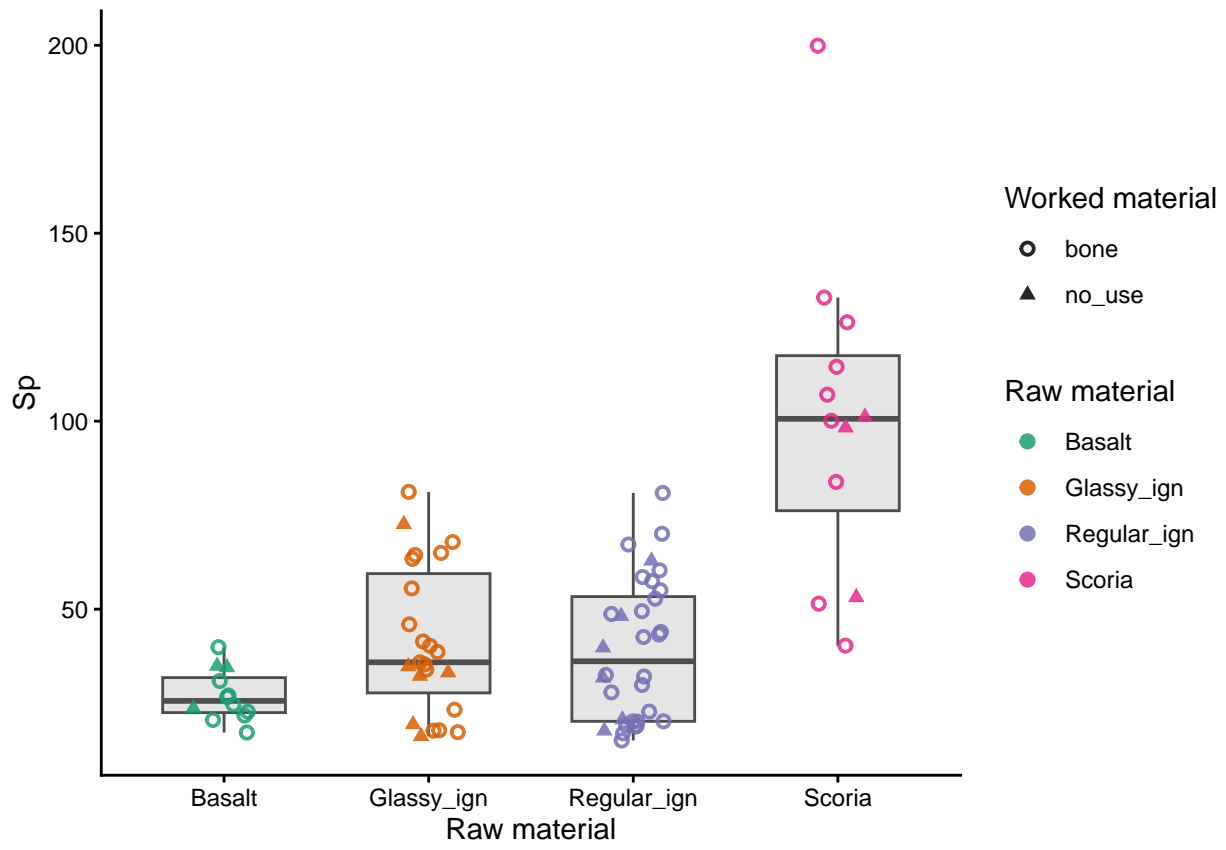
## [24] Ssk



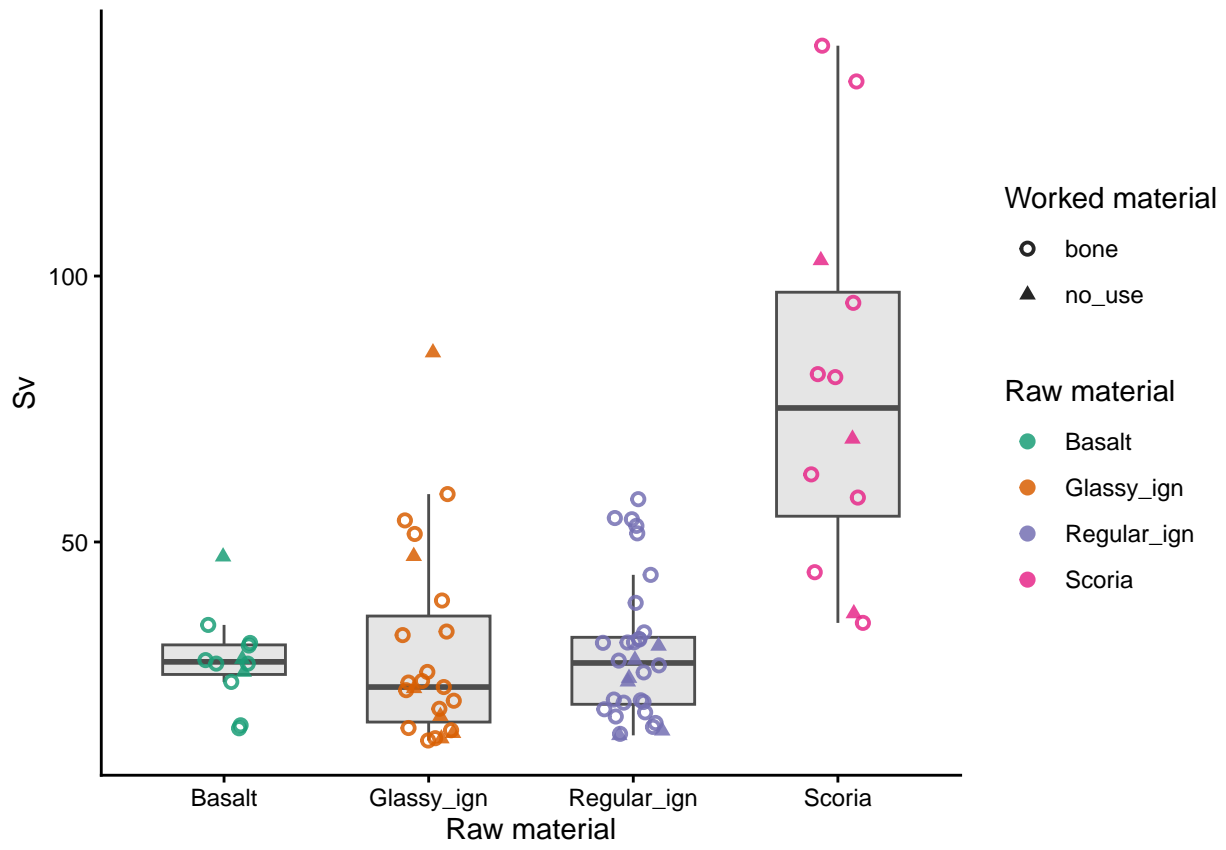
## [25] Sku



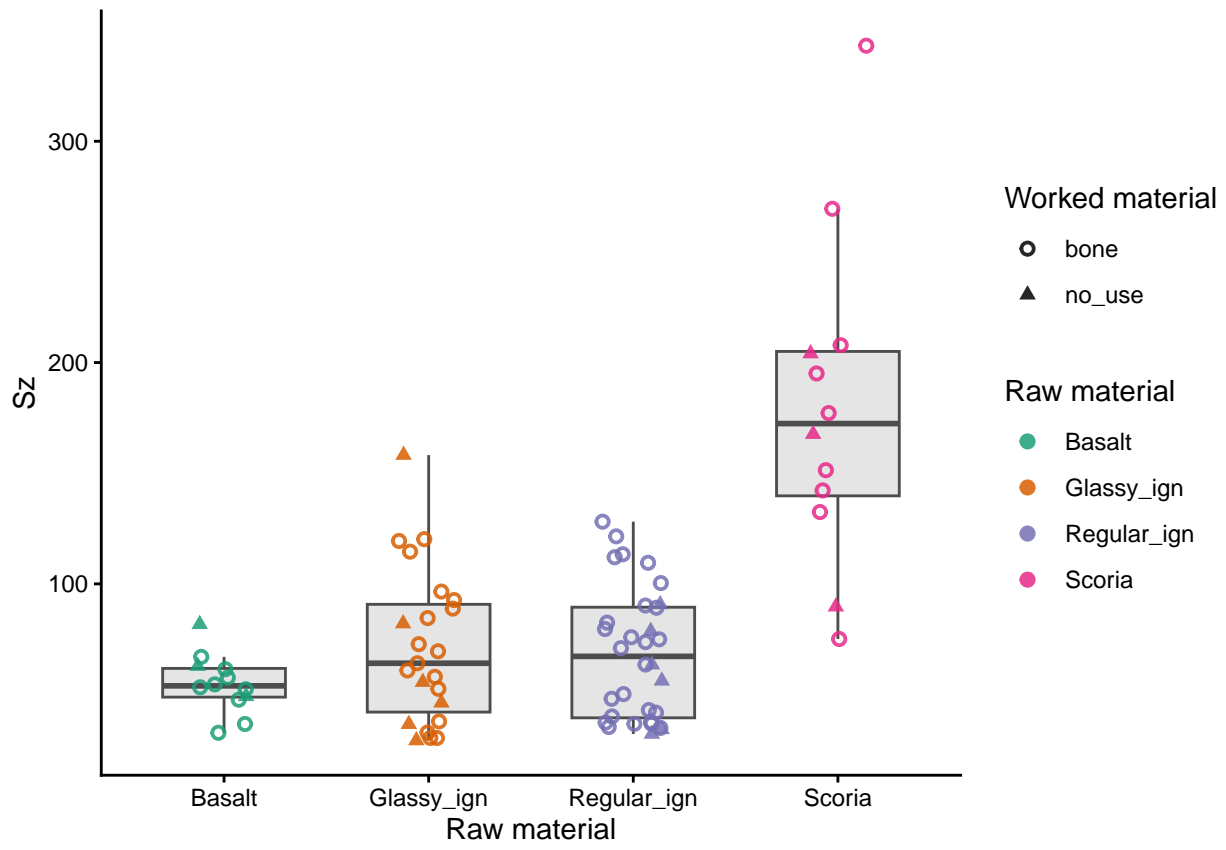
## [26] Sp



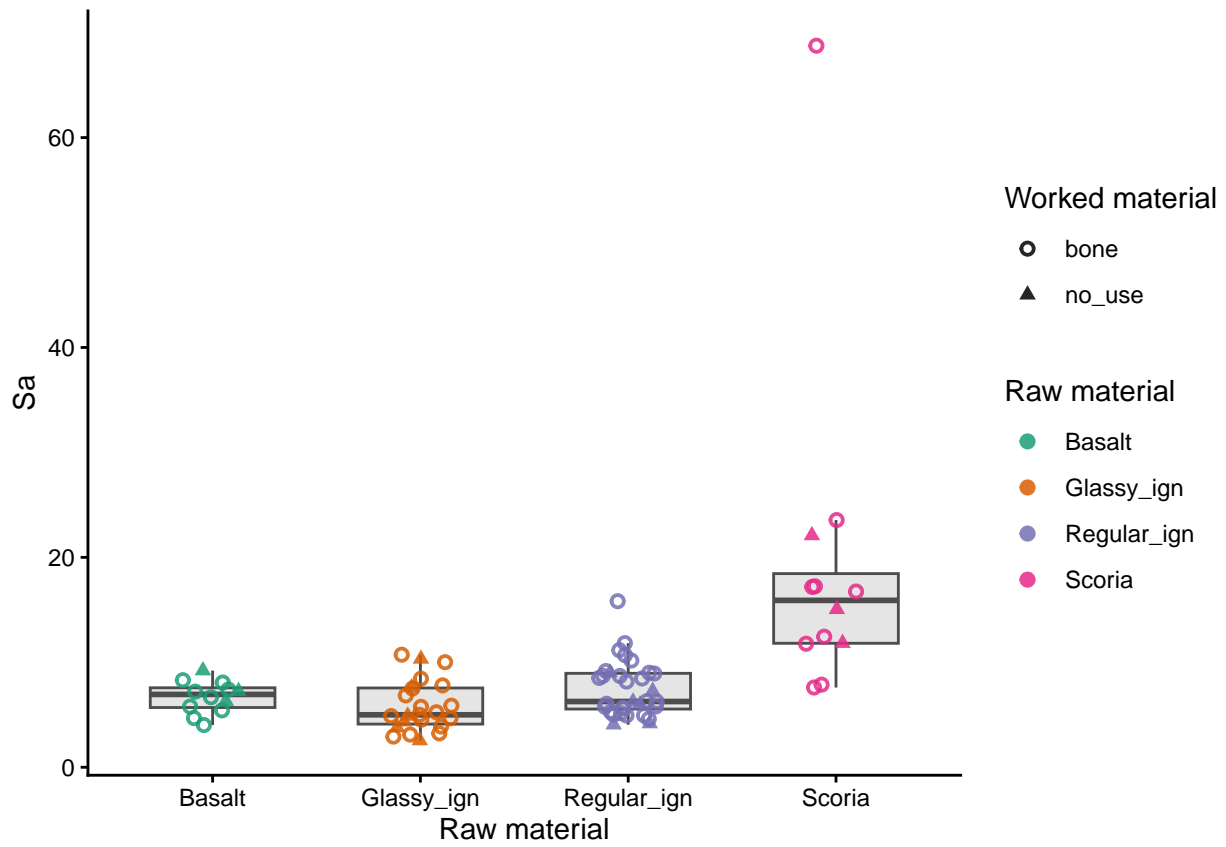
## [27] Sv



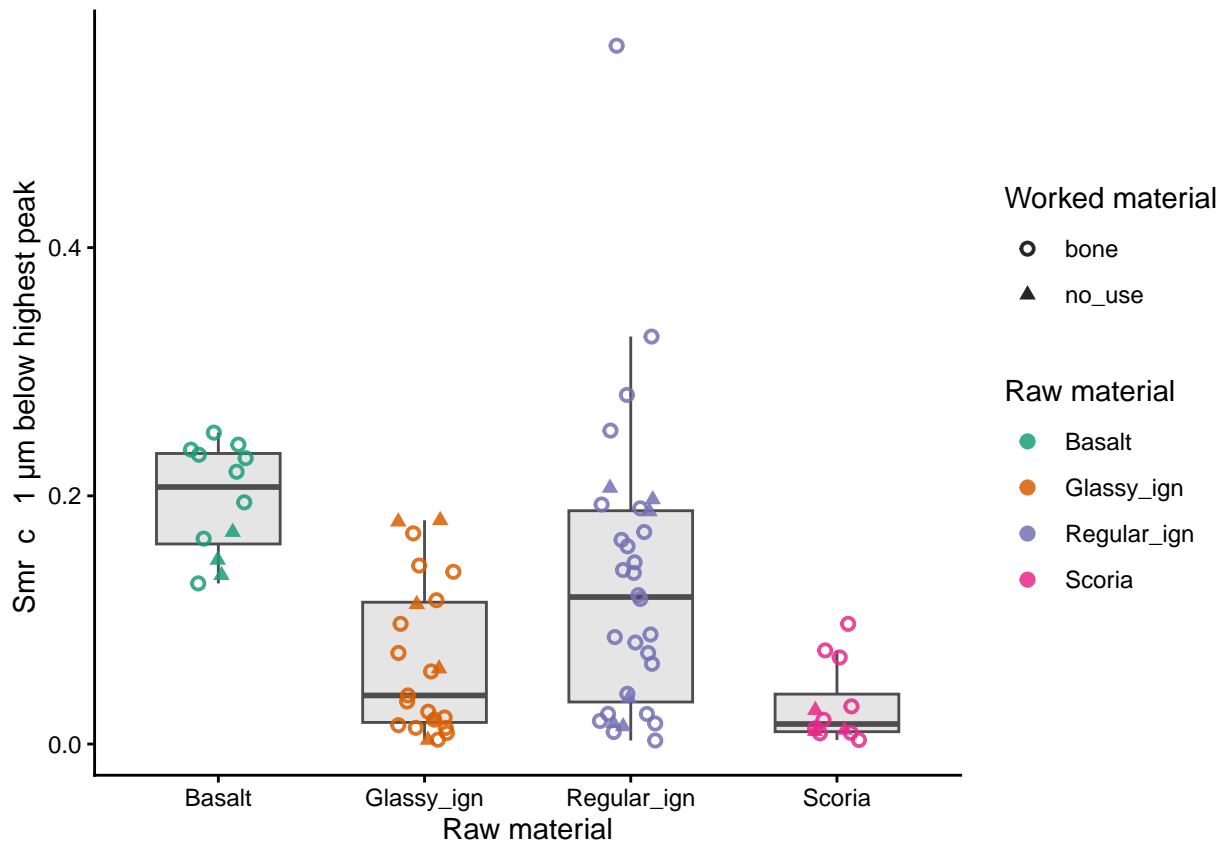
## [28] Sz



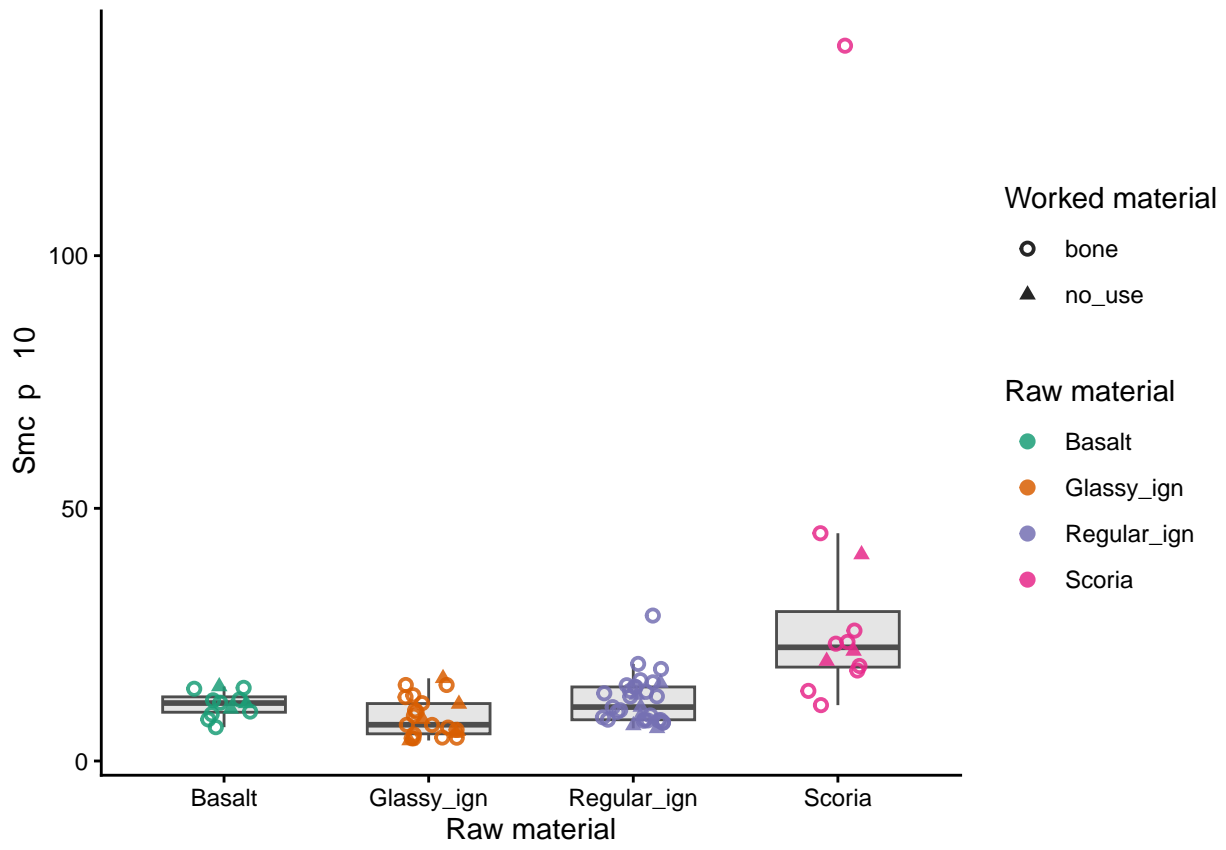
## [29] Sa



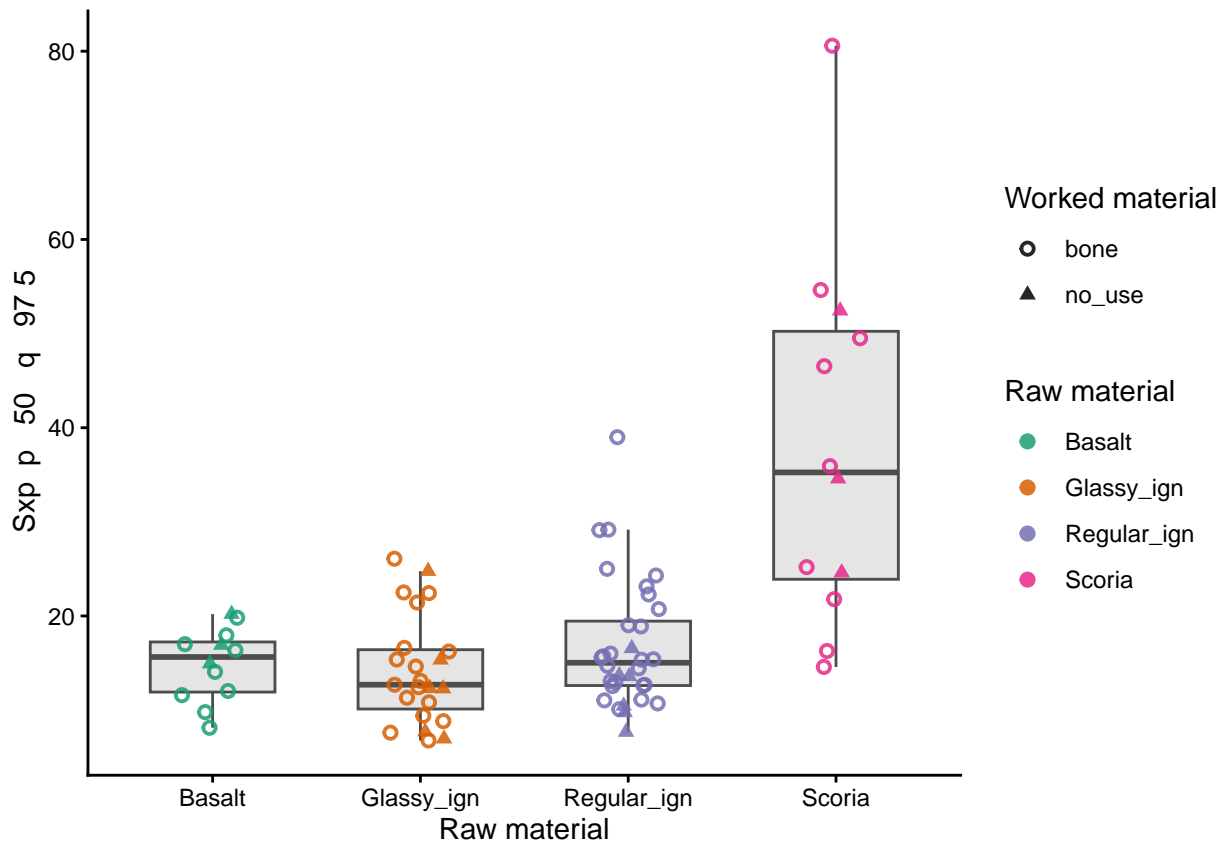
## [30] Smr...c...1.µm.below.highest.peak.



