



The Use of ArcGIS for Data Collection and Sample Characterization of the Urban Acoustic Environment

The example in SALVE

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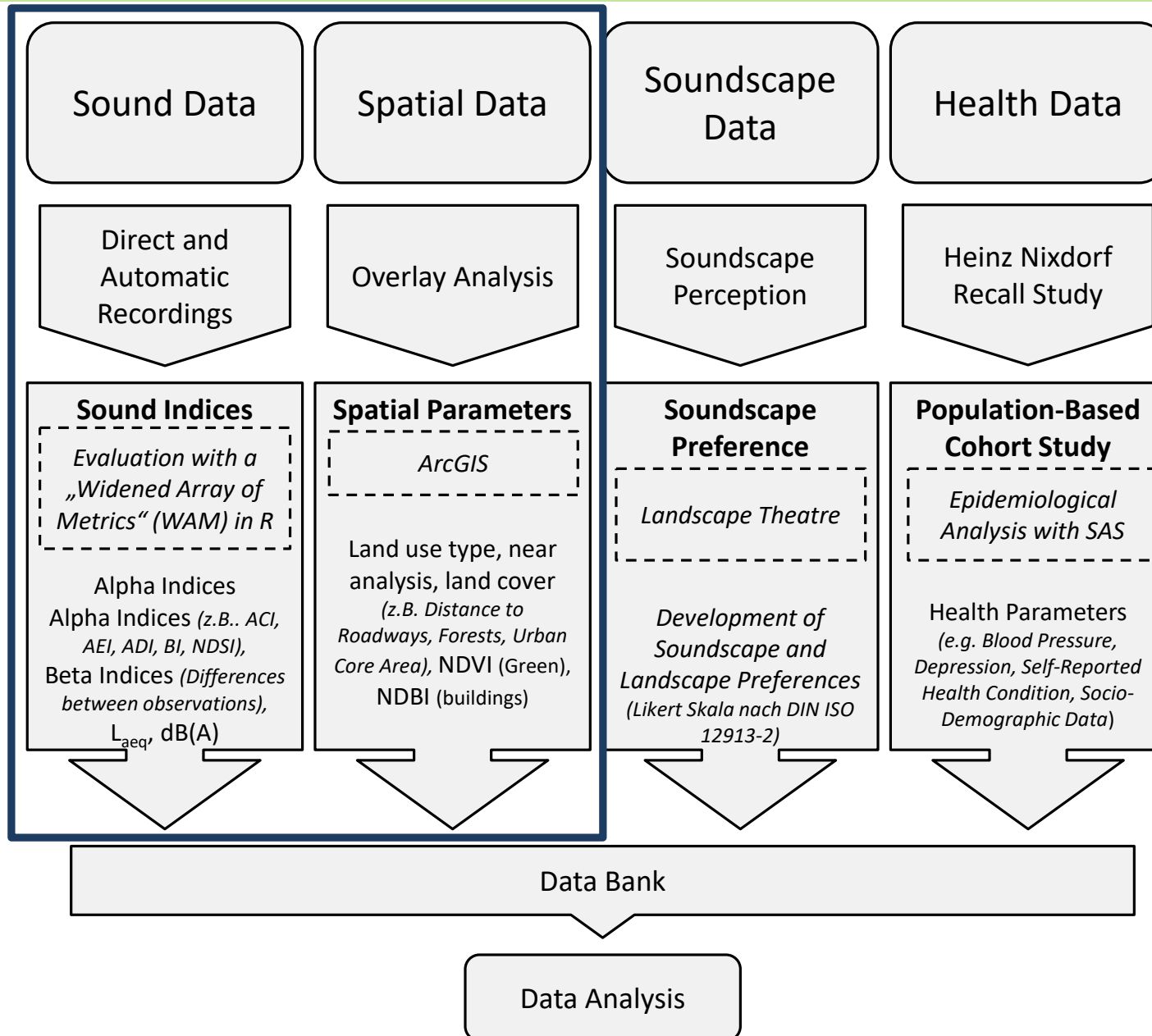


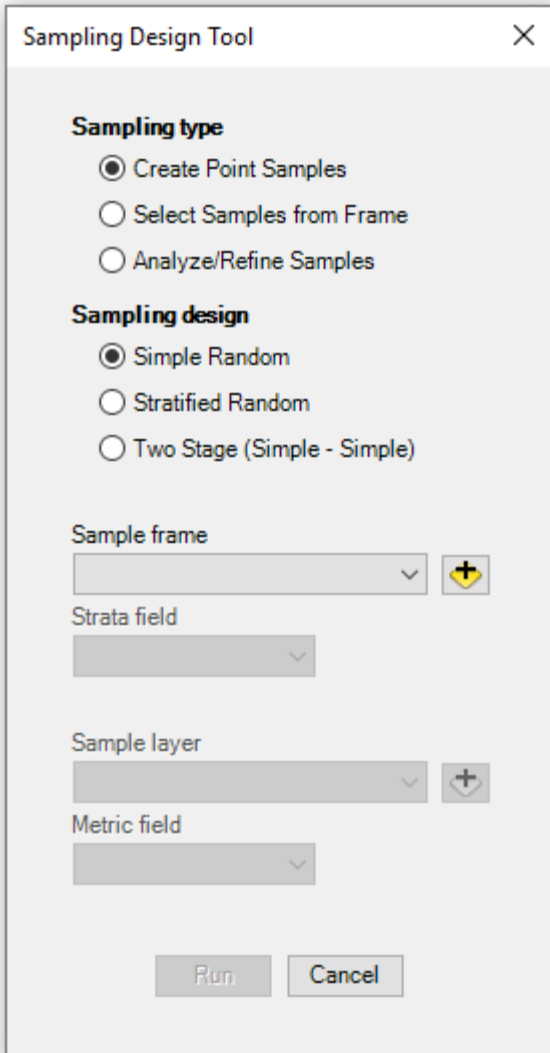
- In health research, health policy, and spatial planning the focus is limited to noise
- The acoustic environment below the ‘noise’ threshold is not fully investigated
 - opportunity for frequency and sound composition investigations related to cities and health
- Relationship between the built and acoustic environment and health of interest to spatial planning and epidemiology

