

Impact of Physical Objects on Scaring of Geese

With an agroecological approach towards the issue of geese as a pest in agriculture

Påverkan av fysiska objekt på skrämsel av gäss - Med ett agroekologiskt grepp på gäss som skadegörare i lantbruket

Elias Kvarnbäck

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Elias Kvarnbäck

Supervisor:	Johan Månsson, SLU, Department of Ecology, Wildlife Ecology Unit	
Assistant supervisor:	Johan Elmberg, Kristianstad University, Department of Environmental Science and Bioscience	
Examiner:	Matias Jonsson, Swedish University of Agricultural Sciences, SLU, Department of Ecology	
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Swedish University of Agricultural Sciences

Faculty of Landscape Architecture, Horticulture and Crop Production Science (LTV) Department of Biosystems and Technology

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Abstract

The interactions and conflicts between geese and agricultural interests have risen in the last decades in Sweden. A range of measures are used by humans to disturb and scare geese with the goal to counteract crop damage. The hypothesis of this master thesis is that proximity to physical objects taller than 70 cm above the ground, e.g., woody perennials, houses or naturally occurring topographical features, makes greylag and barnacle geese easier to scare off crops. The incentive to inquire the effect of physical objects on scaring is that landscape and field features such as hedges, agroforestry and buffer strips are often suggested as agroecological practices. Presence of such element is relevant since geese tend to prefer to forage on fields with good visibility range. The data collected could however not prove that geese are more easily scared/disturbed as they are closer situated to physical objects. Among mixed flocks and greylag goose flocks, proximity to physical objects even made them harder to scare away from agricultural fields.

Keywords: Geese, barnacle goose, greylag goose, large grazing birds, agroecological practices, flight initiation distance

Preface

The broad scope of agroecology presents many opportunities for master thesis topics. Especially if one embraces the food system approach and not just the agroecosystem. I struggled for a while to choose my topic, but a basic yet very powerful sentence that made me finalize my choice, was the following (Smith & Smith 2015 p. 87):

How adaptations enable an organism to function in the prevailing environment – and conversely how those same adaptations limit its ability to function in other environment – is the key to understanding the distribution and abundance, the ultimate objective of the science of ecology.

During my time as a master student, I've sometimes felt that the approach in classical ecology differs a lot from the agroecological approach, where the latter is much more applied and value driven. The interactions between geese on agricultural land and humans, should therefore be seen from an applied perspective, if it is to be considered agroecology.

To write a master thesis, has been quite a different process and feeling than taking other university courses. Without my helpful and cheerful supervisors Johan Månsson and Johan Elmberg, I'd probably have felt lost and bored. The skills, that I've trained and come to value the most while writing my master thesis, are probably statistics in R as well as practical field work. Critical thinking and writing came naturally to me after 1.5 years of courses in agroecology.

Table of contents

1.	Introd	uctio	on	10
	1.1.	Нур	pothesis and aim	11
	1.2.	Stu	ıdy design	12
2.	Backg	roun	nd	13
	2.1.	Ge	ese	13
	2.1.	.1.	Taxonomy and species identification	
	2.1.	.2.	Physiology and ecology	15
	2.1.	.3.	Population and damage trends	16
	2.2.	Leg	gislation	18
	2.3.	Div	versification and agroecological practices	19
3.	Materi	als a	and methods	20
	3.1.	Spe	ecies, scaring sites and limitations	21
	3.2.	Sca	aring procedures	22
	3.3.	Ma	terials	24
	3.3.	.1.	Field materials	24
	3.3.	.2.	Data treatment and software usage	24
	3.3.	.3.	Statistical tests	26
4. Results		s		29
	4.1.	Ge	neral features of the data set	29
	4.1.	.1.	Distance to nearest object	32
	4.1.	.2.	Correlation plots	34
	4.2.	Lin	ear models	
5.	Discus	ssior	٩	40
	5.1.	Dis	cussion of results	40
	5.1.	.1.	Agroecological approach	42
	5.2.	Me	thodological discussion	44
	5.2	.1.	Reproducibility	44