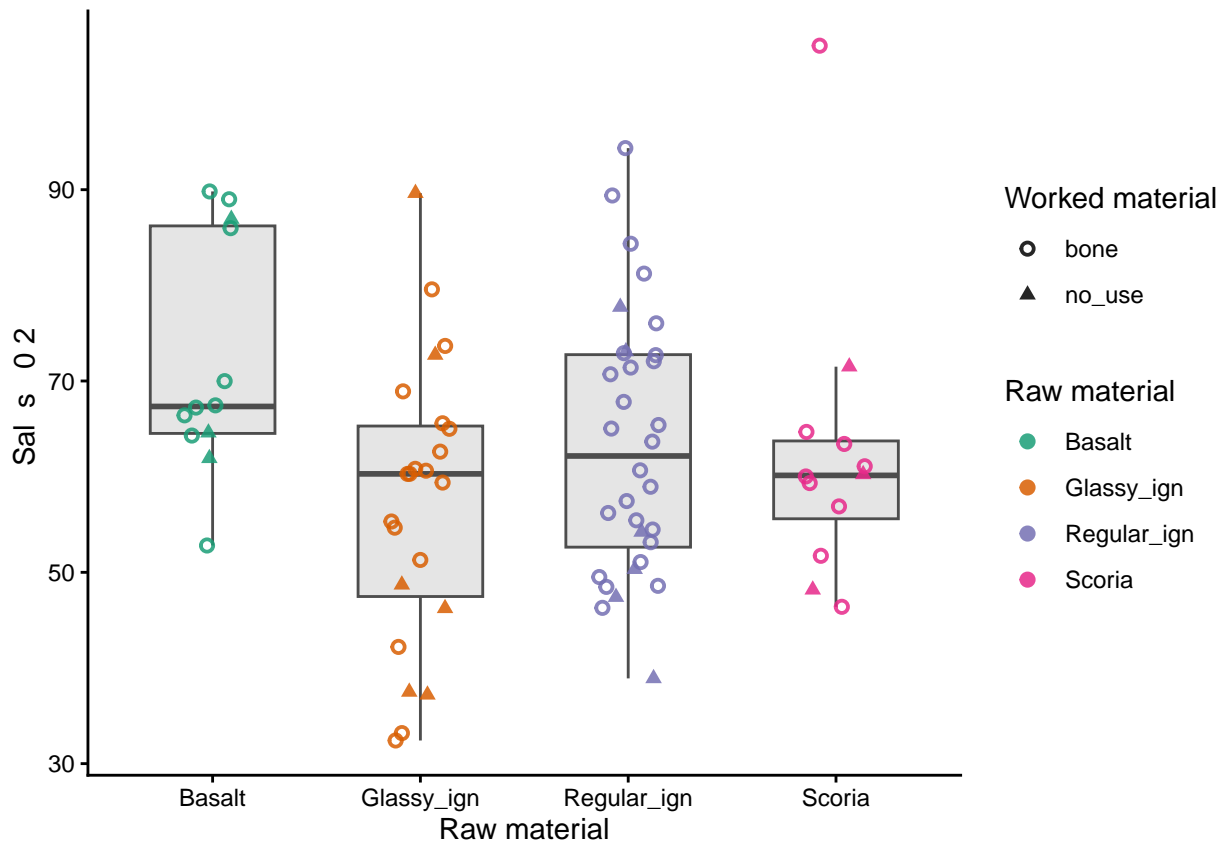
## [31] Smc..p...10..



## [32] Sxp...p...50...q...97.5..

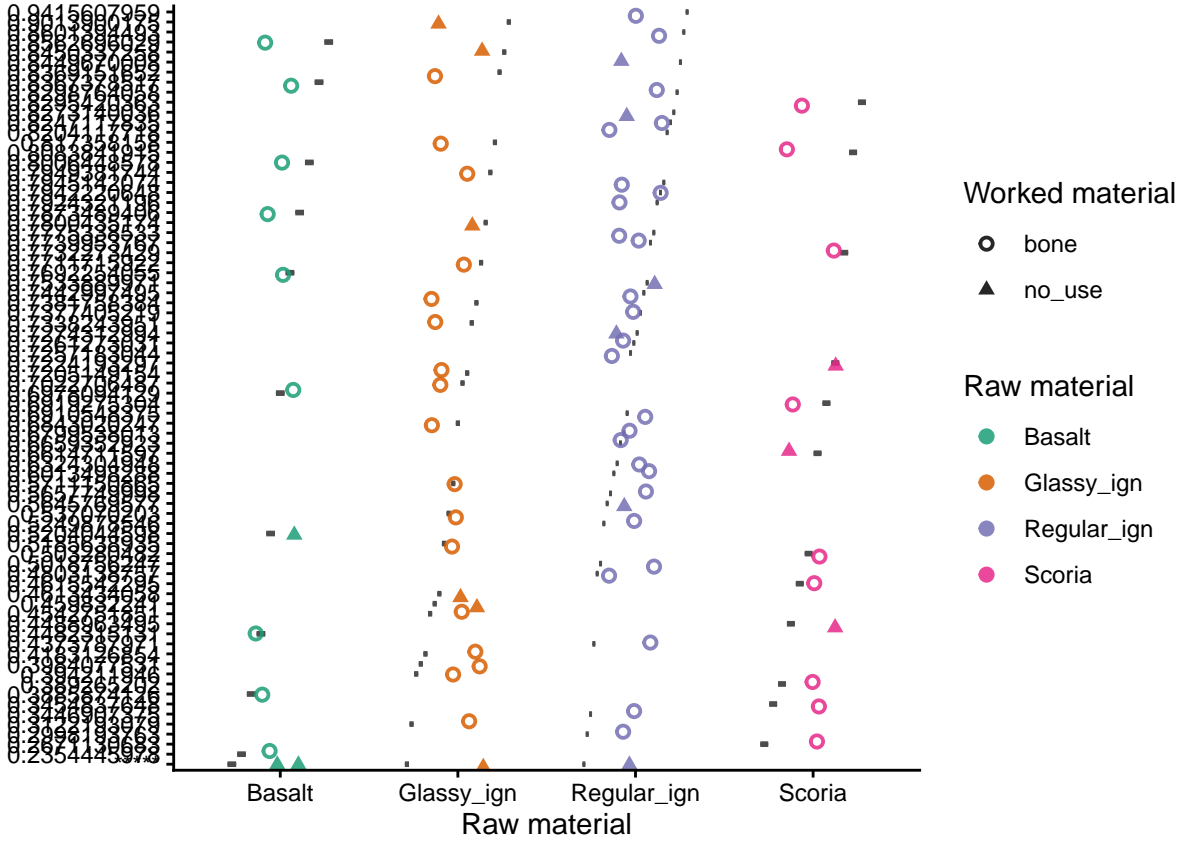


## [33] Sa1...s...0.2.

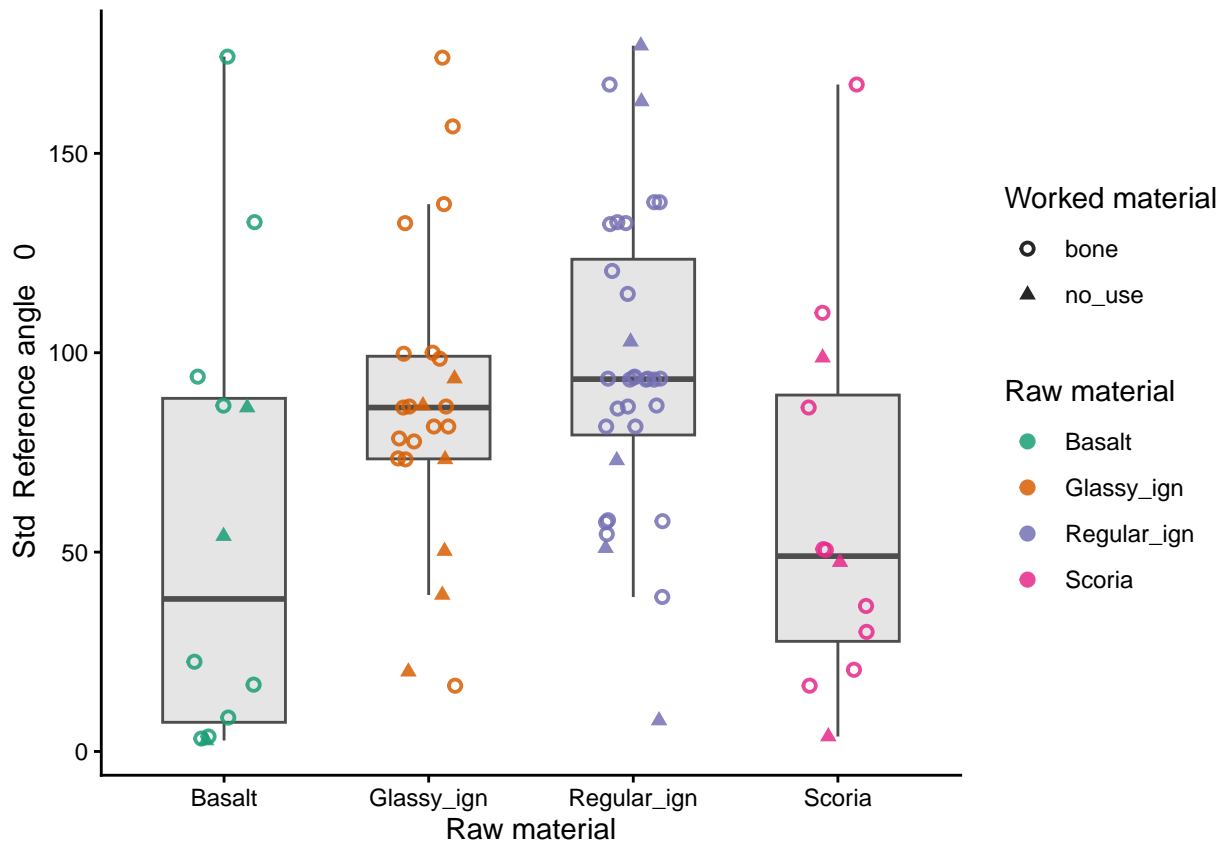


## [34] Str...s...0.2.

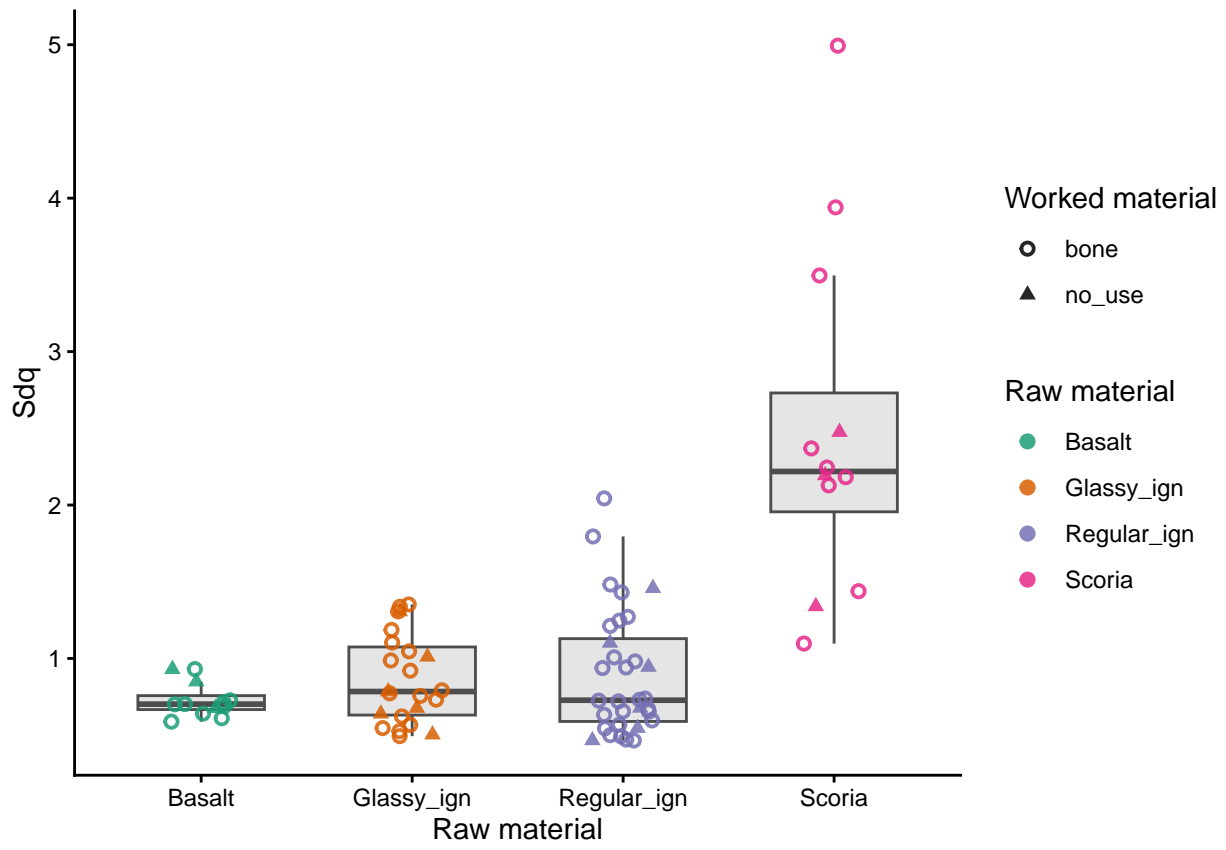
Str s 02



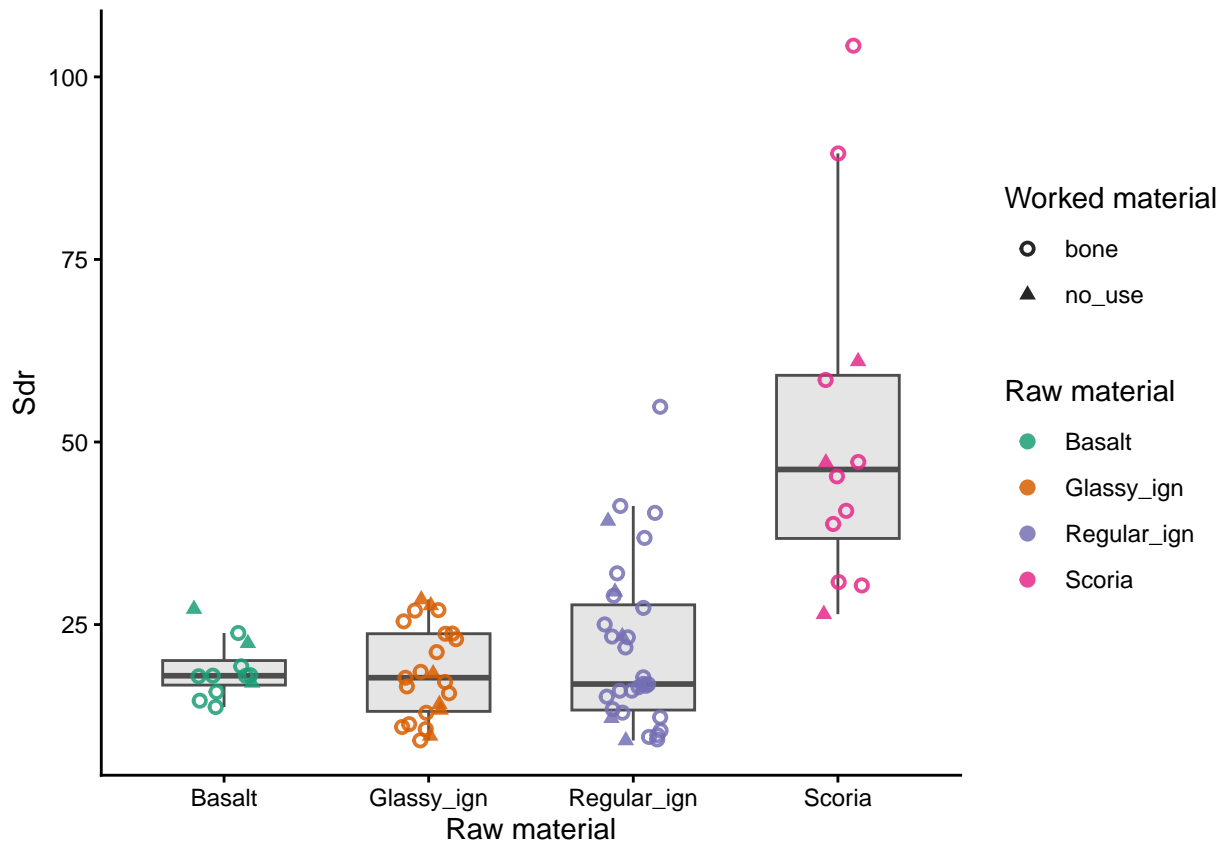
## [35] Std..Reference.angle...0..