1. Measurements of different acoustic environments in Bochum, Germany under consideration of spatial-temporal structures (landuse, season, time of day) - focus on human habitation
 - *Broaden automated and comprehensive procedures for the classification of sound data (dB, Hz, time) into indices (metrics)*
2. Analysis of associations between sound indices, spatial structure and health in a population cohort (HNR)
3. Survey to determine soundscape and landscape preferences, together

Interdisciplinary Research Field
Healthy Urban Soundscapes

SALVE Data Structure





The screenshot shows the 'Sampling Design Tool' dialog box with the following settings:

- Sampling type:** Create Point Samples, Select Samples from Frame, Analyze/Refine Samples
- Sampling design:** Simple Random, Stratified Random, Two Stage (Simple - Simple)
- Sample frame:** [Empty dropdown menu]
- Strata field:** [Empty dropdown menu]
- Sample layer:** [Empty dropdown menu]
- Metric field:** [Empty dropdown menu]

Buttons: Run, Cancel



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Sample Design Tool Add-In

1. Spatial sampling

2. Scalable data requirements

3. Random selection

4. Multiple sampling designs: Simple, stratified, and two-stage sampling designs.

5. Sample unit-based sampling: Points or polygons are selected from a sample frame.

6. Area-based sampling: Random points are generated within a polygon.

7. Analysis: set sample sizes; allocate samples among strata.

8. Computations: Mean, standard error, and confidence intervals

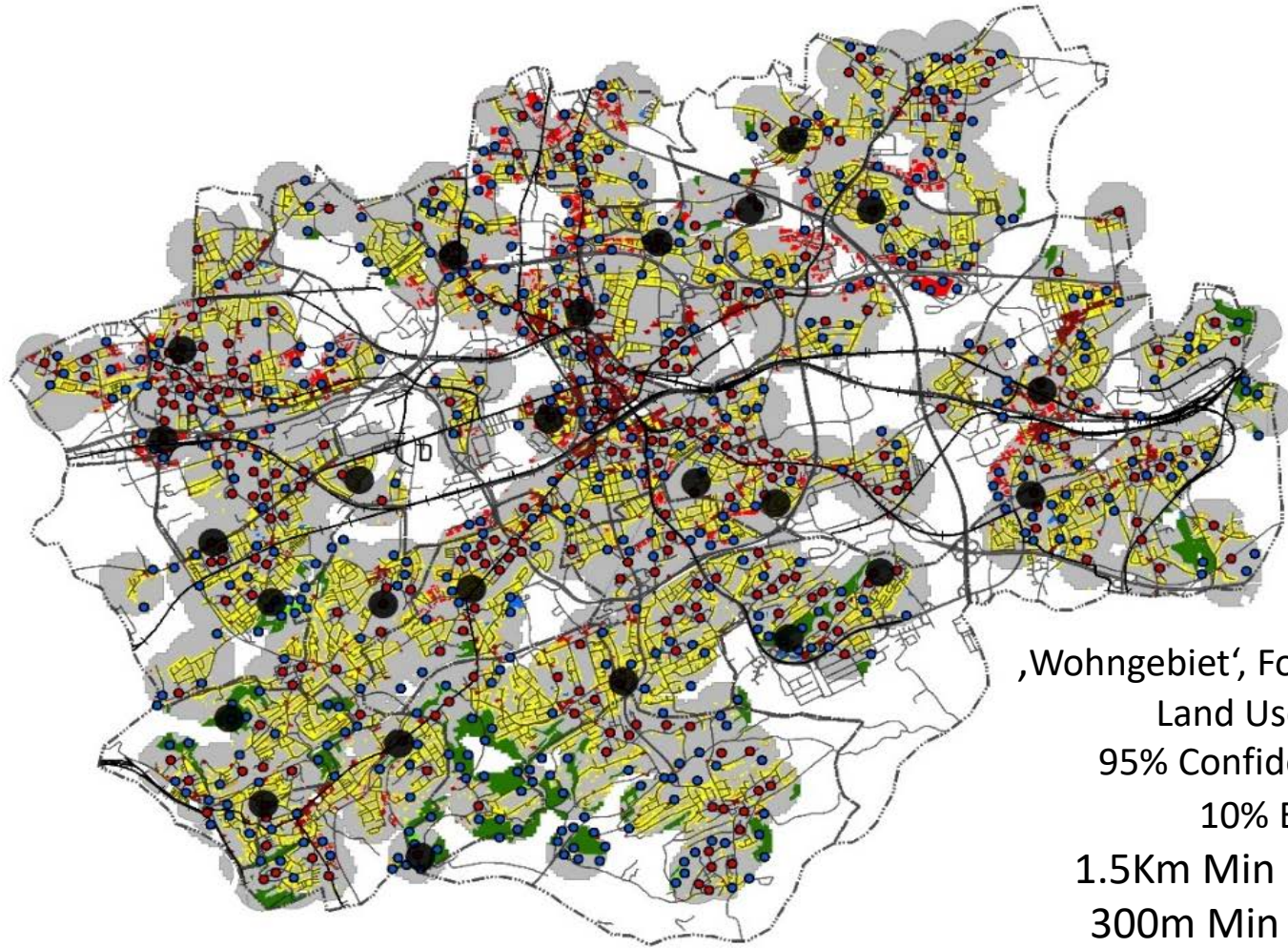
9. Output: Geographic positions; sample size estimates; sample statistics