6.	Conclusion		49
	5.2.3.	Data treatment and potential improvement	47
	5.2.2.	Validity due to practical issues	45

Abbreviations

CAB	County Administrative Board/Länsstyrelsen
EU	European Union
FID	Flight Initiation Distance
GIS	Geographic Information System
GPS	Global Positioning System
LGB	Large Grazing Birds, i.e., (in Sweden) swans, geese and cranes
NVV	Naturvårdsverket/Swedish Environmental Protection Agency
SLU	Sveriges Lantbruksuniversitet/Swedish University of Agricultural Sciences
T-goose	
/geese	Target Goose/Geese. In this thesis referring to greylag goose and barnacle goose
WP	Waypoint

1. Introduction

The interactions between wild geese and humans have increased in the last 50-70 years. This is true for both Europe (Fox & Madsen 2017; Fox et al. 2017) and Sweden (Montràs-Janer 2021). The way we farm affect the numbers of geese, and farming practices and productivity are reciprocally affected by the numbers and distribution of geese. There are several goose species in the world; in Sweden nine of them occur naturally and annually (Artdatabanken 2021b). This thesis targets two of them: greylag goose (*Anser anser*) and barnacle goose (*Branta leucopsis*). Hereafter collectively called *target geese* (*T-geese*). Also, if not specified differently, the words "goose" and "geese" refer to wild individuals/populations and not domesticated ones.

Together with common cranes (*Grus grus*) – T-geese are the bird species that render the highest number of damage reports from Swedish farmers to county administrative boards (CABs) (Montràs-Janer 2021). Between the year 2000 and 2015, barnacle geese in single species flocks or as a part of mixed flocks prompted 804 damage reports (ibid.). Greylag geese in single species flocks or as a part of mixed flocks prompted 772 damage reports (ibid.).

In contrast to many other farmland bird populations, most goose populations that are part of European flyways have benefitted from the agricultural intensification and rationalisation in the last 50-70 years (Fox & Madsen 2017). Agroecological practices, on the other hand, usually aim at diversifying the agroecosystem itself and the landscape where it's imbedded (Wezel et al. 2014; Gliessman 2015). Agroecological practices incorporated directly on cropland include intercropping and agroforestry, but also to actively manage landscape features surrounding cropland, e.g., hedges, buffer strips and field islets. A pronounced goal of integrating or re-integrating such landscape features is to provide habitats for natural enemies to pests inside cropland (Wezel et al. 2014). Compared to an agricultural landscape of monocultures with vast fields, the mentioned agroecological practices typically lead to increased patchiness and occurrence of physical objects on and around cropland (Gliessman 2015; Snapp & Pound 2017). As geese prefer to forage on arable land with extended visibility range (Fox et al. 2017; Rosin et al 2012), it's relevant to study how they react to scaring/disturbance depending on how closely they are situated to physical objects that reduce their visibility range. Qualitative features of physical objects that reduce visibility range can be assumed to have an impact on both habitat provision for goose predators, and how much the visibility range is actually decreased. In the results of this thesis, no distinction between qualitative aspects of physical objects is however applied,Objects that are more typical of agroecological practices, e.g.,trees or other taller vegetation in buffer zones are treated the same as houses or road embankments. Of course, it would be more agroecologically relevant to only look at physical objects that are also suggested as agroecological practices. However, that would require a different study design where I would be confined by where these objects were situated and not where the geese happen to be. Data collection would take much longer time using that type of approach.

Passive or active scaring of geese from agricultural land is utilized to prevent crop damage, and one way to see how sensitive geese are to scaring, is to measure flight initiation distance (FID). It simply shows at which distance geese (or any birds) initiate flight from a human being or other actively disturbing element. Geese that show high values of FID are more easily scared away from agricultural land and can therefore be assumed to cause less crop damage. Therefore, it's interesting to see if variables, such as distance to physical objects from geese, leads to increasing FID.

1.1. Hypothesis and aim

The underlying research question for this thesis is how the dependent variable FID is affected by the independent variable of distance to surrounding physical objects. Other independent variables, that is flock constellation, flock size (number of LGB individuals) and date of the scaring trial will also be analysed, yet there's just one main hypothesis:

 H_1 =Distance to objects taller than 70 cm above the ground, shows a significant negative association with FID among greylag and barnacle geese, i.e., ρ (rho) should be negative at an alpha level of 0.05.