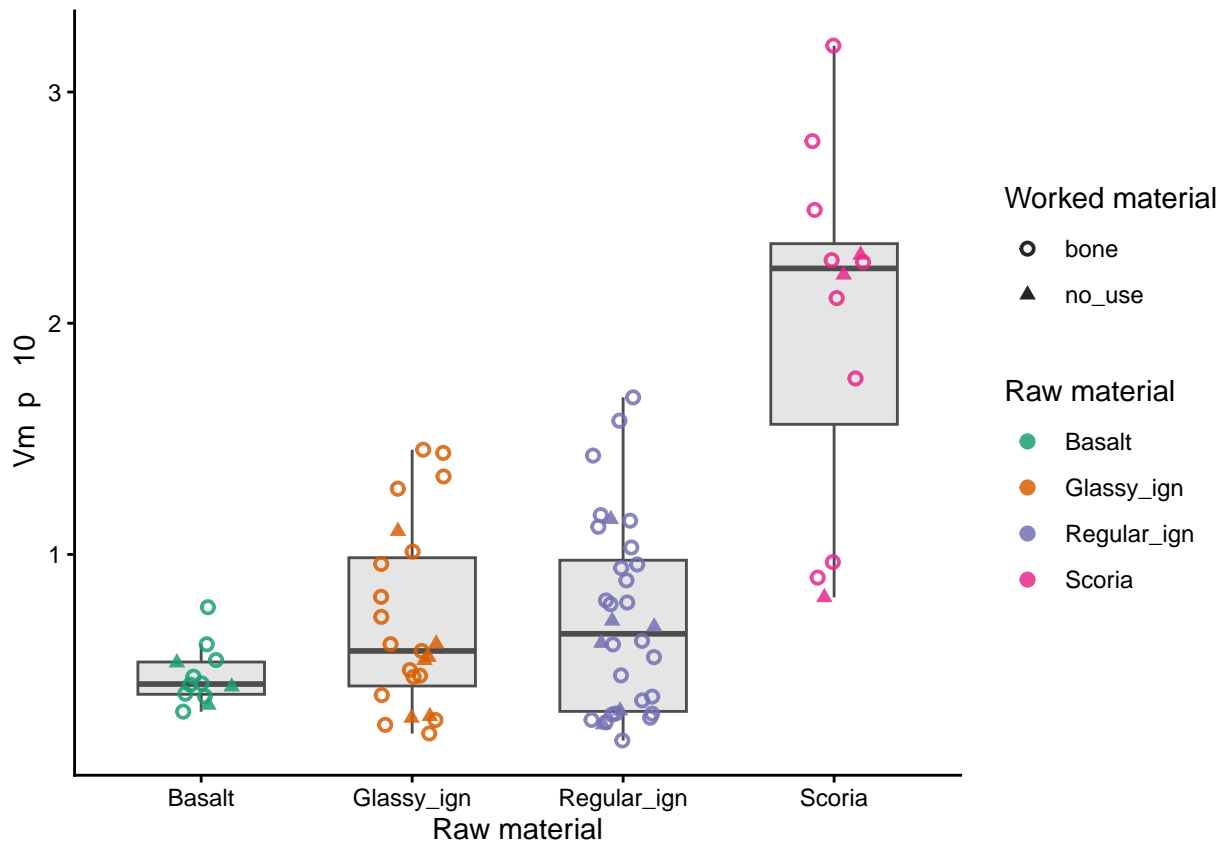
## [36] Sdq



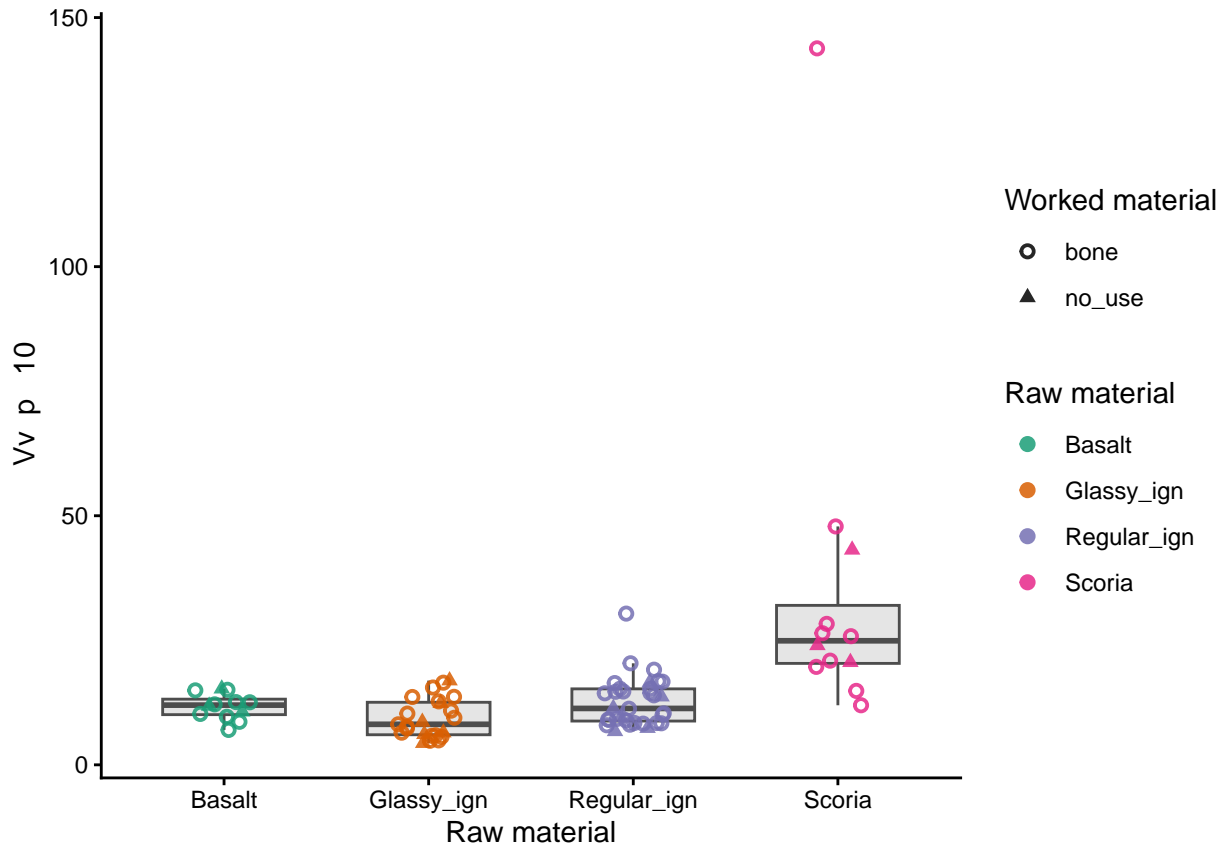
## [37] Sdr



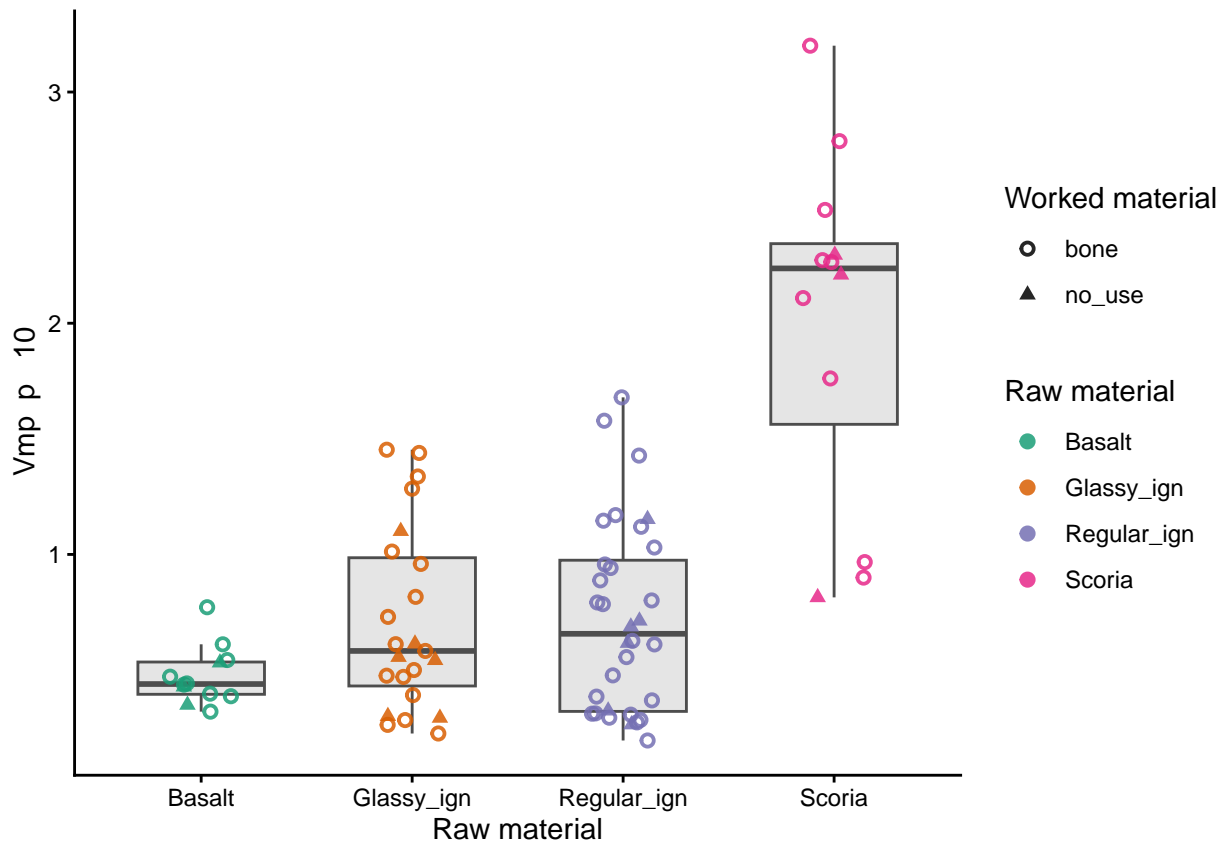
## [38] Vm..p...10..



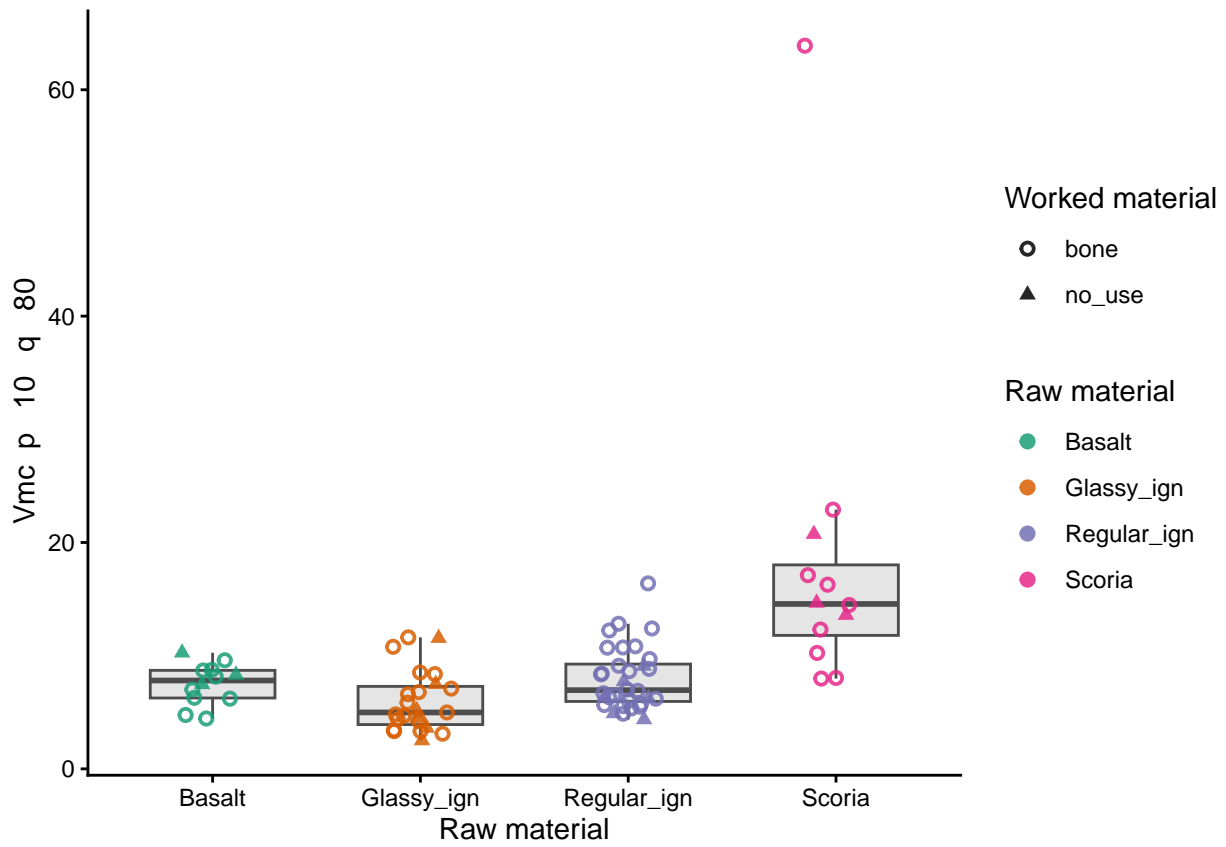
## [39] Vv..p...10..



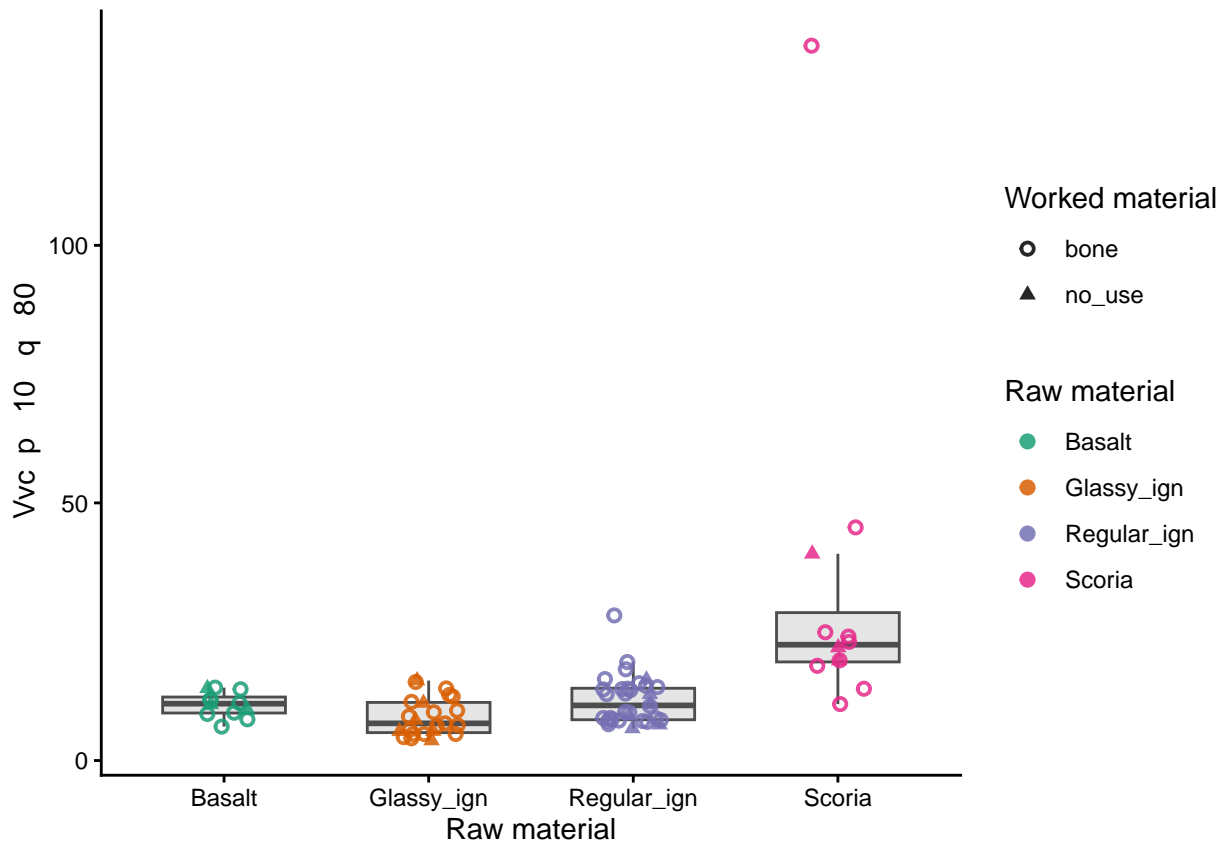
## [40] Vmp..p...10..



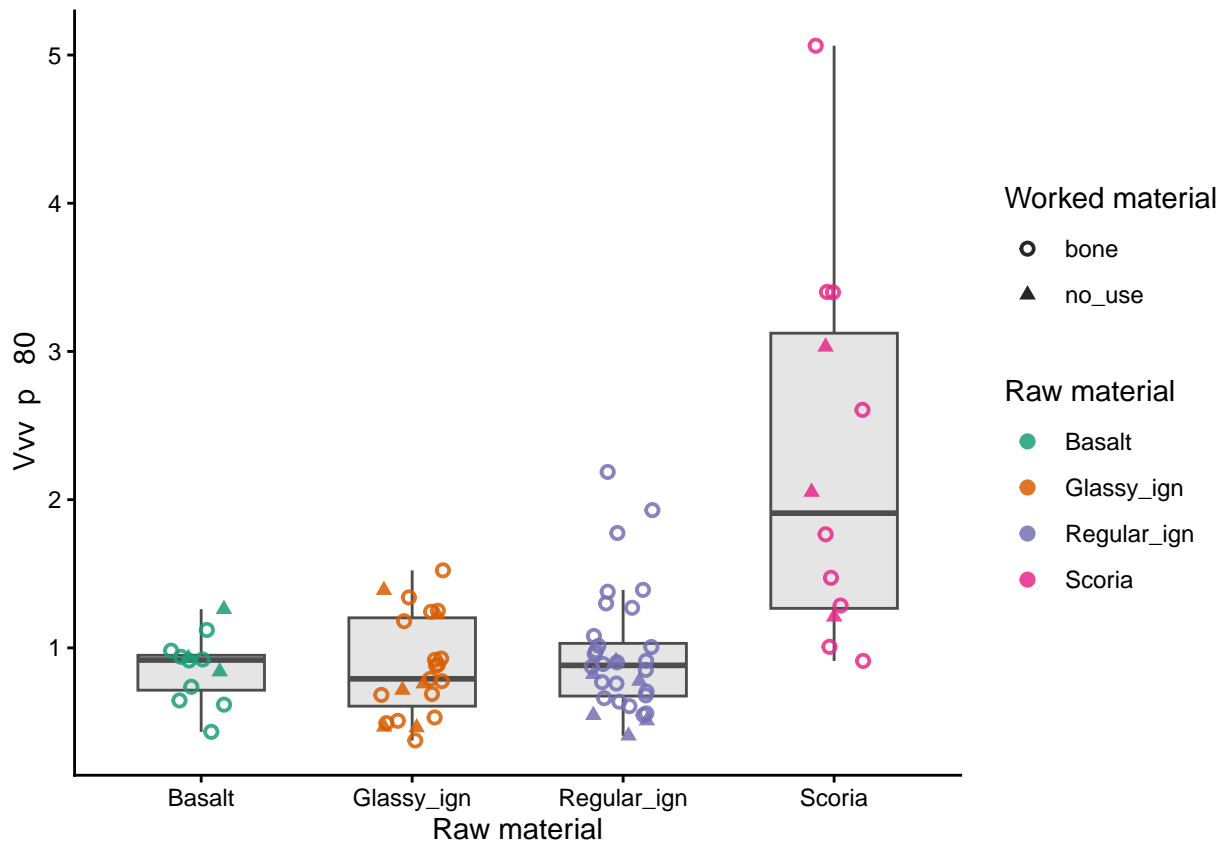
## [41] Vmc..p...10...q...80..