Source: <https://coastalscience.noaa.gov/project/sampling-design-tool-arcgis/>

Sample Design – Bochum, Germany



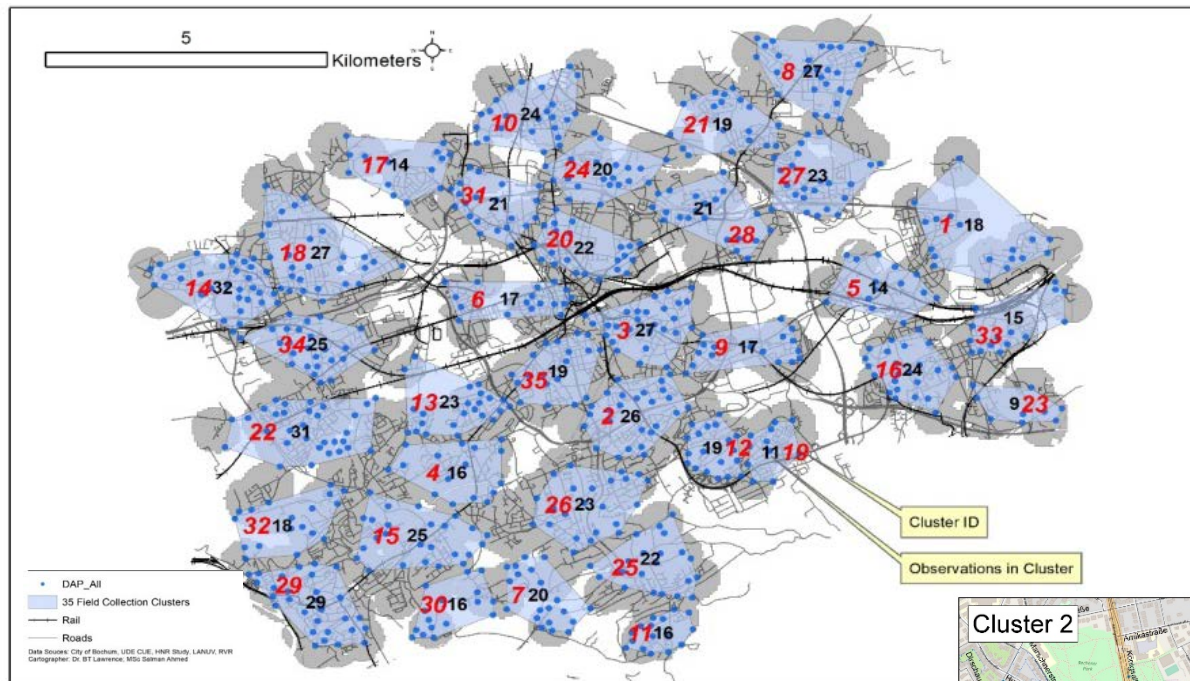
Bochum



,Wohngebiet', Forests, Park
 Land Use Stratified
 95% Confidence (DAP)
 10% Error (DAP)
 1.5Km Min Dist AADs
 300m Min Dist DAPs

<ul style="list-style-type: none"> ● Automated Aural Devices ● DAP Participant Polygon Samples ● DAP Participant Envelope Samples — Rail — Roads ■ HNR Participant Envelope 	<p>Sampled Land Use Areas</p> <p>Land Use Category</p> <ul style="list-style-type: none"> ■ Bebaute Flaechen, dem Wohnen dienend, bis 3 Geschosse ■ Bebaute Flaechen, dem Wohnen dienend, bis 5 Geschosse ■ Bebaute Flaechen, dem Wohnen dienend, ueber 5 Geschosse ■ Gemeindebedarfsflaechen (Kindergaerten, Hort, Jugend- u. Altenheime / Wohnanlagen) 	<ul style="list-style-type: none"> ■ Gewerbeflaechen (Gebaeude / Anlagen) ■ Mischbauflaechen ■ Landwirtschaftliche Hof- und Gebaeudeflaechen (Gebaeude u. Anlagen) ■ Gruenflaechen (gestaltet, im hausnahen Bereich der Kriterien 10,20,30,40) ■ Laubwald ■ Mischwald
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Direct Sampling – Networked Clusters



Grouping Analysis in ArcGIS

- >Spatial Statistics
 - >Mapping Clusters
- K=20 Nearest Neighbors Alg.*

Minimum Bounding Geom.