Corresponding null hypothesis is then:

 H_0 =There's no negative association between distance to objects taller than 70 cm above the ground and FID for greylag and barnacle geese.

The limit is set to 70 cm above the ground, because this is roughly the height at which the eyes of T-geese are positioned when they're standing on the ground with partly or fully stretched necks. Logically, physical objects taller than 70 have the largest relevance for visibility range among geese.

Note that mere a correlation between FID and proximity shouldn't lead to confirmation of H_1 , i.e., the correlation must be negative. The statistical tests should therefore be negatively one-tailed.

In a linear model the alternative hypothesis could also be described as: $\hat{y} = a + b^*x$.

In this equation $> \hat{y} < is$ the FID and thus the dependent/response variable. >a < is the intercept of the y-axis. >-b < is the coefficient describing the negative correlation and >x < is the distance to a physical object surrounding T-geese.

The ultimate aim of the thesis is that the knowledge generated can be utilized by farmers that wish to adopt agroecological practices to combat geese as a pest on cropland.

1.2. Study design

The results gathered and presented in this thesis are based on an *observational prospective* study design with a *deductive* hypothesis.

The study design is termed observational since the subjects (geese) were treated as similar as possible, and with no purpose of treating the subjects differently. Additionally, other independent variables that I measured, e.g., species constellations and distance to objects from the flocks, were out of my control. And since this is an observational study, I can just point at correlations rather than ultimate causation of FID. Although, I tried to avoid scaring the same geese more than two days in a row, I can't be sure that this was actually the case. This is further discussed in the methodological design.

The study design is prospective because I collected the data myself and the data have not been published anywhere else.

The hypothesis formulated is deductive since it's based on already existing hypotheses claiming that geese prefer foraging sites with good visibility compared to reduced visibility.

2. Background

2.1. Geese

2.1.1. Taxonomy and species identification

In daily English and Swedish language, the words *geese* and *gäss* refer to the two genera of *Anser* and *Branta*, see figure 1.

Class	•Aves (birds)
Order	•Anseriformes
Family	•Anatidae
Sub family	•Anserinae
Genera	•Anser •Branta [True] geese"

Figure 1. Taxonomic hierarchy from birds down to true geese, i.e., the genera Anser and Branta. Sometimes the term "True geese" is reserved for the Anser and Branta genera. In this paper, no distinction is however made between geese and "True geese".

As the species concept is debatable within biology, some goose populations are subject to taxonomic debate. In this thesis, I will however just rely on the taxonomic

classification by SLU Artdatabanken (2021b). This implies there are six *Anser* species and three number of *Branta* species annually and naturally occurring in Sweden, see table 1.

Table 1. Goose species annually and naturally occurring in Sweden (SLU Artdatabanken 2021b). Note that there is a taxonomic discussion about species distinction of some populations. Hybrids between species may also occur. Further subspecies could also be distinguished. T-geese in bold letters.

Anser genus	Branta genus
Lesser white-fronted goose - A. erytrhropus	Barnacle goose - B. leucopsis
Taiga bean goose - A. fabalis fabalis	Canada goose - B. canadensis
Tundra bean goose – <i>A. fabalis rossicus</i>	Brent goose - B. bernicla
Greylag goose - A. anser	
Greater white-fronted goose - A. albifrons	
Pink-footed goose - A. brachyrhynchus	

Due to appearance similarities, some goose species can be hard to distinguish visually from each other in field: this is particularly true for bean goose, pink-footed goose and subadult greater white-fronted goose (Svensson 2009). In the case of greylag goose and barnacle goose, it's usually much easier to distinguish them from other goose species based on plumage. Especially when they're observed on the ground and with a field scope, as is the case in my method, see chapter 2.