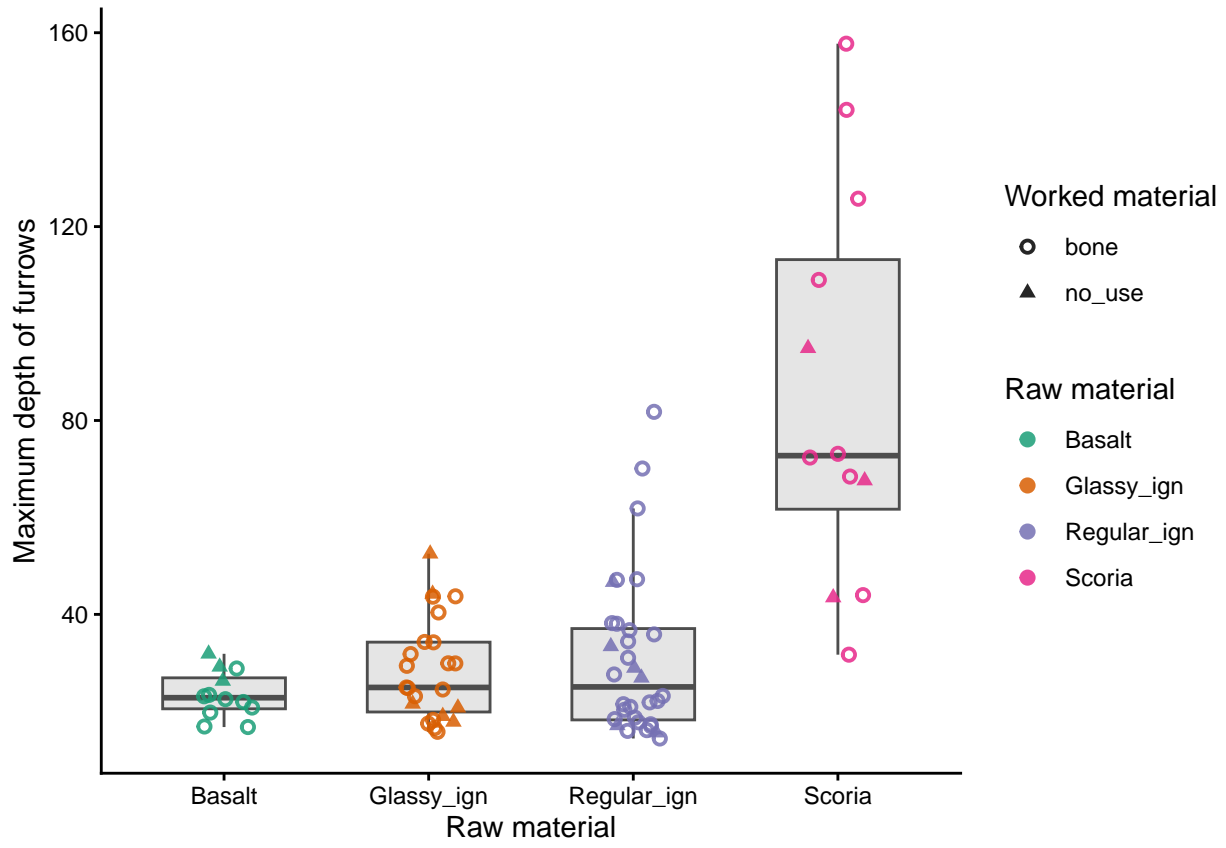
## [42] Vmc..p...10...q...80..



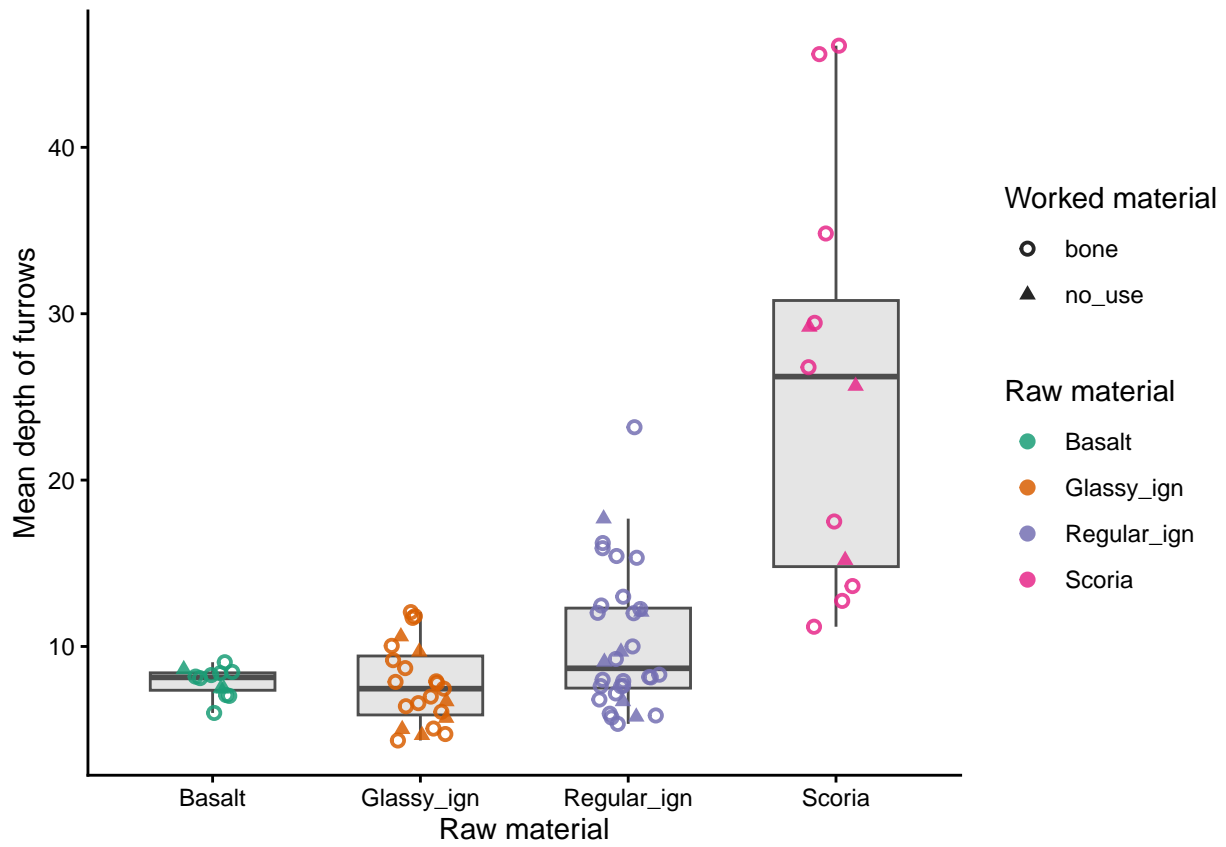
## [43] Vvv...p...80..