- >Data Management
 - >Features
- Convex Hull*

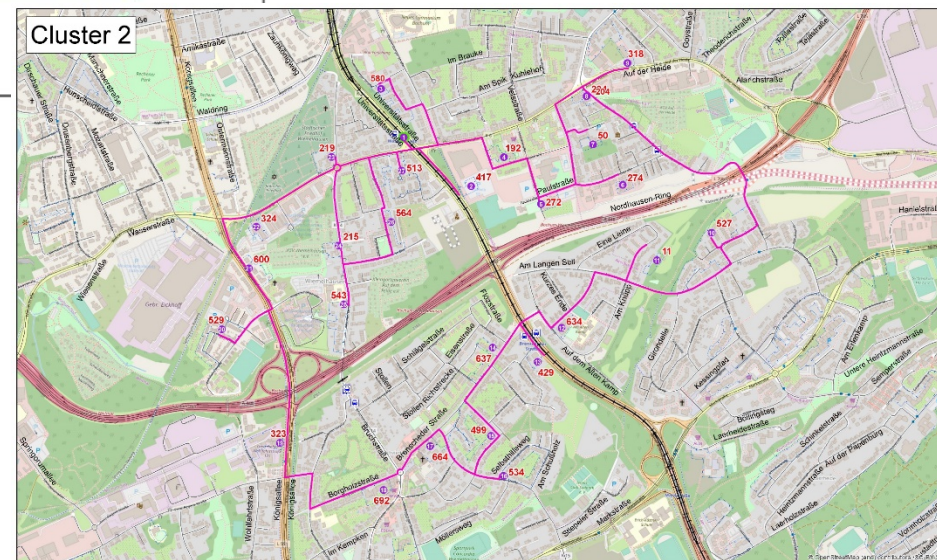
Network Analyst

- >Network Dataset (B.Rds)
- >Distance Field
- >Time Field
- > 500m Prox.
- >Two-Way Travel Allowed

Manpower Estimate

NOTES:

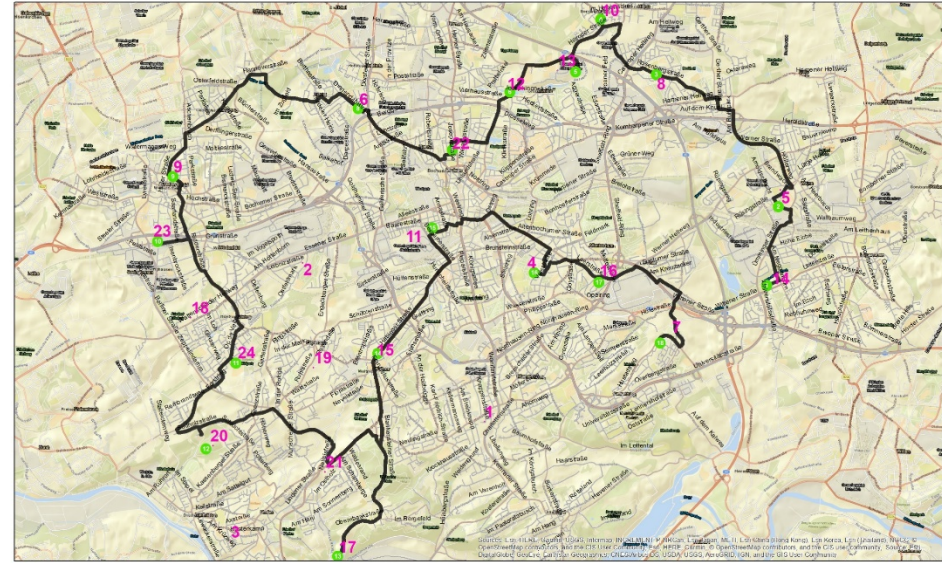
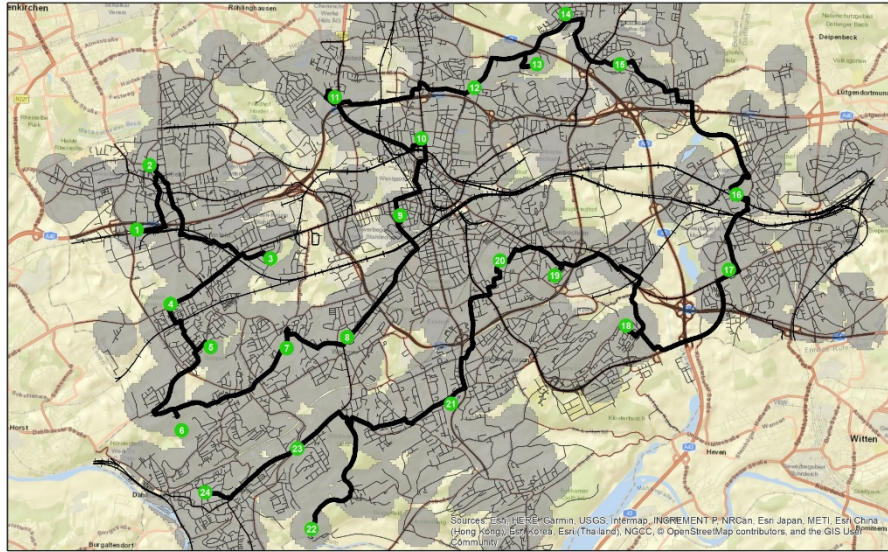
Total Distance: 13 km
 Total Travel Time at 5 km/h: 2 hrs 30 min
 Total Data Points: 26 (8 mins per point)
 Total Estimated time: 6 hrs 30 min



NOTES:
 Total Distance: 13 km
 Total Travel Time at 5 km/h: 2 hrs 30 min
 Total Data Points: 26 (8 mins per point)
 Total Estimated time: 6 hrs 30 min

0 0.125 0.25 0.5 0.75 1 Kilometers

Automated Device Data Harvest



Network Analyst affords a flexible and open access approach to extensive long-term and geographically complex field data collection.