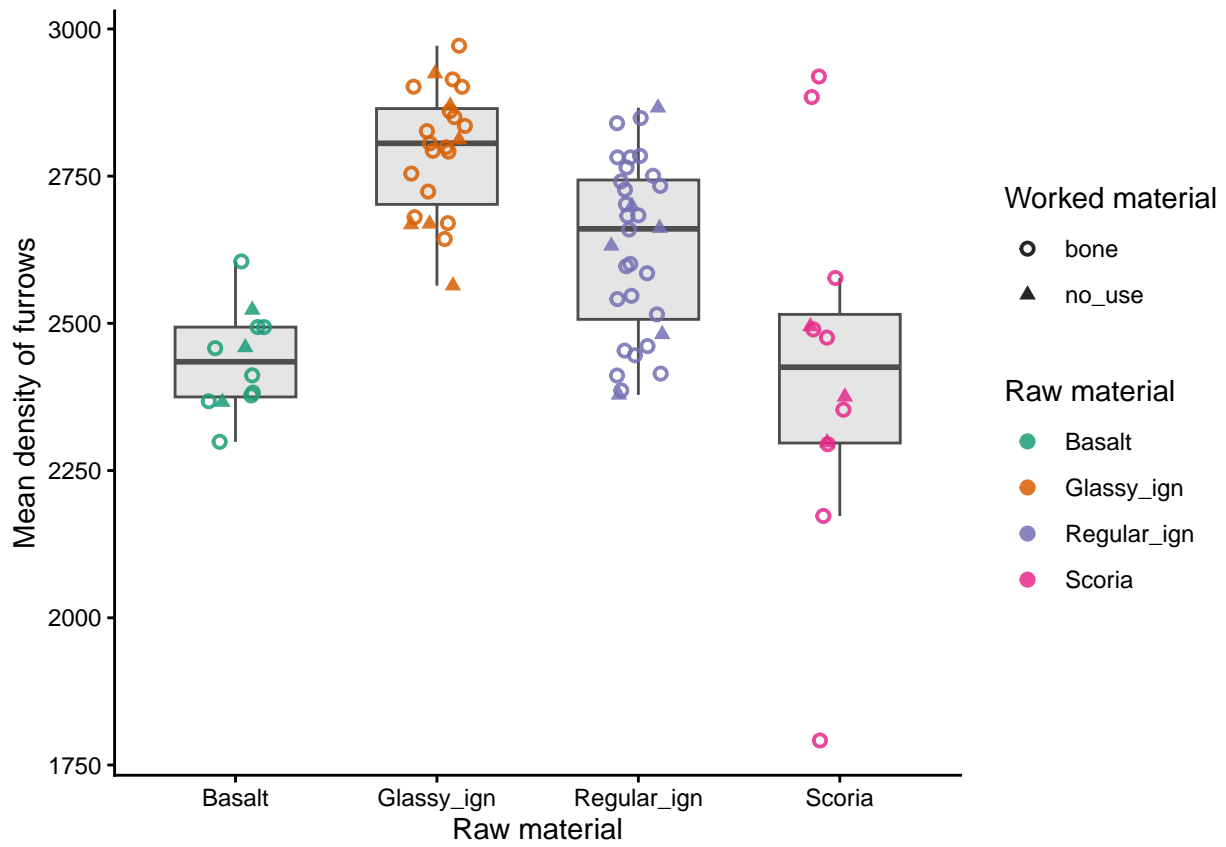
## [44] Maximum.depth.of.furrows



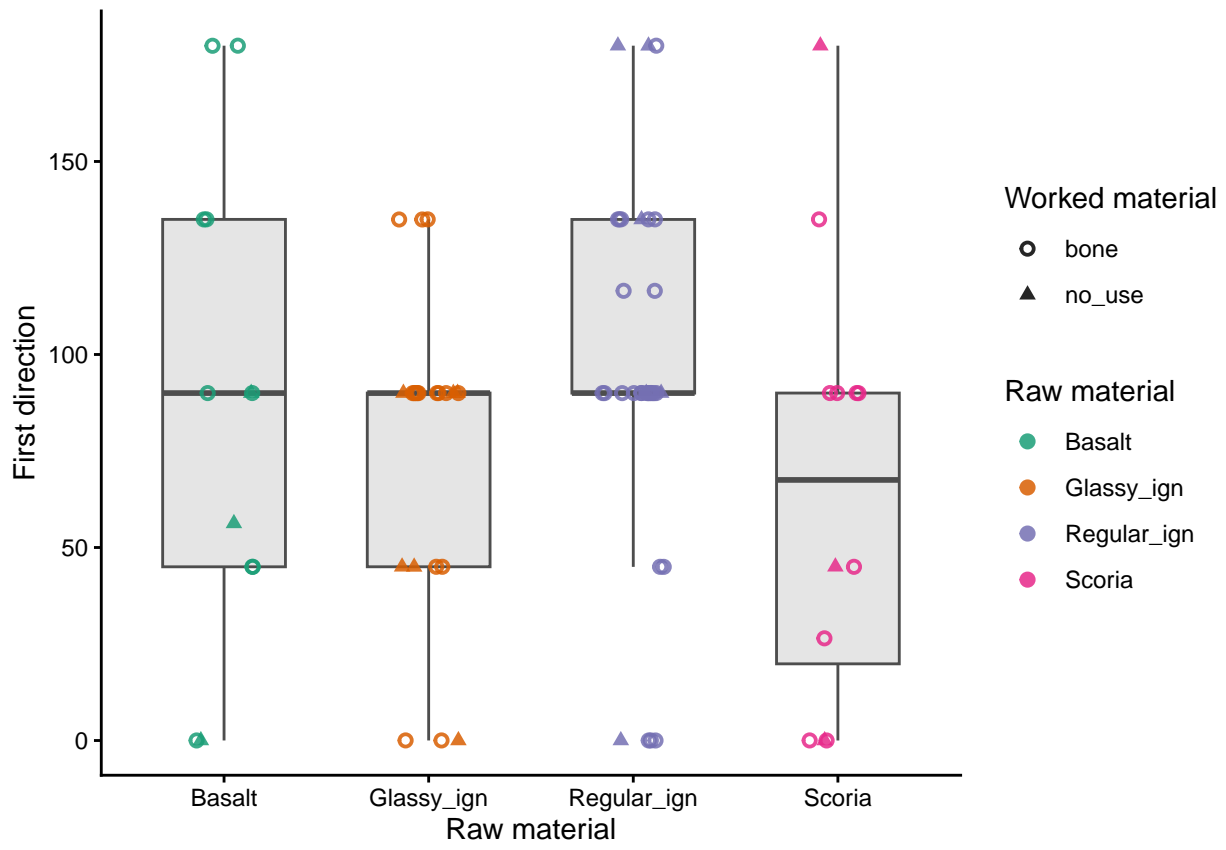
## [45] Mean.depth.of.furrows



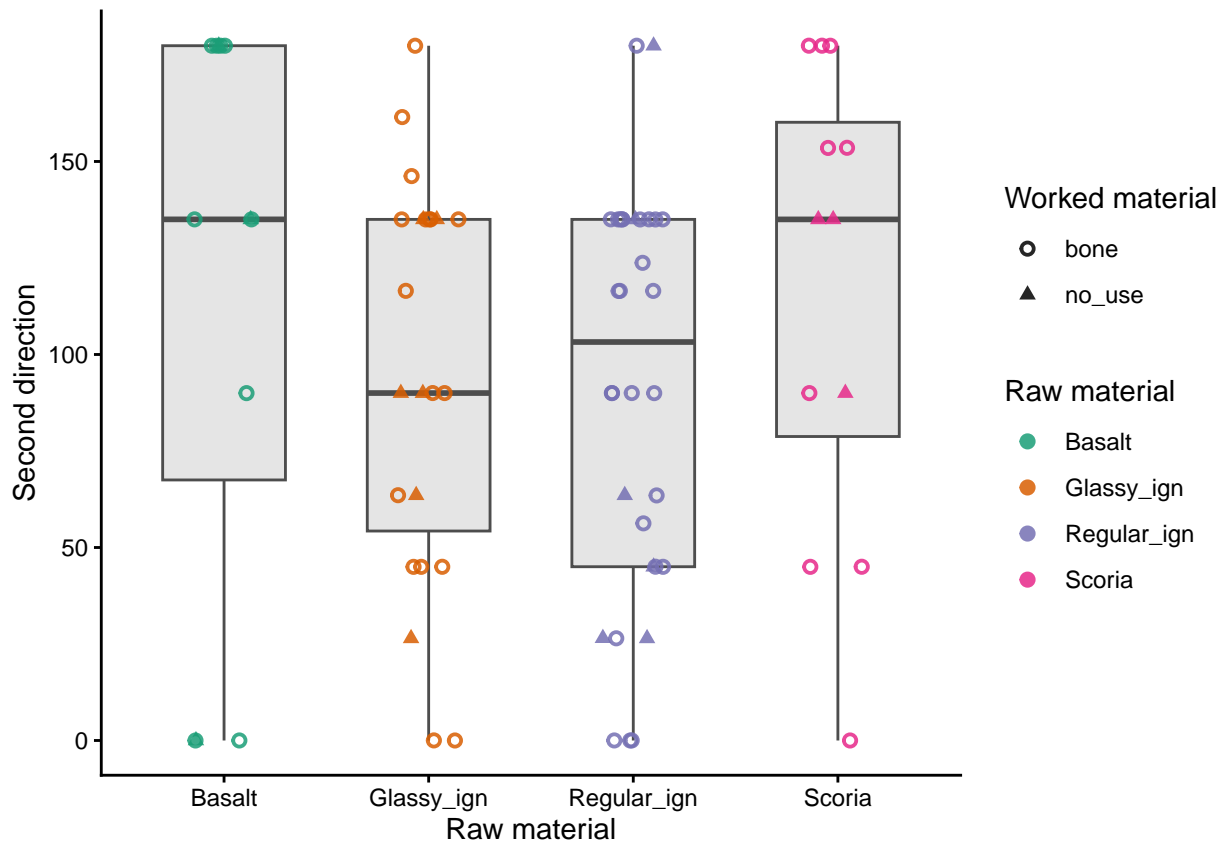
## [46] Mean.density.of.furrows



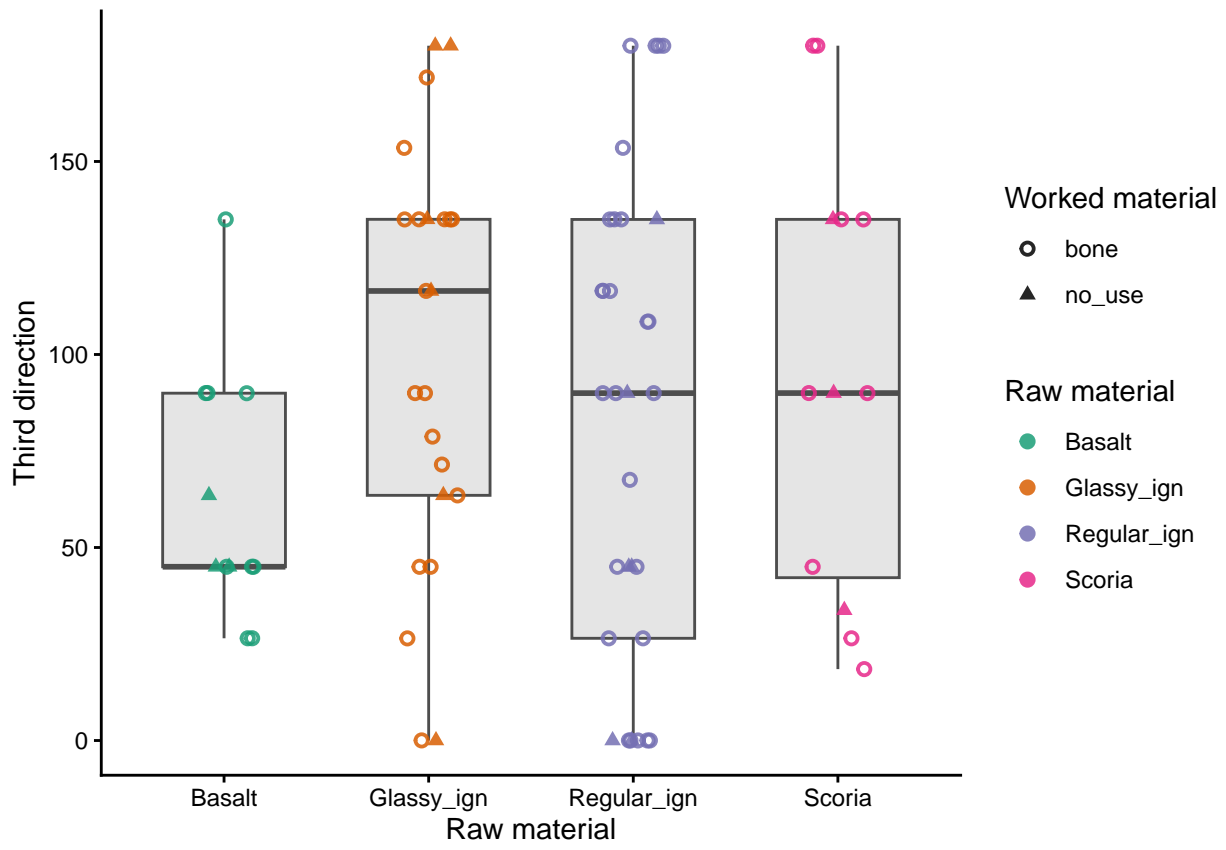
## [47] First.direction



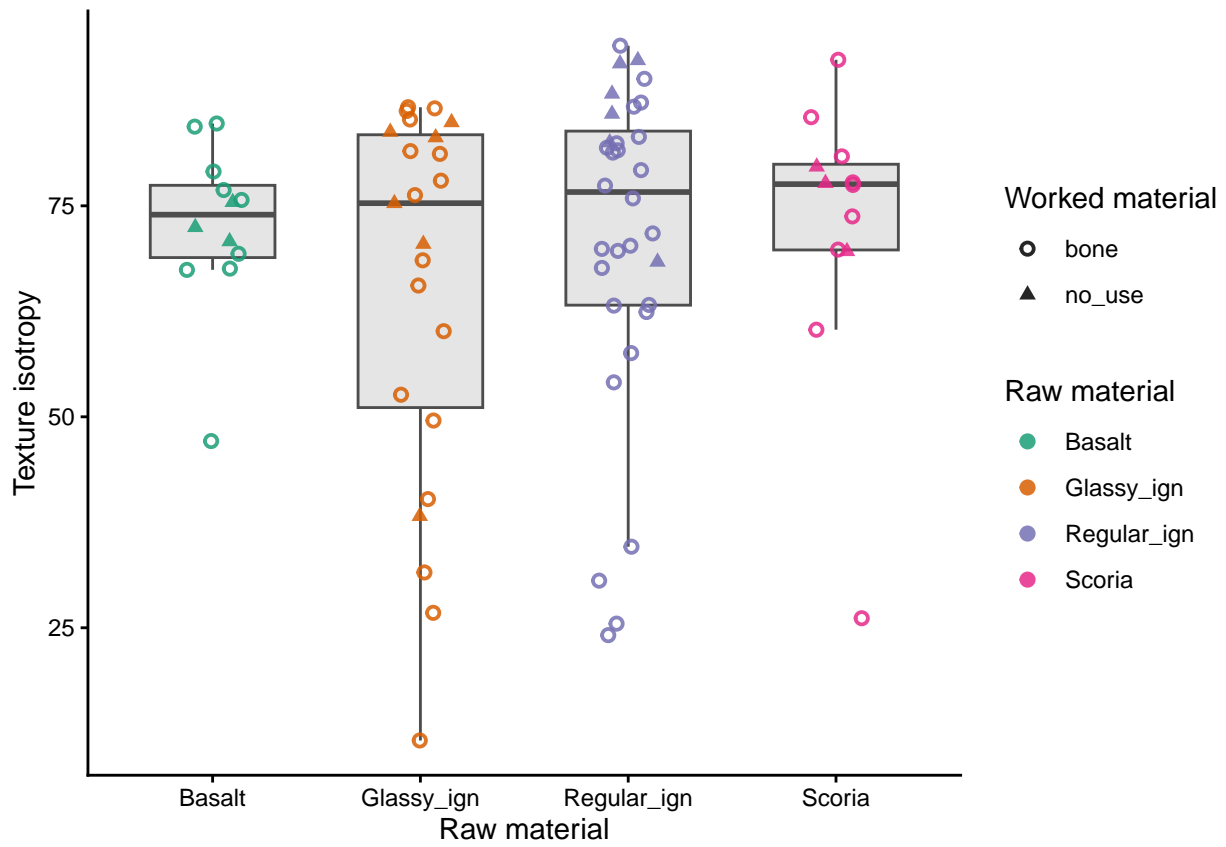
## [48] Second.direction



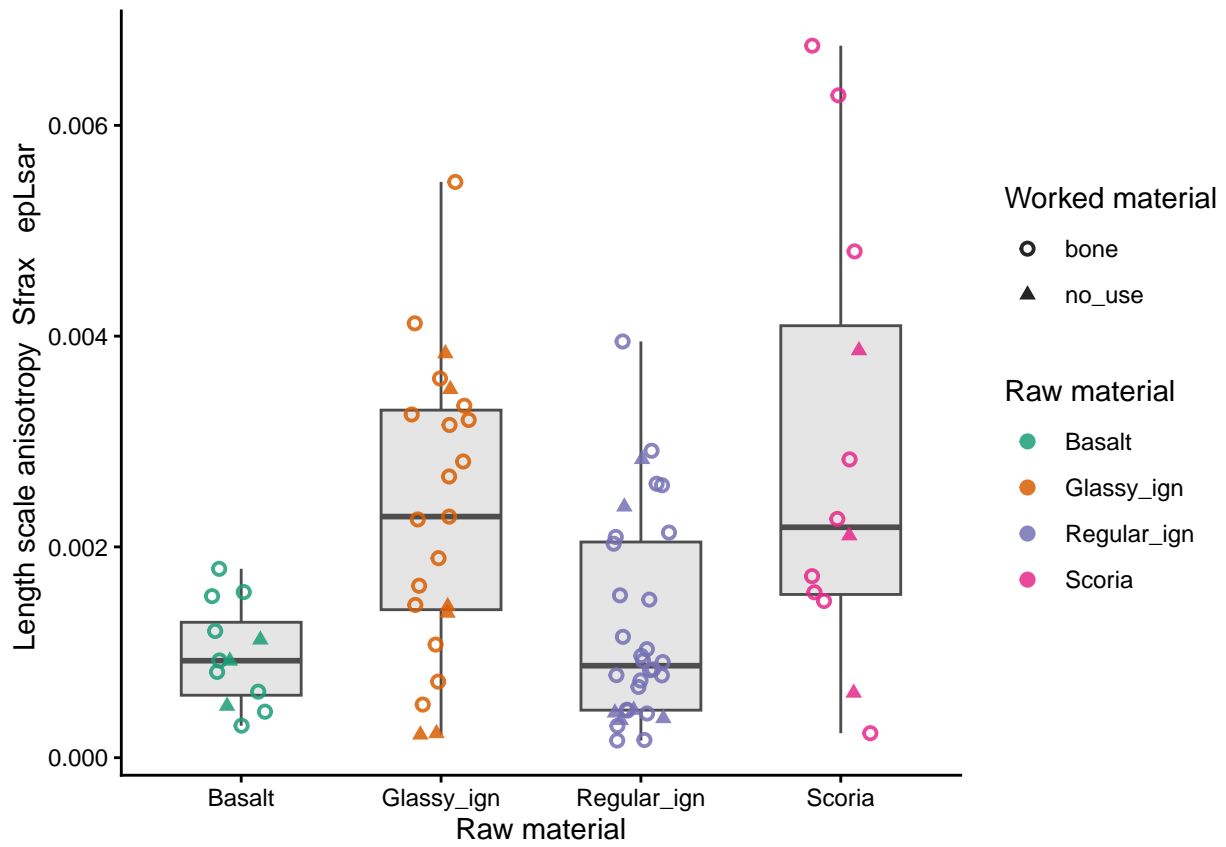
## [49] Third.direction



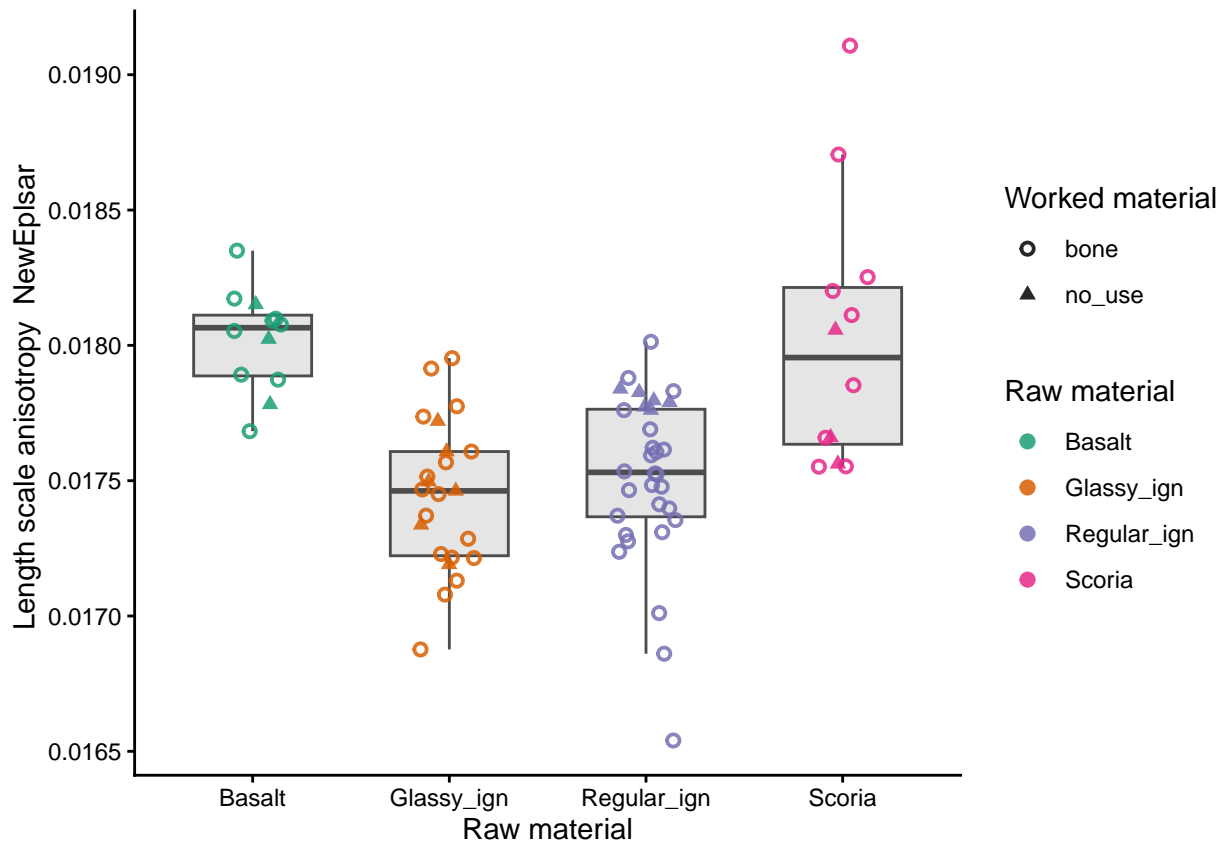
## [50] Texture.isotropy



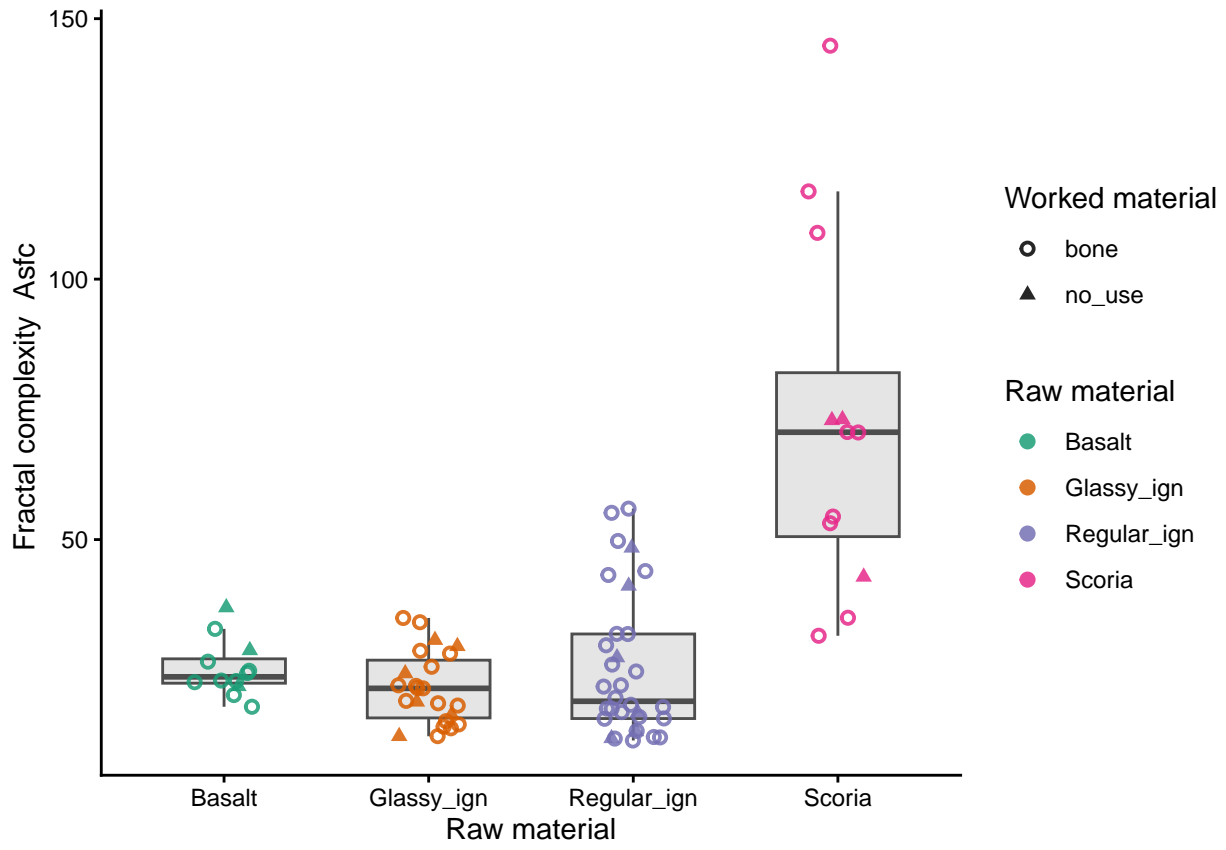
## [51] Length.scale.anisotropy..Sfrax...epLsar.



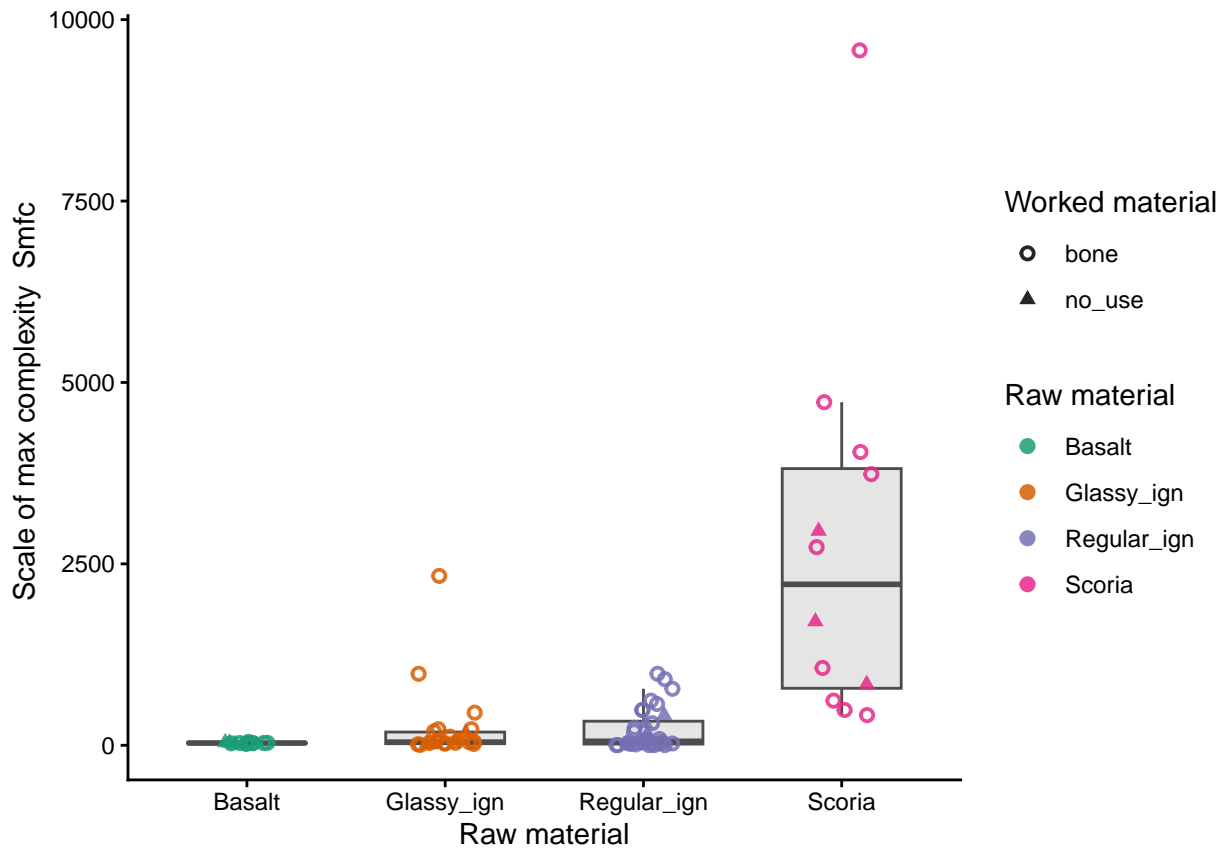
## [52] Length.scale.anisotropy..NewEplsar.



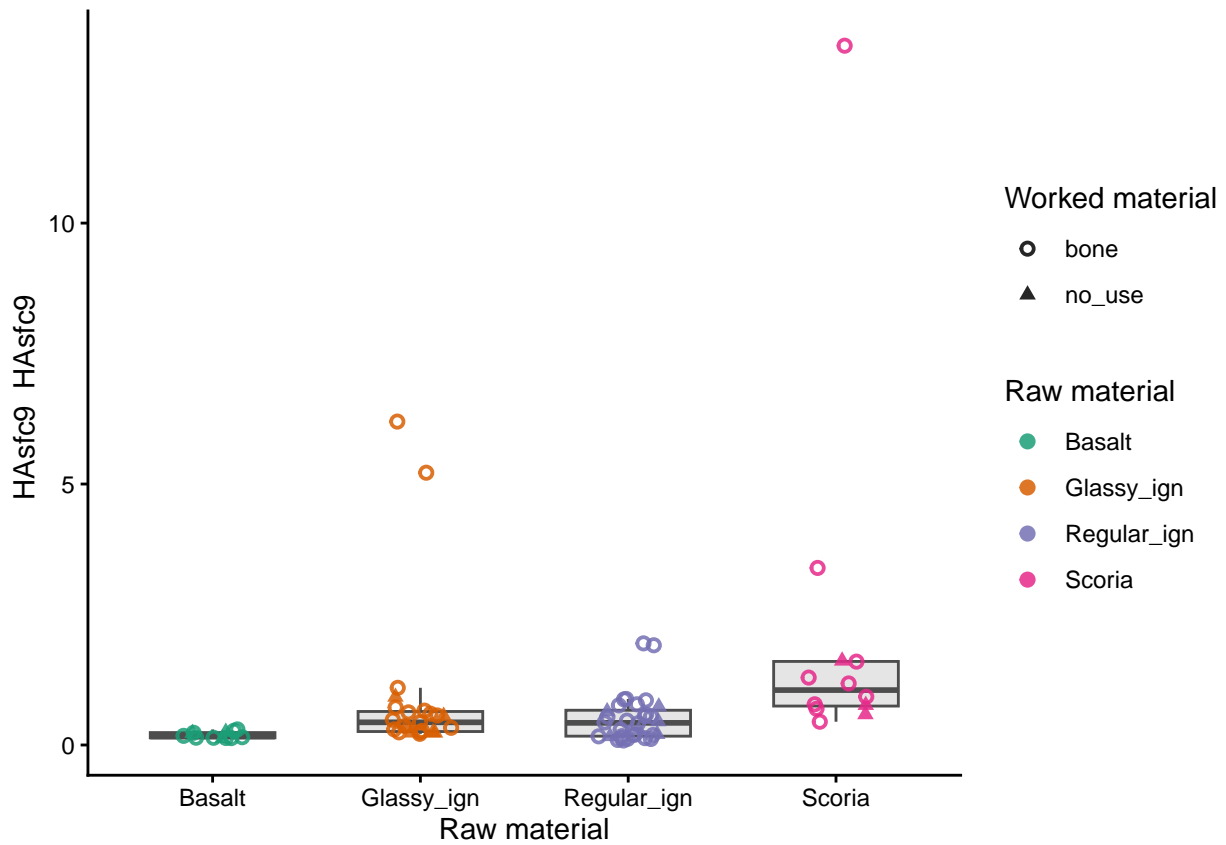
## [53] Fractal.complexity..Asfc.



## [54] Scale.of.max.complexity..Smfc.

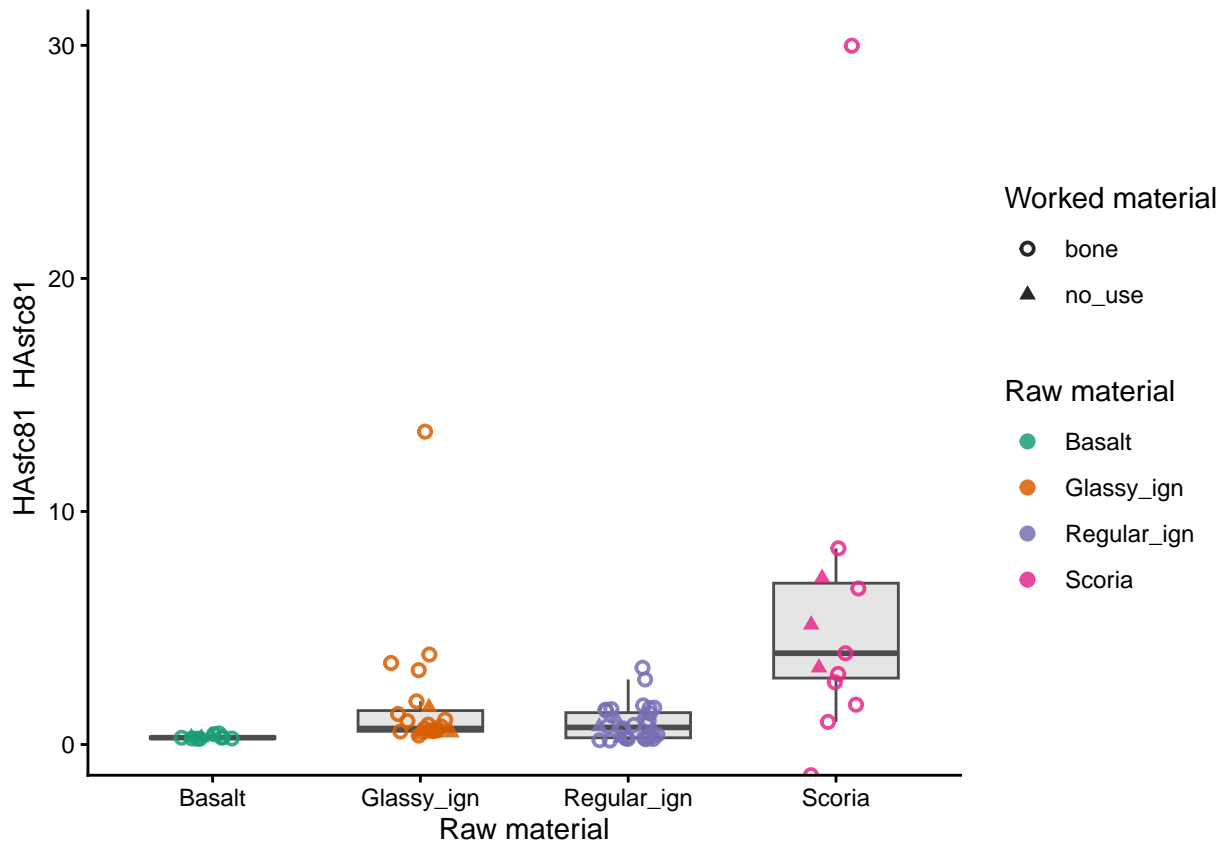


## [55] HAsfc9..HAsfc9.



```
## [56] HAsfc81..HAsfc81.
```

```
## Warning: Removed 1 row containing non-finite outside the scale range ('stat_boxplot()').
## Removed 1 row containing non-finite outside the scale range ('stat_boxplot()').
```



Scatterplots of selected variables combined by “Worked material” and “Motion”

```
# Main parameters
# Sq, Sa → overall roughness
# Vmc → load-bearing core volume
# Asfc / Smfc → complexity (use-wear intensity)
# Furrow depth & density → contact mechanics

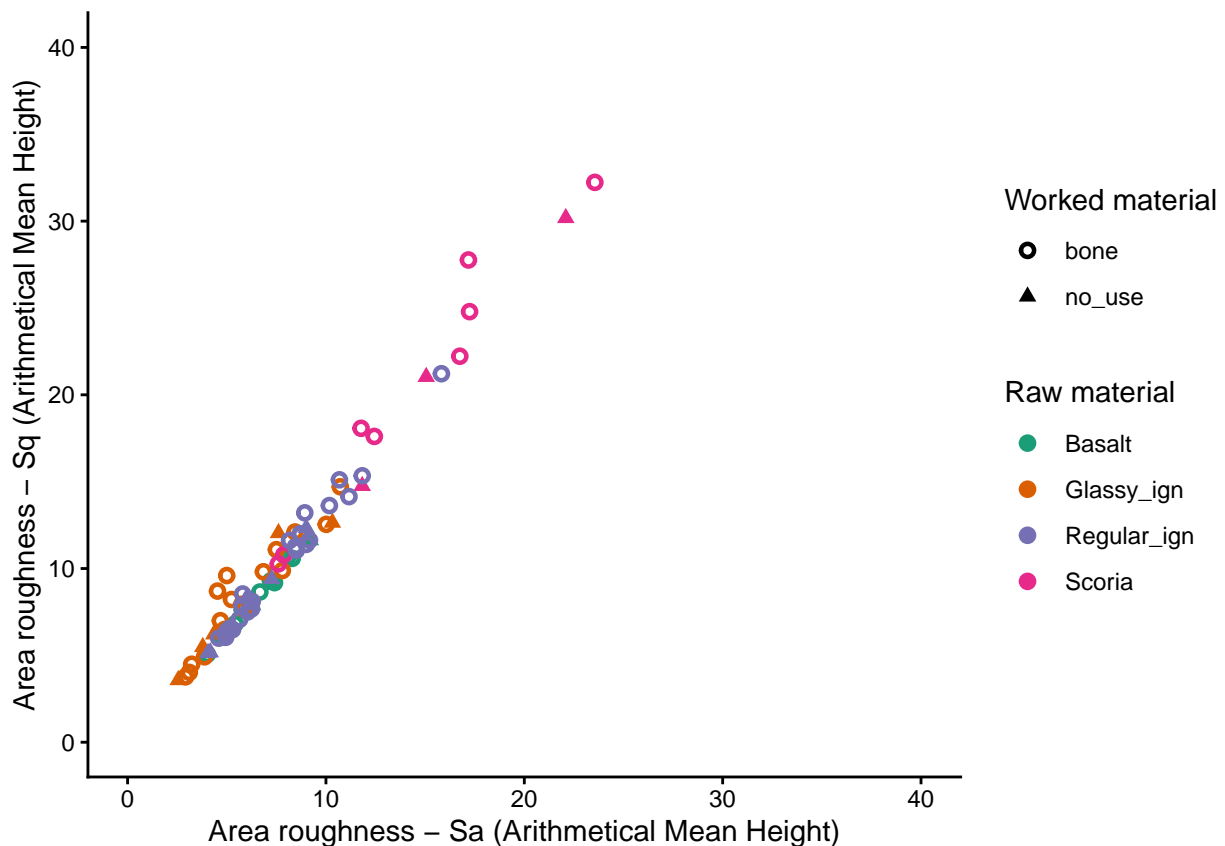
# Sa vs. Sq, overall roughness

Sa_Sq <- ggplot(confocaldata) +
  geom_point(
    aes(
      x = Sa,
      y = Sq,
      colour = Raw_material,
      shape = Worked_material
    ),
    size = 1.5,
    stroke = 1.2
  ) +
  coord_cartesian(
    xlim = c(0, 40),
```

```

ylim = c(0, 40)
) +
theme_classic() +
labs(
  x = "Area roughness - Sa (Arithmetical Mean Height)",
  y = "Area roughness - Sq (Arithmetical Mean Height)",
  colour = "Raw material",
  shape = "Worked material"
) +
scale_colour_brewer(palette = "Dark2") +
scale_shape_manual(
  values = c(
    bone = 1, # open circle
    no_use = 17 # filled triangle
  )
)
)
print(Sa_Sq)

```



```

file_out <- paste0(file_path_sans_ext(info_in[["file"]]), "_scatterplot_Sa-Sq", ".pdf")
ggsave(filename = file_out, plot = Sa_Sq, path = "analysis/plots", device = "pdf")

```

```
## Saving 6.5 x 4.5 in image
```

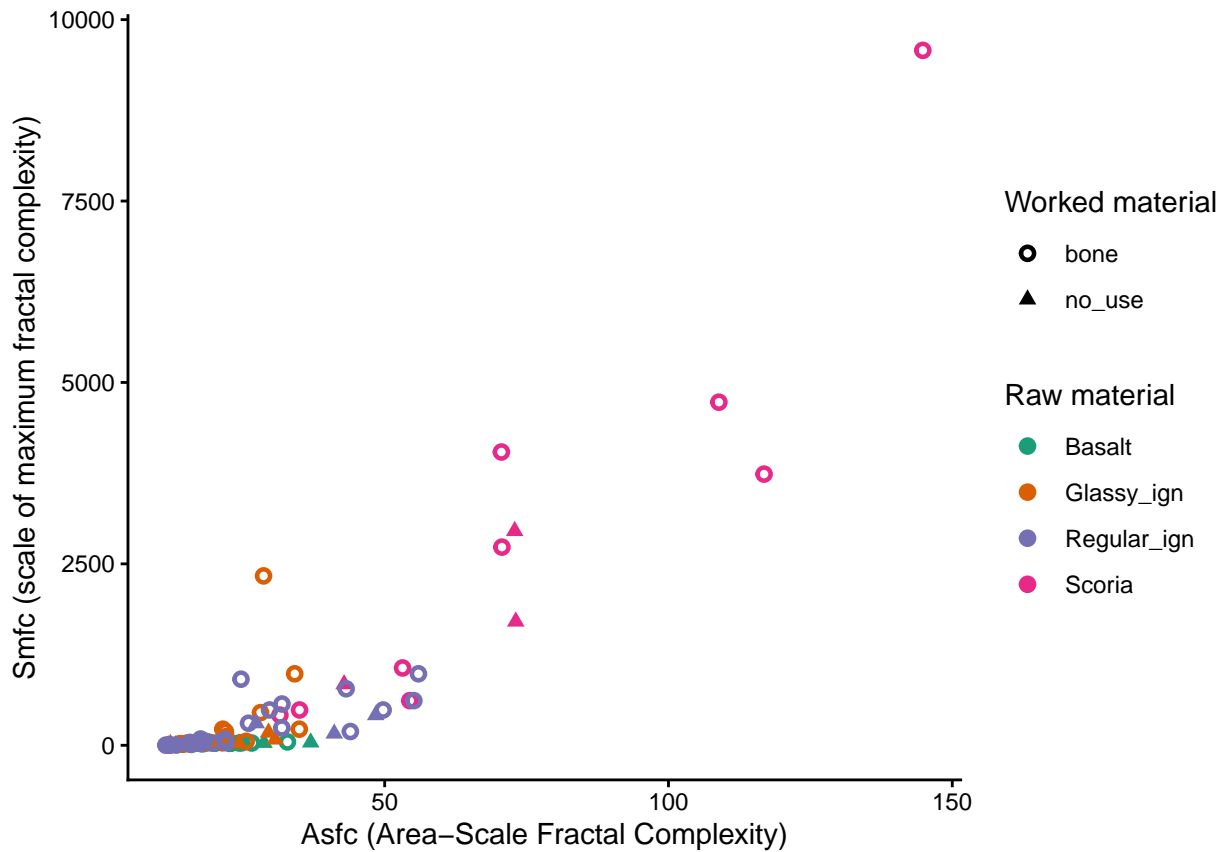
```

# Asfc / Smfc → complexity (use-wear intensity)

asfc_smfc <- ggplot(confocaldata) +
  geom_point(
    aes(
      x = Fractal.complexity..Asfc.,
      y = Scale.of.max.complexity..Smfc.,
      colour = Raw_material,
      shape = Worked_material
    ),
    size = 1.5,
    stroke = 1.2
  ) +
  theme_classic() +
  labs(
    x = "Asfc (Area-Scale Fractal Complexity)",
    y = "Smfc (scale of maximum fractal complexity)",
    colour = "Raw material",
    shape = "Worked material"
  ) +
  scale_colour_brewer(palette = "Dark2") +
  scale_shape_manual(
    values = c(
      bone = 1,
      no_use = 17
    )
  )
)

print(asfc_smfc)

```



```
file_out <- paste0(file_path_sans_ext(info_in[["file"]]), "_scatterplot_asfc_smfc", ".pdf")
ggsave(filename = file_out, plot = asfc_smfc, path = "analysis/plots", device = "pdf")
```

```
## Saving 6.5 x 4.5 in image
```

```
# epLsar vs. Asfc
```

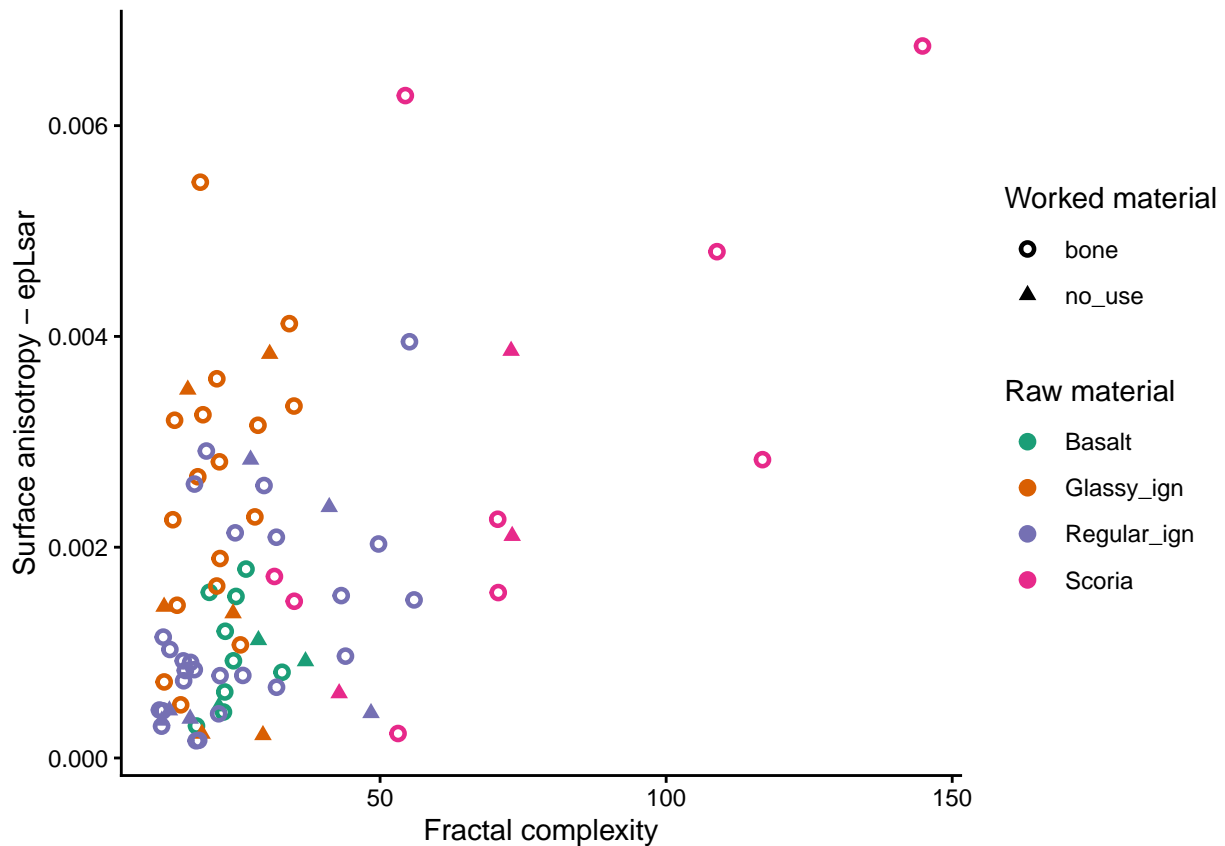
```
ep_As <- ggplot(confocaldata) +
  geom_point(
    aes(
      x = Fractal.complexity..Asfc.,
      y = Length.scale.anisotropy..Sfrax...epLsar.,
      colour = Raw_material,
      shape = Worked_material
    ),
    size = 1.5,
    stroke = 1.2
  ) +
  theme_classic() +
  labs(
    x = "Fractal complexity",
    y = "Surface anisotropy - epLsar",
    colour = "Raw material",
    shape = "Worked material"
  ) +
  scale_colour_brewer(palette = "Dark2") +
```

```

scale_shape_manual(
  values = c(
    bone = 1,
    no_use = 17
  )
)

print(ep_As)

```



```

file_out <- paste0(file_path_sans_ext(info_in[["file"]]), "_scatterplot_Asfc-epLsar", ".pdf")
ggsave(filename = file_out, plot = ep_As, path = "analysis/plots", device = "pdf")

```

```
## Saving 6.5 x 4.5 in image
```

```
# Vmc → load-bearing core volume against Sq
```

```

Sq_Vmc <- ggplot(confocaldata) +
  geom_point(
    aes(
      x = Sq,
      y = Vmc..p...10...q...80...,
      colour = Raw_material,
      shape = Worked_material
    ),
    size = 1.5,

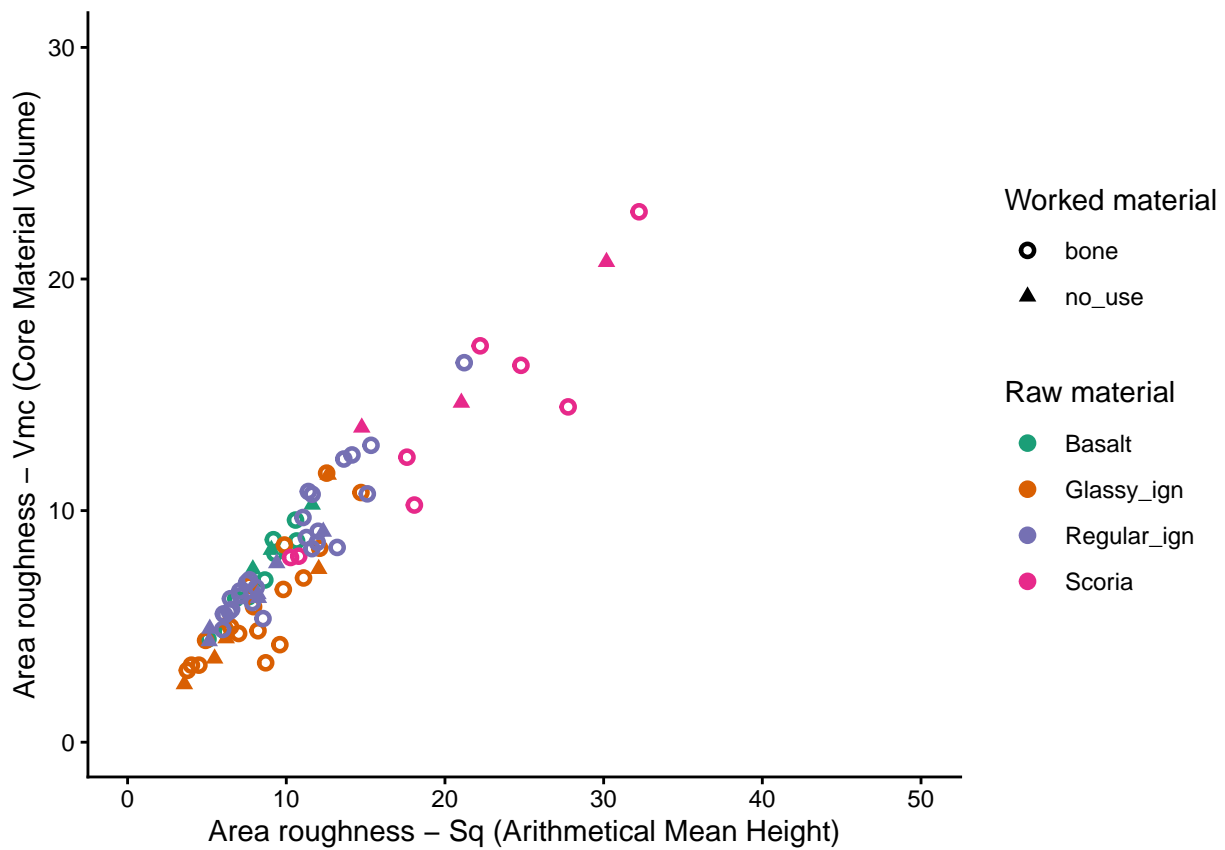
```

```

stroke = 1.2
) +
coord_cartesian(
  xlim = c(0, 50),
  ylim = c(0, 30)
) +
theme_classic() +
labs(
  x = "Area roughness - Sq (Arithmetical Mean Height)",
  y = "Area roughness - Vmc (Core Material Volume)",
  colour = "Raw material",
  shape = "Worked material"
) +
scale_colour_brewer(palette = "Dark2") +
scale_shape_manual(
  values = c(
    bone = 1,
    no_use = 17
  )
)
)

print(Sq_Vmc)