Summary of Collected Data



Device	Sample Points	Sample Days per year	Measurement Time (in min.)	Total recorded time (in min.)
Manual (DAP)	730	4	5	14 600
Automated (AAP)	24	48 x 365	3	1 261 440
Binaural (BR)	24	8	3	576



**1 276 616 min \cong 21 276 hours \cong 886 days (equiv.)
of audible sound data collected in Bochum**

Widened Array of Metrics (WAM)



Formative Topical
Development
(2016-2018)

- Advanced Planning Studios at TU Dortmund and University of Duisburg Essen
- *R Studio; Soundscape Ecology*

Rapid Systematic
Literature Review
in Context of
SALVE (2018)

- Ten year search of ScienceDirect and The Journal of the Acoustical Society of America
- Keywords: *Sound, Soundscape, Acoustic, Urban, Environment, Ecology, Metric, Index;*
- Boolean (AND, OR, NOT)
- N= 1,142

Screening,
Review, Final
Selection
(2018-2019)

- Screening for
 - English
 - Metrics or indices with dB, Hz and time
- Literature - N=57; N=17
- *43 Potential Metrics*
- **WAM = 33-34 Alpha, Beta, dB Metrics**

WAM Overview of Findings -Alpha



Table 1 – Alpha and beta sound indices, their origin and calculation package in R

Index	Abb.	Principle	Reference	Calculation Package
Bioacoustic Index	Bio	Biophony Level	19	soundecology
Amplitude Index	M()	Median of Amplitude Envelope	20	Seewave
Temporal Entropy	H _t	Envelope Complexity	21	Seewave
Acoustic Richness	AR	Envelope Complexity and Intensity	20	Seewave
Spectral Entropy	H _f	Spectrum Complexity	21	Seewave
Acoustic Entropy Index	H	Envelope and Spectrum Complexity	21	Seewave
Acoustic Diversity Index	ADI	Spectrum Complexity	22	soundecology
Acoustic Evenness Index	AEI	Envelope and Spectrum Complexity	23	soundecology
Acoustic Complexity Index	ACI	Spectrogram Complexity	24	soundecology
Number of Peaks	NP	Spectrogram Complexity	25	Seewave
Normalized Difference Soundscape Index	NDSI	Ratio of Anthrophony to Biophony	26	soundecology
Shannon's Index	H'	Spectrum Complexity	26	Seewave
Spectral Persistence	-	Duration of Repeated Clusters	17, 18	Kaleidoscope ART 1

WAM - Alpha Scripts in R Studio



```
# Make sure to first install the soundecology and data.table
packages from CRAN repository
```

```
# Load the required packages and libraries
p <- c("parallel", "pracma", "oce", "ineq", "gsw", "seewave",
"soundecology", "data.table", "tuneR", "vegan")
lapply(p, require, character.only = TRUE);
```

```
# Set the working directory and retrieve the wave files
oldwd <- getwd()
setwd("C:/SALVE_SoundAnalysis")
files <- dir(pattern = "wav$");
```

```
# Prepare the function to calculate eight alpha indices
indices <- function(x){
  x <- readWave(x)
  return(c(sh(meanspec(x, plot=FALSE)),
    acoustic_evenness(x)$aei_left,
    ACI(x),
    acoustic_diversity(x)$adi_left,
    bioacoustic_index(x)$left_area,
    H(x),
    nrow(fpeaks(meanspec(x, plot=TRUE))),
    ndsi(x, fft_w = 512, anthro_min = 1000, anthro_max =
2000, bio_min = 2001, bio_max = 11500)$ndsi_left));
```

```
# Prepare the first object where the results will be written in
n <- length(files)
num <- rep(NA, n)
res <- data.frame(Hf=num, AEI=num, ACI=num, ADI=num,
BIO=num, H=num, NP=num, NDSI=num, row.names=files);
```

```
# Run a for loop with the prepared function to calculate the
eight indices for all wav files in the directory
for(i in 1:n) res[i,] <- indices(files[i]);
```

```
# Calculate and store the remaining three alpha indices AR, M
and Ht in a second object
ARI <- AR(getwd(), datatype="files");
```

```
# Join the three alpha indices from the second object to the first
object containing the rest of the indices
res$M <- ARI$M
res$Ht <- ARI$Ht
res$AR <- ARI$AR;
```

```
setDT(res, keep.rownames = "Filename");
```

```
# Export all eleven alpha indices to a csv file
write.table(res,
file="C:/SALVE_SoundAnalysis/DAP_Summ_Alpha.csv", sep=" ",
row.names=FALSE)
```

LiDO3

Hochleistungsrechner

powered by

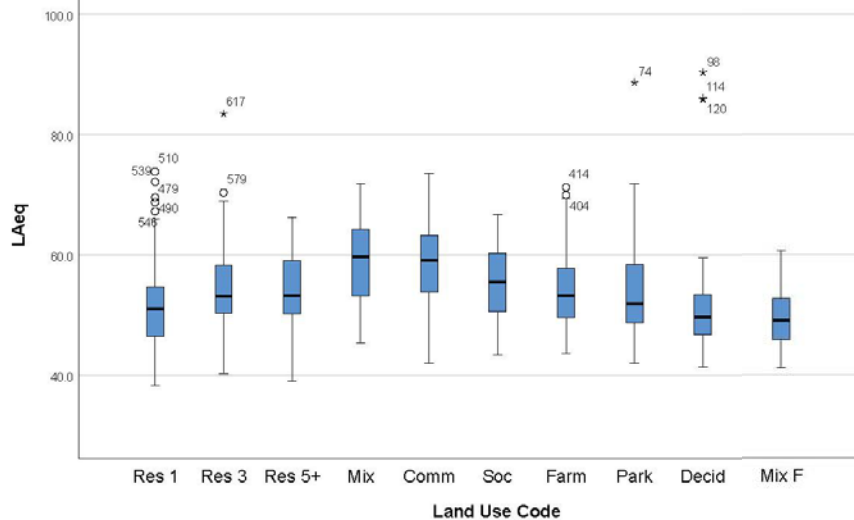
itmc

- Alpha and Beta scripts for DAP observations (n=2 920) can be reasonably calculated with a standard well-equipped computer (4 core, 3.5Hz, 16G Ram) in about 48 hours run time
- Alpha and Beta scripts for AAP observations (n=420 480) must be calculated with the LiDo3 super computer at TU Dortmund (360 quad cores, TBs ram); 250+ days with standard well-equipped computer