```



```

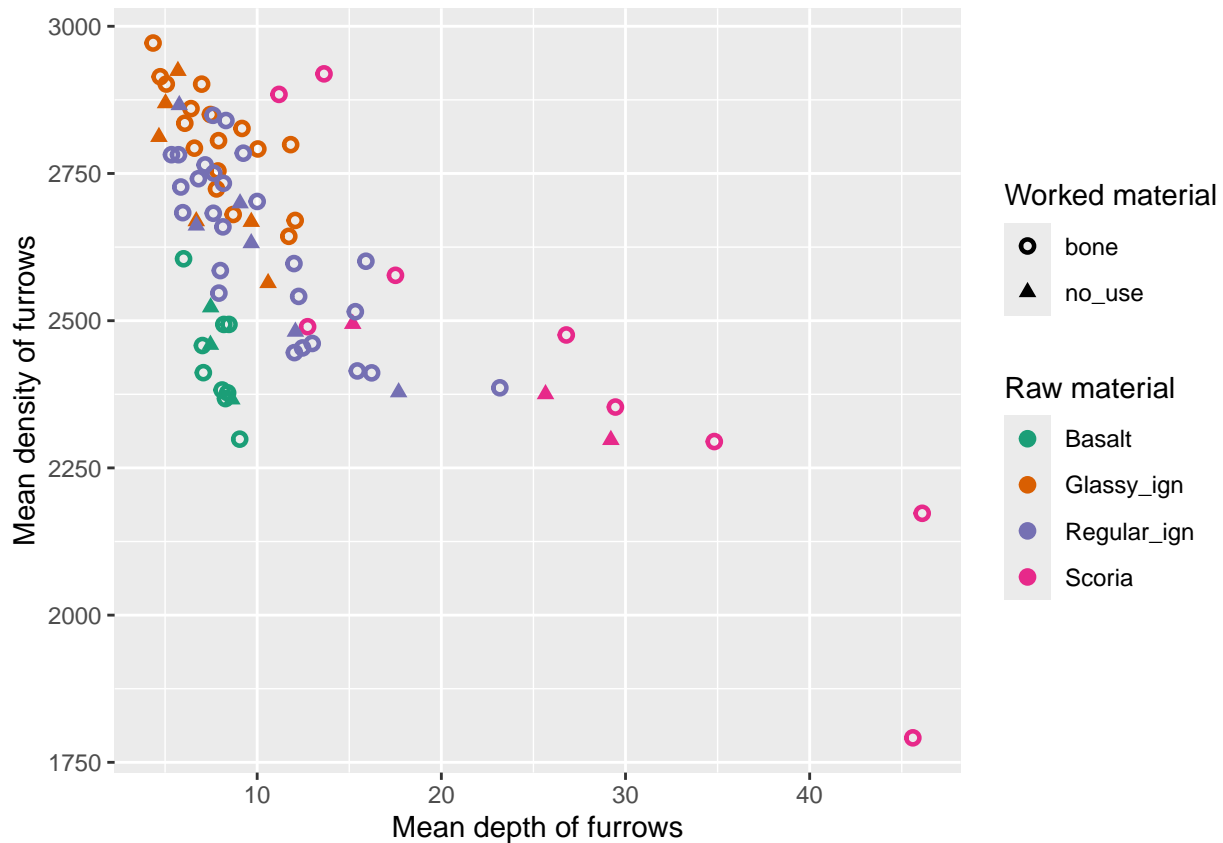
file_out <- paste0(file_path_sans_ext(info_in[["file"]]), "_scatterplot_Sq-Vmc", ".pdf")
ggsave(filename = file_out, plot = Sq_Vmc, path = "analysis/plots", device = "pdf")

```

```
## Saving 6.5 x 4.5 in image
```

```
# Furrow depth & density → contact mechanics
```

```
furrows <- ggplot(confocaldata) +  
  geom_point(  
    aes(  
      x = Mean.depth.of.furrows,  
      y = Mean.density.of.furrows,  
      colour = Raw_material,  
      shape = Worked_material  
    ),  
    size = 1.5,  
    stroke = 1.2  
  ) +  
  labs(  
    x = "Mean depth of furrows",  
    y = "Mean density of furrows",  
    colour = "Raw material",  
    shape = "Worked material"  
  ) +  
  scale_colour_brewer(palette = "Dark2") +  
  scale_shape_manual(  
    values = c(  
      bone = 1,  
      no_use = 17  
    )  
  )  
  
print(furrows)
```



```
file_out <- paste0(file_path_sans_ext(info_in[["file"]]), "_scatterplot_furrows", ".pdf")
ggsave(filename = file_out, plot = furrows, path = "analysis/plots", device = "pdf")
```

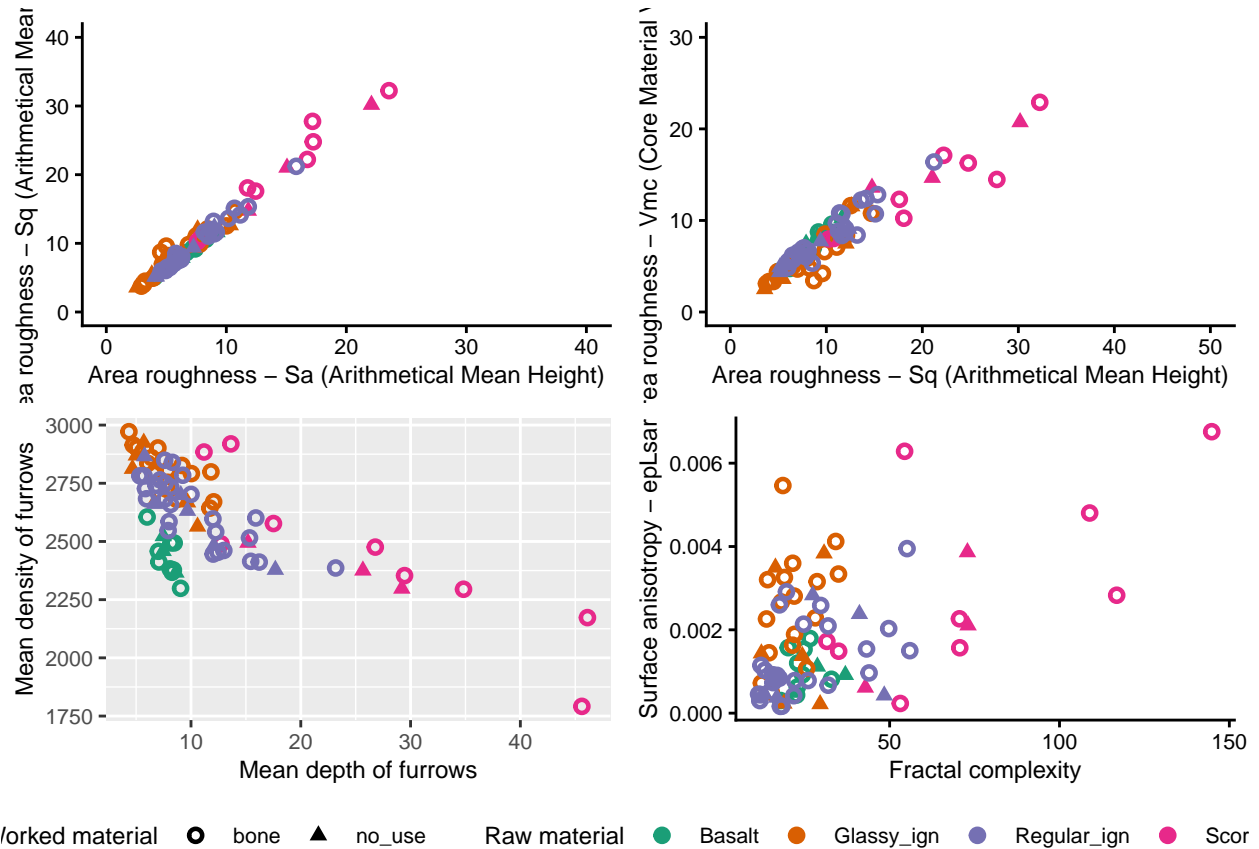
```
## Saving 6.5 x 4.5 in image
```

```
# combine all in a single image (rearrange sizes)
```

```
small_theme <- theme(
  axis.title = element_text(size = 9),
  axis.text  = element_text(size = 8),
  legend.title = element_text(size = 9),
  legend.text  = element_text(size = 8)
)
```

```
Sa_Sq <- Sa_Sq + small_theme
Sq_Vmc <- Sq_Vmc + small_theme
furrows <- furrows + small_theme
ep_As <- ep_As + small_theme
```

```
ggarrange(
  Sa_Sq, Sq_Vmc, furrows, ep_As,
  common.legend = TRUE,
  legend = "bottom"
)
```



```
ggsave("analysis/plots/scatterplots.png")
```

```
## Saving 6.5 x 4.5 in image
```

### Abbott–Firestone–derived volume parameters (raw material vs worked material)

```
af_long <- confocaldata %>%
  pivot_longer(
    cols = c(Vmc..p...10...q...80..., Vvc..p...10...q...80..., Vvv..p...80...),
    names_to = "Parameter",
    values_to = "Value"
  )

# only Vmc
p_vmc <- ggplot(
  af_long %>% filter(Parameter == "Vmc..p...10...q...80..."),
  aes(
    x = Worked_material,
    y = Value,
    fill = Worked_material
  )
) +
  geom_boxplot(
```

```

alpha = 0.7,
outlier.shape = 21,
outlier.size = 2
) +
facet_wrap(~ Raw_material, scales = "free_y") +
labs(
  title = "",
  x = "Worked material",
  y = expression(Vmc~(mu*m^3/mu*m^2)),
) +
theme_minimal(base_size = 14) +
theme(
  legend.position = "none",
  strip.text = element_text(face = "bold")
)

# all together

p_af_all <- ggplot(
  af_long,
  aes(
    x = Worked_material,
    y = Value,
    fill = Worked_material
  )
) +
geom_boxplot(alpha = 0.7, outlier.shape = NA) +
facet_grid(
  Parameter ~ Raw_material,
  scales = "free_y"
) +
labs(
  title = "Abbott-Firestone-Derived Volume Parameters",
  x = "Worked material",
  y = expression(mu*m^3/mu*m^2)
) +
theme_minimal(base_size = 13) +
theme(
  legend.position = "none",
  strip.text = element_text(face = "bold"),
  axis.text.x = element_text(angle = 45, hjust = 1)
)

# "Higher Vmc values indicate an expansion of the core material portion of the Abbott-Firestone curve,

```

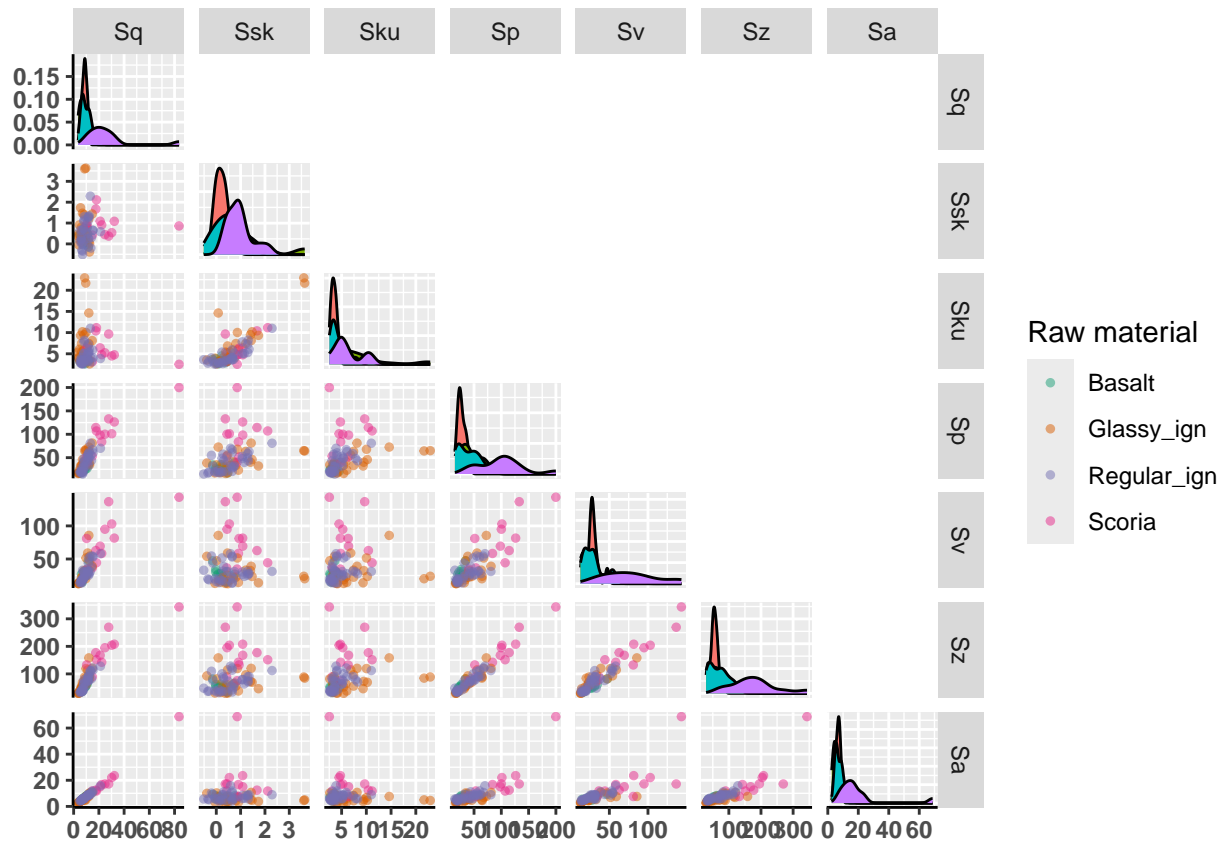
Scatterplot matrix for the ISO 25178 Area scale, Height and volume parameters

```

ggpairs(
  data = confocaldata,
  columns = 23:29,
  cardinality_threshold = 30,

```





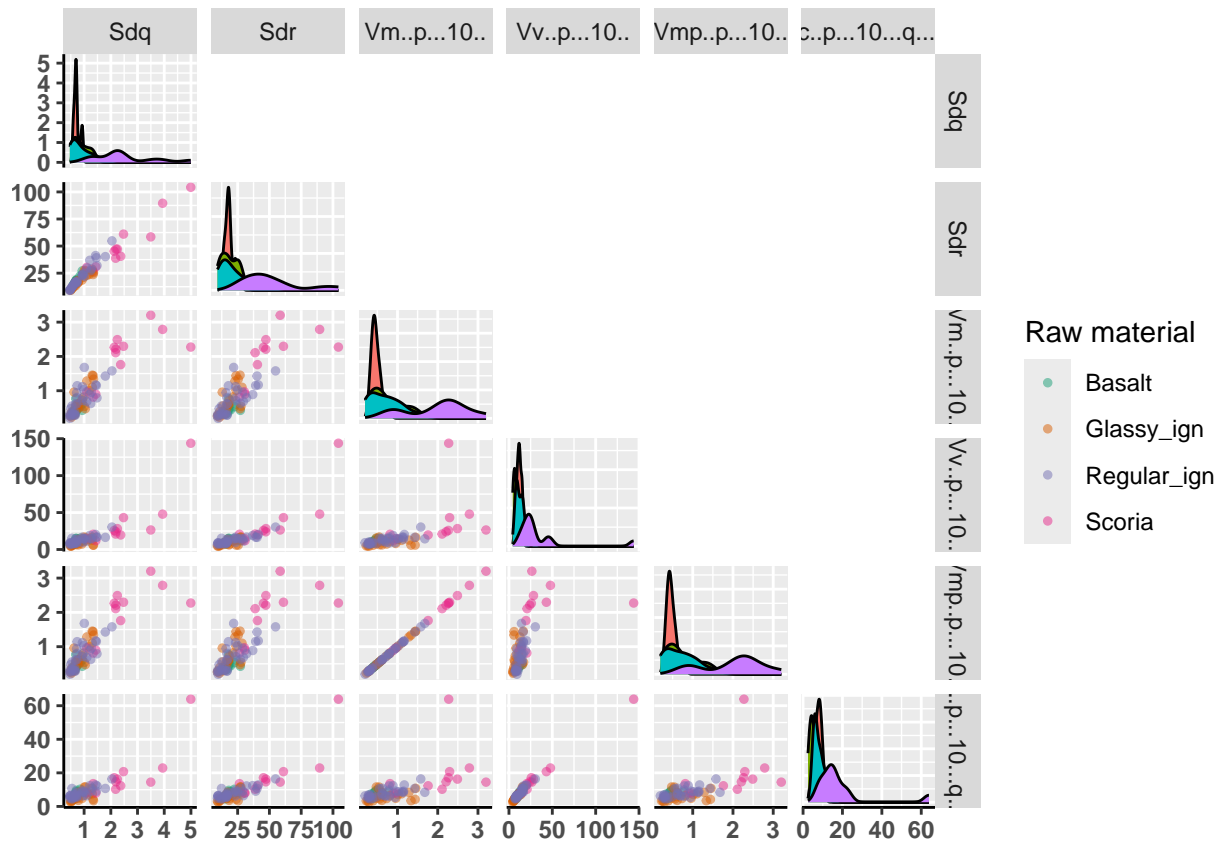
```
ggsave("analysis/plots/confocalarea_matrix.png")
```

```
## Saving 6.5 x 4.5 in image
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
```

```
# Volume parameters
```

```
ggpairs(
  data = confocaldata,
  columns = 36:41,
  cardinality_threshold = 30,
  mapping = ggplot2::aes(colour = Raw_material),
  lower = list(
```





```
ggsave("analysis/plots/confocalvolume_matrix.png")
```

```
## Saving 6.5 x 4.5 in image
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
## Ignoring unknown labels:
## * colour : "Raw material"
```

## Plot confostats for the ISO 25178 Area-scale, Height and volume parameters

```
heightconfostats <- select(confostats, Worked_material, Raw_material, Sq.mean, Ssk.mean, Sku.mean, Sp.mean,
library(ggplot2)

p1 <- ggplot(heightconfostats,
aes(x = Raw_material, y = Sq.mean, colour = Worked_material)) +
```

```

geom_boxplot() +
scale_colour_brewer(palette = "Dark2") +
theme_classic() +
labs(x = "", colour = "Raw material")

p2 <- ggplot(heightconfostats,
             aes(x = Raw_material, y = Ssk.mean, colour = Worked_material)) +
geom_boxplot() +
scale_colour_brewer(palette = "Dark2") +
theme_classic() +
labs(x = "", colour = "Raw material")

p3 <- ggplot(heightconfostats,
             aes(x = Raw_material, y = Sku.mean, colour = Worked_material)) +
geom_boxplot() +
scale_colour_brewer(palette = "Dark2") +
theme_classic() +
labs(x = "", colour = "Raw material")

p4 <- ggplot(heightconfostats,
             aes(x = Raw_material, y = Sp.mean, colour = Worked_material)) +
geom_boxplot() +
scale_colour_brewer(palette = "Dark2") +
theme_classic() +
labs(x = "", colour = "Raw material")

p5 <- ggplot(heightconfostats,
             aes(x = Raw_material, y = Sv.mean, colour = Worked_material)) +
geom_boxplot() +
scale_colour_brewer(palette = "Dark2") +
theme_classic() +
labs(x = "", colour = "Raw material")

p6 <- ggplot(heightconfostats,
             aes(x = Raw_material, y = Sz.mean, colour = Worked_material)) +
geom_boxplot() +
scale_colour_brewer(palette = "Dark2") +
theme_classic() +
labs(x = "", colour = "Raw material")

p7 <- ggplot(heightconfostats,
             aes(x = Raw_material, y = Sa.mean, colour = Worked_material)) +
geom_boxplot() +
scale_colour_brewer(palette = "Dark2") +
theme_classic() +
labs(x = "", colour = "Raw material")

# joint and rescale plots

small_theme <- theme(
  axis.title = element_text(size = 9),
  axis.text  = element_text(size = 8),
  legend.title = element_text(size = 9),

```

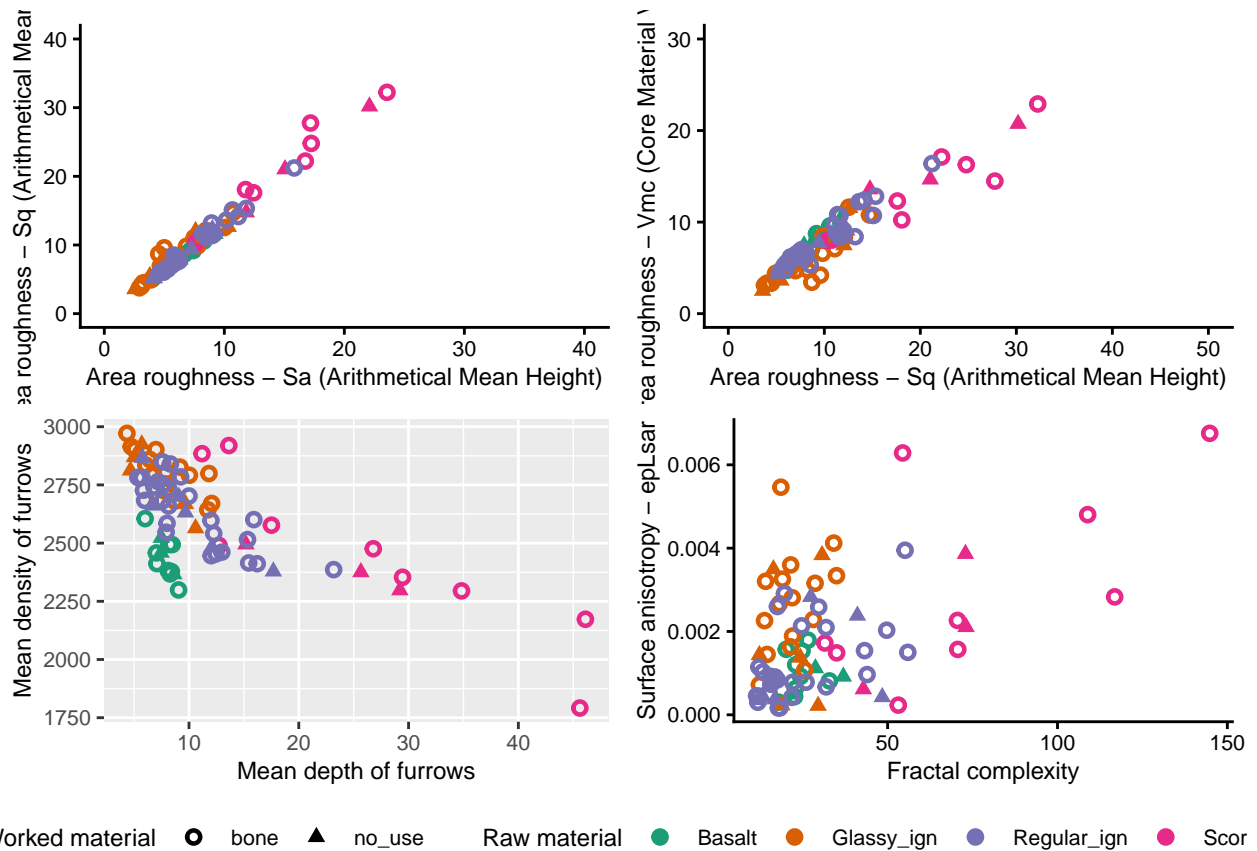
```

legend.text = element_text(size = 8)
)

Sa_Sq <- Sa_Sq + small_theme
Sq_Vmc <- Sq_Vmc + small_theme
furrows <- furrows + small_theme
ep_As <- ep_As + small_theme

ggarrange(
  Sa_Sq, Sq_Vmc, furrows, ep_As,
  common.legend = TRUE,
  legend = "bottom"
)

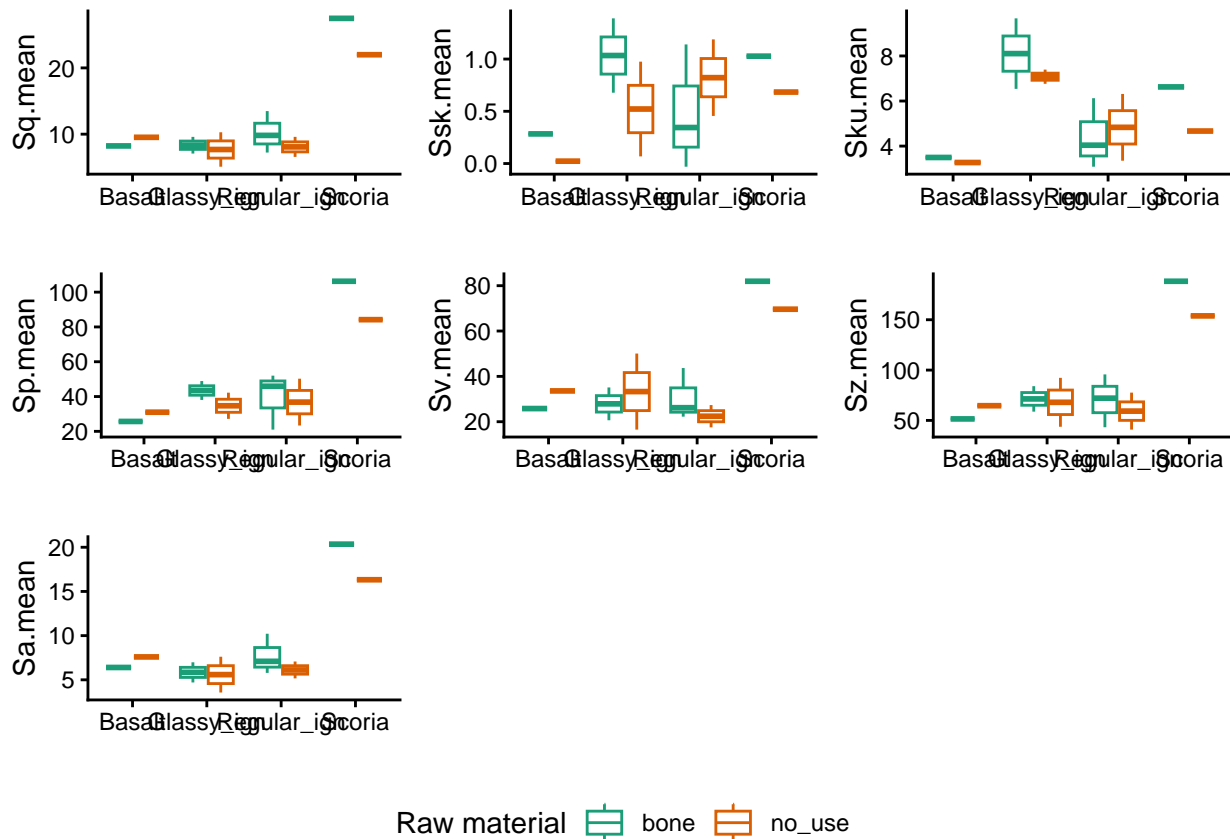
```



```

ggarrange(p1, p2, p3, p4, p5, p6, p7, common.legend = TRUE, font.label = list(size=8), legend="bottom")

```



```
ggsave("analysis/plots/confostats_boxplots.png")
```

```
## Saving 6.5 x 4.5 in image
```

```
# Now Volume parameters
```

```
volumeconfostats <- select(confostats,Raw_material,Worked_material, Vm..p...10...mean,Vv..p...10...mean
```

```
p8 <- ggplot(volumeconfostats, aes(x=Raw_material, y=Vm..p...10...mean, colour=Worked_material)) +
  geom_boxplot() +
  scale_colour_brewer(palette = "Dark2") +
  theme_classic() +
  labs(x="", colour="Raw material")
```

```
p9 <- ggplot(volumeconfostats, aes(x=Raw_material, y=Vv..p...10...mean, colour=Worked_material)) +
  geom_boxplot() +
  scale_colour_brewer(palette = "Dark2") +
  theme_classic() +
  labs(x="", colour="Raw material")
```

```
p10 <- ggplot(volumeconfostats, aes(x=Raw_material, y=Vmp..p...10...mean, colour=Worked_material)) +
  geom_boxplot() +
  scale_colour_brewer(palette = "Dark2") +
  theme_classic() +
  labs(x="", colour="Raw material")
```

```

p11 <- ggplot(volumeconfostats, aes(x=Raw_material, y=Vmc..p...10...q...80...mean, colour=Worked_material)) +
  geom_boxplot() +
  scale_colour_brewer(palette = "Dark2") +
  theme_classic() +
  labs(x="", colour="Raw material")

p12 <- ggplot(volumeconfostats, aes(x=Raw_material, y=Vvc..p...10...q...80...mean, colour=Worked_material)) +
  geom_boxplot() +
  scale_colour_brewer(palette = "Dark2") +
  theme_classic() +
  labs(x="", colour="Raw material")

p13 <- ggplot(volumeconfostats, aes(x=Raw_material, y=Vvv..p...80...mean, colour=Worked_material)) +
  geom_boxplot() +
  scale_colour_brewer(palette = "Dark2") +
  theme_classic() +
  labs(x="", colour="Raw material")

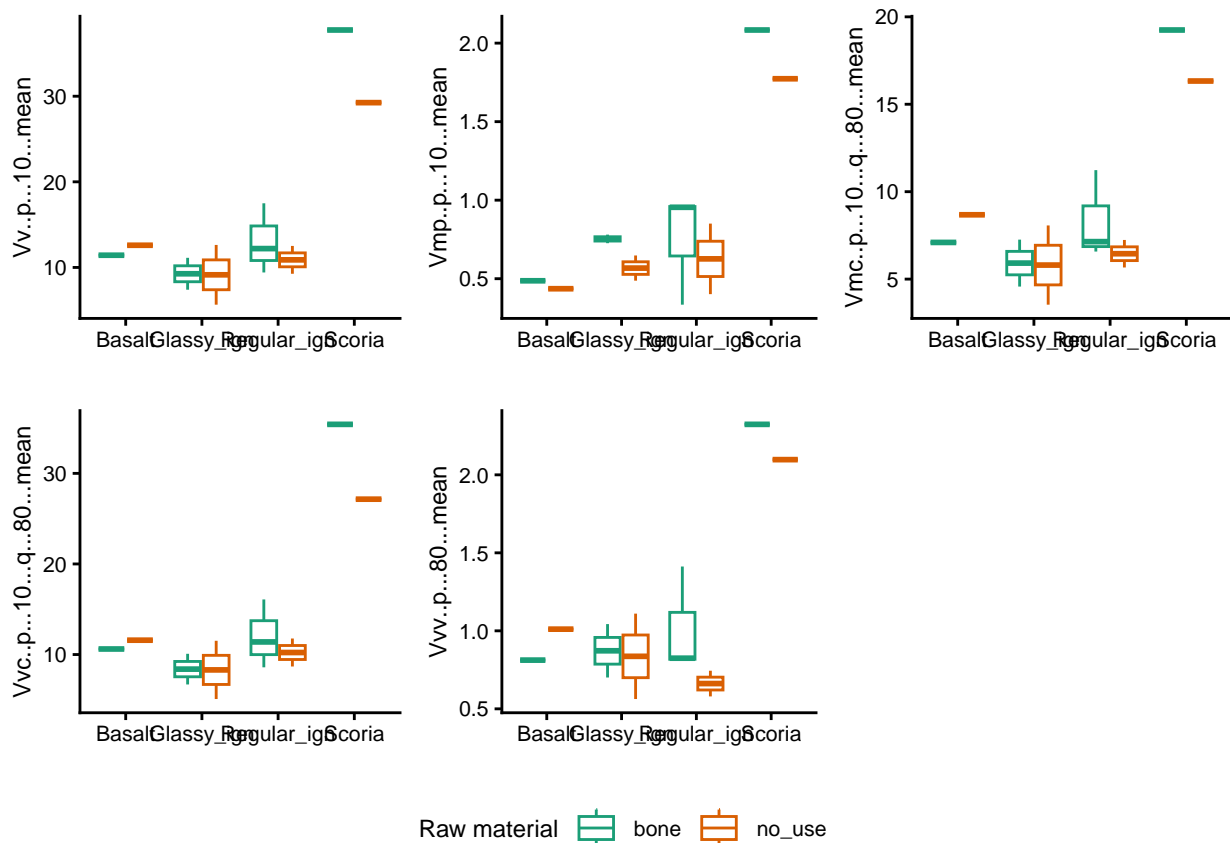
# joint and rescale plots

small_theme <- theme(
  axis.title = element_text(size = 9),
  axis.text = element_text(size = 8),
  legend.title = element_text(size = 9),
  legend.text = element_text(size = 8)
)

p8 <- p8 + small_theme
p9 <- p9 + small_theme
p10 <- p10 + small_theme
p11 <- p11 + small_theme
p12 <- p12 + small_theme
p13 <- p13 + small_theme

ggarrange(
  p9, p10, p11, p12, p13,
  common.legend = TRUE,
  legend = "bottom"
)

```



```
ggsave("analysis/plots/confostatsvolume_boxplots.png")
```

```
## Saving 6.5 x 4.5 in image
```

## Multivariate pattern exploration, PCA biplot

```
pca_vars <- c(
  "Sq", "Ssk", "Sku",
  "Vmc.p...10...q...80..", "Vvv.p...80..",
  "Fractal.complexity..Asfc.", "Scale.of.max.complexity..Smfc.",
  "Str..s...0.2."
)

pca_data <- confocaldata %>%
  dplyr::select(all_of(c("Raw_material", "Worked_material", pca_vars))) %>%
  drop_na()

pca_data_clean <- pca_data %>%
  mutate(
    across(
      all_of(pca_vars),
      ~ suppressWarnings(as.numeric(.))
    )
  )
```

```

pca_data_clean <- pca_data_clean %>%
  drop_na(all_of(pca_vars))
pca_res <- PCA(
  pca_data_clean %>% dplyr::select(all_of(pca_vars)),
  scale.unit = TRUE,
  graph = FALSE
)

pca_res <- PCA(
  pca_data_clean %>% dplyr::select(all_of(pca_vars)),
  scale.unit = TRUE,
  graph = FALSE
)

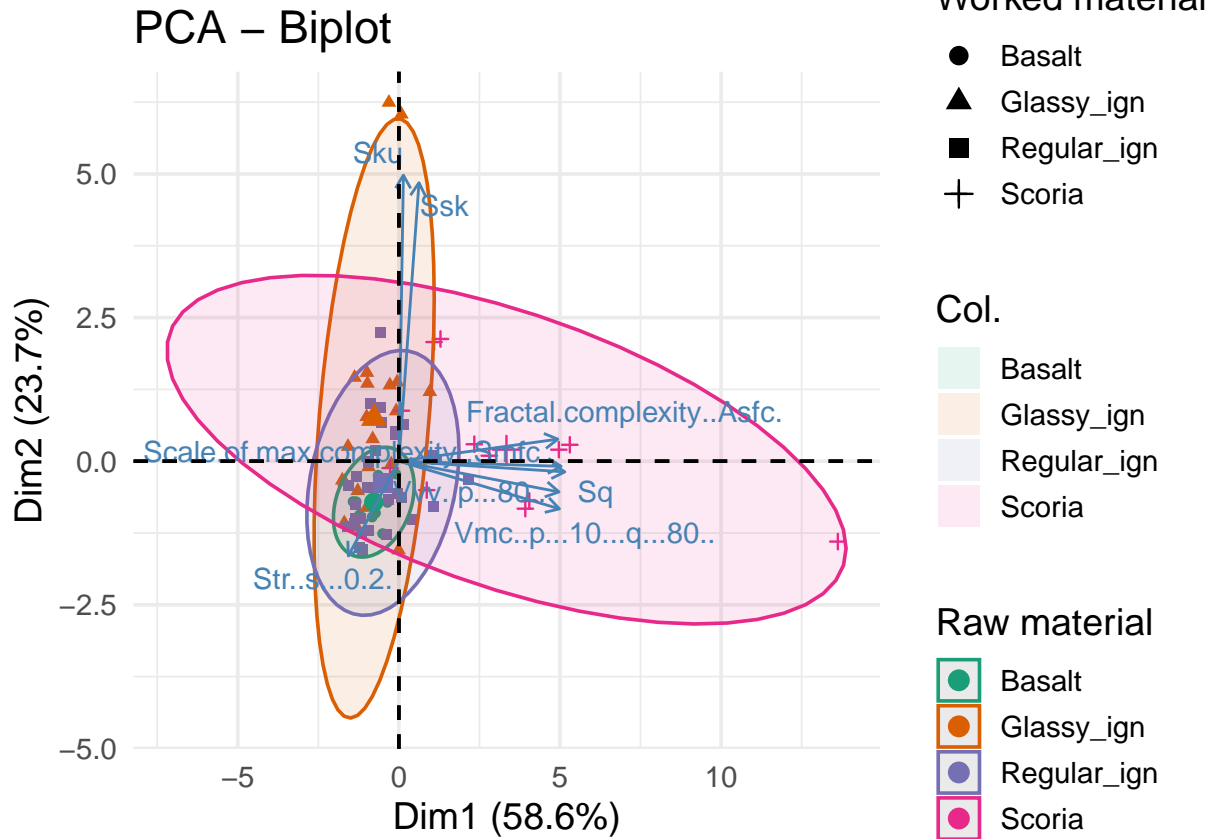
fviz_pca_biplot(
  pca_res,
  geom.ind = "point",
  col.ind = pca_data_clean$Raw_material,
  shape.ind = as.factor(pca_data_clean$Worked_material),
  palette = "Dark2",
  addEllipses = TRUE,
  ellipse.level = 0.95,
  label = "var",
  repel = TRUE,
  legend.title = list(
    col = "Raw material",
    shape = "Worked material"
  )
) +
  theme_minimal(base_size = 14)

```

```

## Warning: Using 'size' aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use 'linewidth' instead.
## i The deprecated feature was likely used in the ggpubr package.
##   Please report the issue at <https://github.com/kassambara/ggpubr/issues>.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.

```



## End and Session info

```
sessionInfo()
```

```
## R version 4.3.0 (2023-04-21)
## Platform: aarch64-apple-darwin20 (64-bit)
## Running under: macOS 26.2
##
## Matrix products: default
## BLAS: /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/lib/libRlapack.dylib; LAPACK v
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## time zone: Europe/Berlin
## tzcode source: internal
##
## attached base packages:
## [1] tools stats graphics grDevices utils datasets methods
## [8] base
##
## other attached packages:
```

```

## [1] factoextra_1.0.7 FactoMineR_2.11 ggthemes_5.1.0 ggpubr_0.6.2
## [5] doBy_4.7.0 GGally_2.4.0 kableExtra_1.4.0 janitor_2.2.1
## [9] knitr_1.50 lubridate_1.9.4 forcats_1.0.1 stringr_1.6.0
## [13] dplyr_1.1.4 purrr_1.2.0 readr_2.1.6 tidyr_1.3.1
## [17] tibble_3.3.0 ggplot2_4.0.1 tidyverse_2.0.0
##
## loaded via a namespace (and not attached):
## [1] tidysselect_1.2.1 viridisLite_0.4.2 farver_2.1.2
## [4] S7_0.2.1 fastmap_1.2.0 TH.data_1.1-5
## [7] digest_0.6.39 timechange_0.3.0 estimability_1.5.1
## [10] lifecycle_1.0.4 cluster_2.1.8.1 Deriv_4.2.0
## [13] multcompView_0.1-10 survival_3.8-3 magrittr_2.0.4
## [16] compiler_4.3.0 rlang_1.1.6 yaml_2.3.10
## [19] ggsignif_0.6.4 labeling_0.4.3 htmlwidgets_1.6.4
## [22] bit_4.6.0 scatterplot3d_0.3-44 xml2_1.3.8
## [25] RColorBrewer_1.1-3 abind_1.4-8 multcomp_1.4-29
## [28] withr_3.0.2 grid_4.3.0 xtable_1.8-4
## [31] emmeans_2.0.0 scales_1.4.0 MASS_7.3-60.0.1
## [34] tinytex_0.58 flashClust_1.01-2 cli_3.6.5
## [37] mvtnorm_1.3-3 crayon_1.5.3 rmarkdown_2.30
## [40] ragg_1.5.0 generics_0.1.4 rstudioapi_0.17.1
## [43] modelr_0.1.11 tzdb_0.5.0 splines_4.3.0
## [46] parallel_4.3.0 vctrs_0.6.5 boot_1.3-32
## [49] Matrix_1.6-5 sandwich_3.1-1 carData_3.0-5
## [52] car_3.1-3 hms_1.1.4 bit64_4.6.0-1
## [55] rstatix_0.7.3 ggrepel_0.9.6 Formula_1.2-5
## [58] systemfonts_1.3.1 glue_1.8.0 ggstats_0.11.0
## [61] codetools_0.2-20 DT_0.33 cowplot_1.2.0
## [64] stringi_1.8.7 gtable_0.3.6 pillar_1.11.1
## [67] htmltools_0.5.8.1 R6_2.6.1 microbenchmark_1.5.0
## [70] textshaping_1.0.4 vroom_1.6.5 evaluate_1.0.5
## [73] lattice_0.22-7 backports_1.5.0 leaps_3.2
## [76] broom_1.0.8 snakecase_0.11.1 Rcpp_1.1.0
## [79] gridExtra_2.3 svglite_2.2.1 xfun_0.54
## [82] zoo_1.8-14 pkgconfig_2.0.3

```