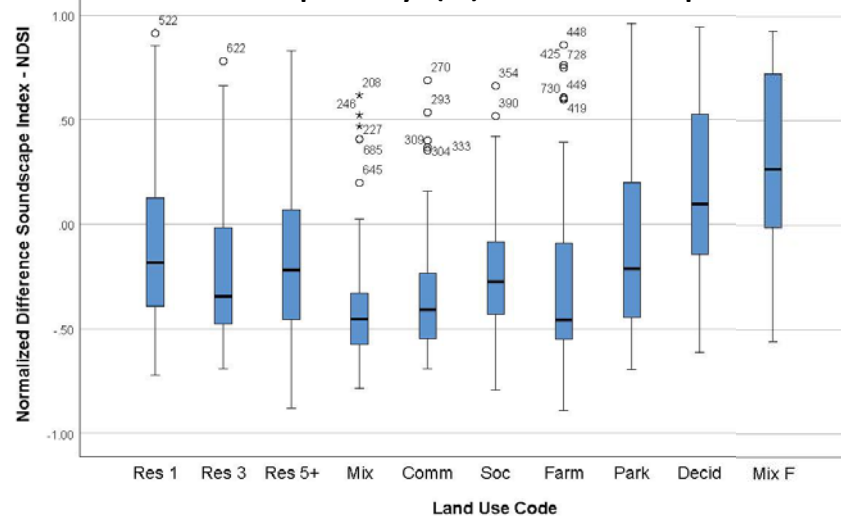
WAM – Spring 2019 Prelim. DAP Results



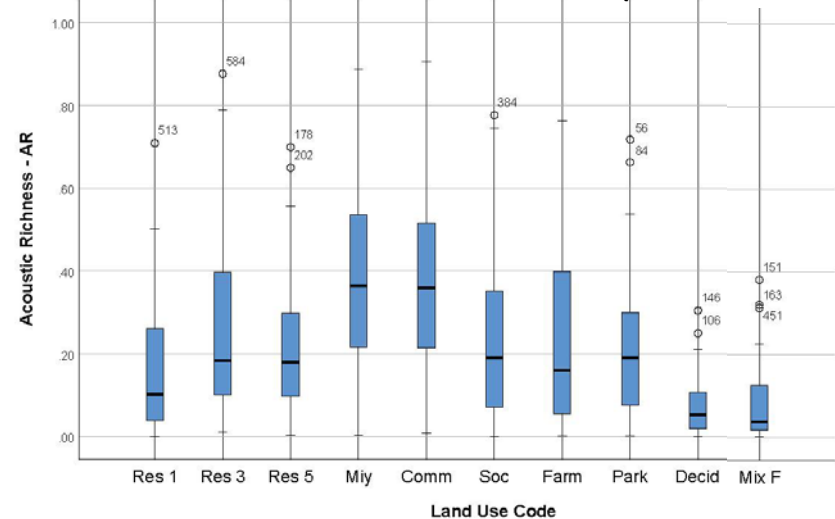
Decibel Level



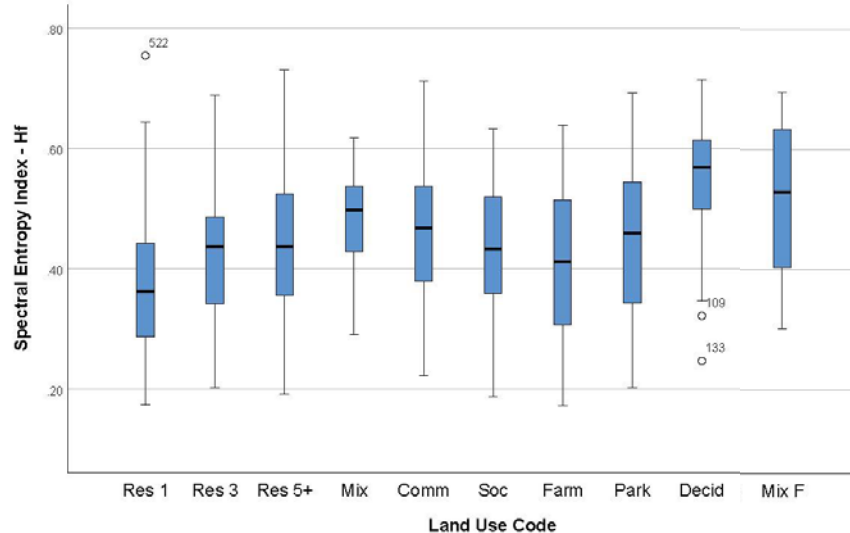
NDSI – Biophony (1) to Anthrophony (-1)



Acoustic Richness – M x Ht (1=loud, even)



Spectral Entropy (Hf) – Hz Evenness (1=even)

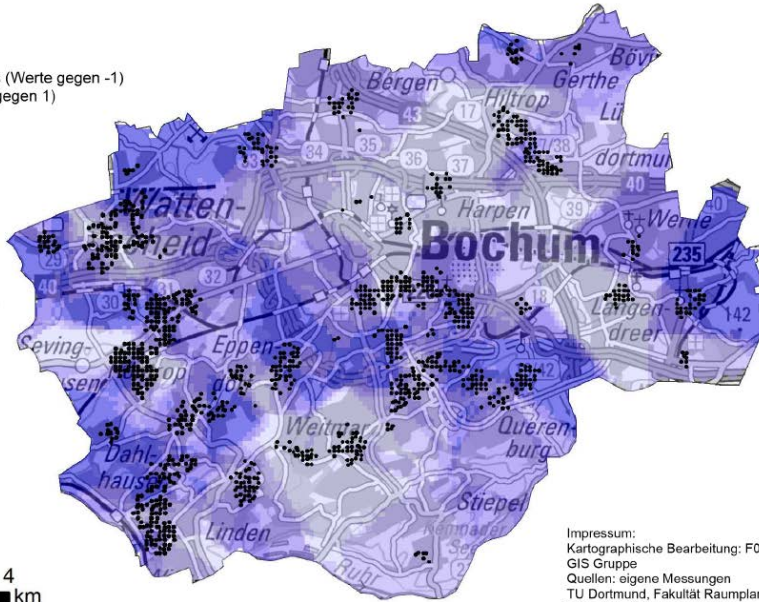
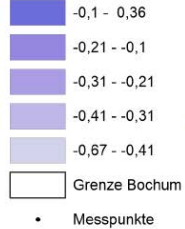


Interpolated ‚Sound Surfaces‘

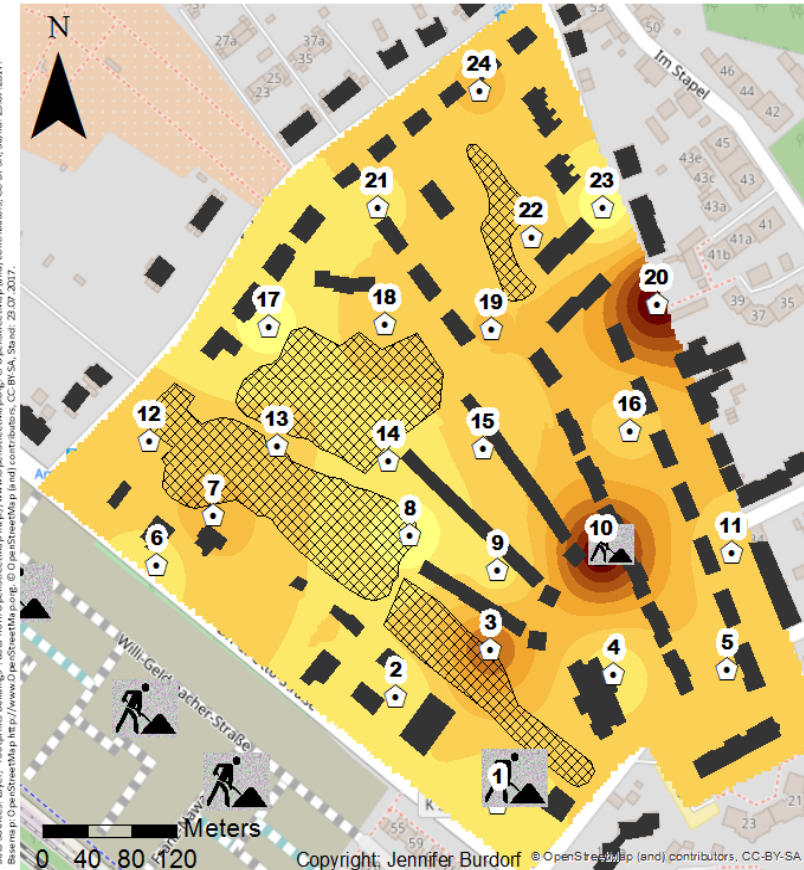
Räumliche Ausbreitung des NDSI

Legende

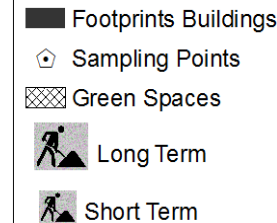
Anteil von Anthroponies (Werte gegen -1) und Biophonies (Werte gegen 1)



Impressum:
Kartographische Bearbeitung: F07,
GIS Gruppe
Quellen: eigene Messungen
TU Dortmund, Fakultät Raumplanung
WiSe 2016/17, SoSe 2017

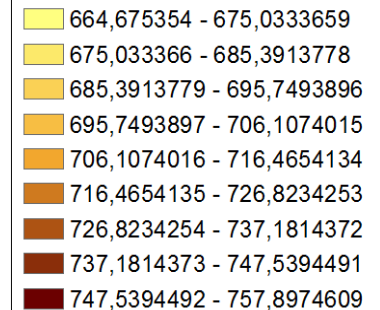


Legend



IDW - ACI

Complexity from low (yellow) to high (brown)



Interpolation

- > Natural Neighbor
- > Inverse Distance Weighting

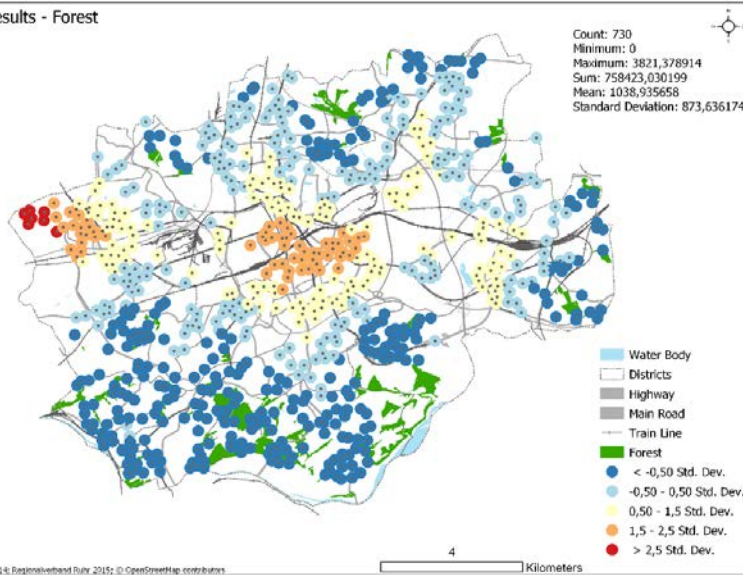
City and Neighborhood Scale

- > Define Sonotopes

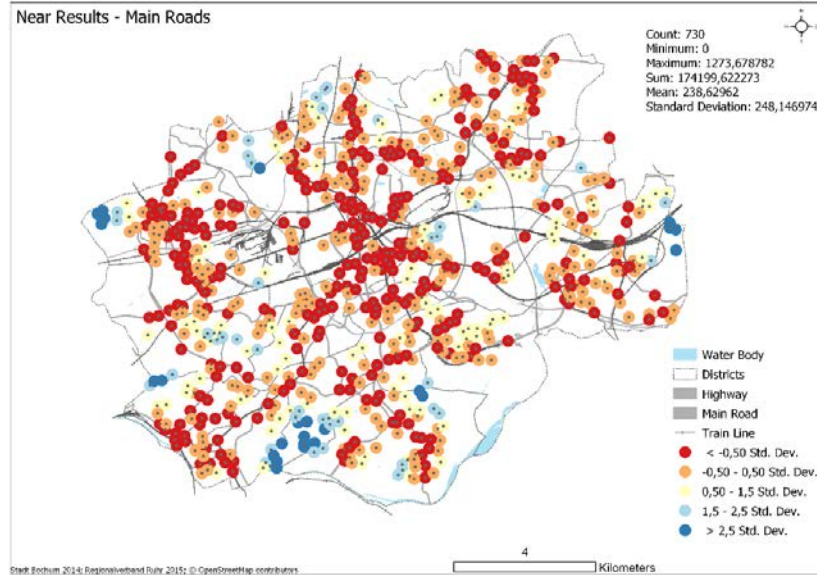
Spatial Data – Near Analysis, Health Data



Near Results - Forest



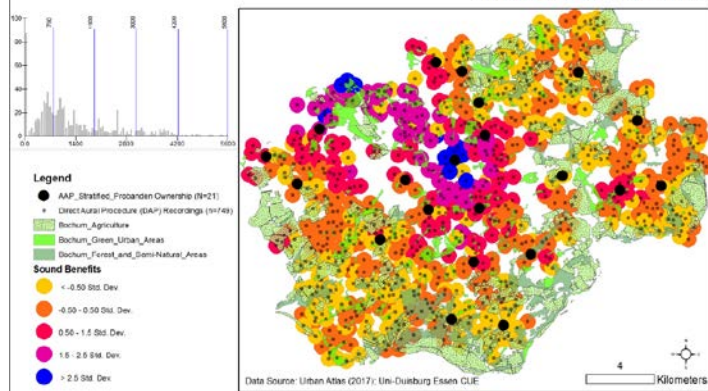
Near Results - Main Roads



Near Analysis > Analysis Tools > Proximity

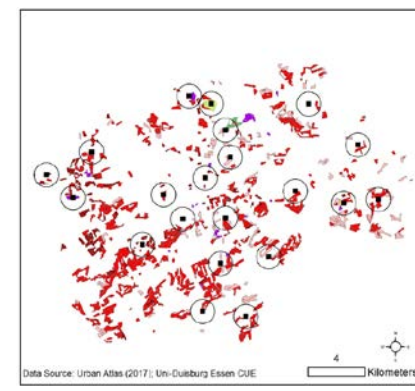
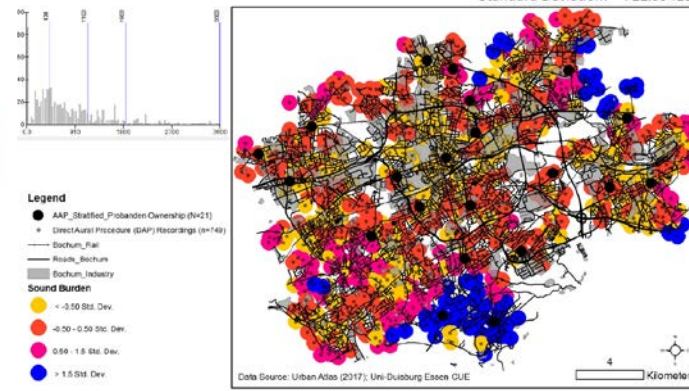
2 Near Results – Sound Benefits

Count: 770
 Minimum: 0
 Maximum: 5555,910156
 Sum: 1027697,402832
 Mean: 1334,671952
 Standard Deviation: 1141,250749



2 Near Results – Sound Burdens

Count: 770
 Minimum: 0
 Maximum: 3620,469971
 Sum: 601513,842136
 Mean: 781,186808
 Standard Deviation: 711,594259



Composite Near Indicators – Social & Environmental Justice

Human Health

- SALVE leveraged several standard spatial analysis methods to develop a defensible sample design strategy for a large set of land uses
- Network analyst helped break down field work into manageable clusters which we designed given personell constraints and is flexible
- Preliminary use indicates WAM may provide a broader understanding of the urban sound environment than decibel focused indices alone
- Spatial datasets including near analysis, NDVI and NDBI will be used to help explain WAM findings in light of spatially distributed health outcomes.

Thank You !!!



For further questions, please contact the team....



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