Figure 2. Greylag goose – Anser anser. (Åsa Berndtsson 2004). https://commons.m.wikimedia.org/wiki/File:Gr%C3%A5g%C3%A5s_Greylag_Goose_(14341925 828).jpg



Figure 3. Barnacle goose – Branta leucopsis. (Tony Hisgett 2013). https://commons.wikimedia.org/wiki/File:Barnacle_Goose_(8677586443).jpg

2.1.2. Physiology and ecology

Geese are obligate herbivores with a simple and short digestive system compared to other herbivorous species. Because of their simple digestive system, they must consume large amounts of plant tissue relatively to their body weight (Fox et al. 2017). Plant tissues with low fibre/roughage content are therefore preferred by geese. Plants with such features are commonly found inside managed agroecosystems, which may lead to intense interactions between agriculture and geese. The green revolution in the 20th century brought crops with higher proportions of protein and starch in the biomass (Fogelfors 2015; Gliessman 2015). This led to increased harvests and harvest indices (harvested biomass/residual biomass of crop) of such crops (Fogelfors 2015). These plant breeding advancements also benefitted foraging efficiency among geese as they already had adaptations to feed on crops with higher proportions of protein and starch than plants in natural ecosystems usually offer (van Eerden 2005; Fox et al. 2017; Fox & Madsen 2017). Many goose populations in Europe have therefore shifted their diet from wild plants to agricultural plants, which is especially true for wintering geese (van Eerden et al. 2005).

Multiple variables affect where geese forage, and one variable that was confirmed in a review article was the size of agricultural fields (Fox et al. 2017). The attractiveness of large fields has been shown to remain also when the same crop has been grown in fields of different size (ibid.) The plausible reason for this is that geese more easily detect and escape from predators when on large fields compared to smaller fields, thanks to the extended visibility range that larger fields provide (Vickery & Gill 1999; Rosin et al. 2012; Fox et al. 2017). The combination of large fields adjacent to roosting sites typically attract geese, and during such conditions, the conflicts between farming interests and geese are accentuated (Fox et al. 2017; Nilsson et al. 2019).

Depending on life history activity, geese can be broadly categorized as

- wintering,
- migrating or
- breeding

T-geese belonging to each of these life history categories forage on agricultural crops in Sweden (SLU Artdatabanken, 2021b). In March, when data were collected for this thesis, geese found in Scania may be wintering, migrating, or breeding as the latter typically starts in March for European populations (Svensson 2009; Carboneras & Kirwan 2020). Diet and response to disturbance also changes depending on life history activity (Carboneras & Kirwan 2020), and that's why it's interesting to analyse date as an independent variable for FID. T-Geese are opportunistic foragers and a larger share of the Swedish breeding population also winter in Sweden and Northern Europe now than 15-30 years ago. (Olsson et al. 2018; Carboneras & Kirwan 2020; SLU Artdatabanken 2021b). This can be attributed to elevated winter temperatures, as well as more winter sown cereals providing food all winter (Fox et al. 2017; Olsson et al. 2018).

2.1.3. Population and damage trends

As with most wild animal populations, it's hard to know exact numbers of individuals. However – based on the criterion of International Union for Conservation of Nature (IUCN) – neither greylag goose nor barnacle goose are red listed in Sweden (SLU Artdatabanken, 2021c; SLU Artdatabanken 2021d). They are both categorized as *Least Concern* (*LC*).

There are six different monitoring systems, with somewhat different methods and objectives, that provide indices about number of T-geese in Sweden (SLU Viltskadecenter 2018). One of these monitoring systems is called *Viltskadestatistik* (eng. game damage statistics). Since game damage statistics focus on geese on cropland, it's the monitoring system that has the highest relevance for this thesis. Figure 4 shows trends for damage caused by T-geese. When Montràs-Janer (2021) compared annual damage between LGB between year 2000 and 2015 in the game damage statistics, she concluded that:

• Barnacle goose caused the second highest number of damage reports (804) and greylag goose caused the third highest number of damage reports (772)

- Barnacle goose caused the second highest number of yield losses (11 531 metric tonnes) and greylag goose caused the third highest yield losses (9 157 metric tonnes)
- Barnacle goose caused the highest number of compensation costs to farmers (1 136 000 euros) and greylag goose caused the third highest number of compensation (738 000 euros).

Socioeconomic reasons may however also explain differences between farmers' willingness to report damages from LGB. Moreover, one and the same farmer may also be differently prone to report damages from year to year (Montràs-Janer 2021). The drivers and factors influencing farmers' willingness to report crop damage are also emphasized as a needed future research perspective by Montràs-Janer (2021).



Figure 4. Reimbursement to farmers due to damage from T-geese. Mixed flocks include reimbursements where T-geese were present with other LGB. The relative damage caused by T-geese in such mixed flocks is however not revealed. The graph is based on game damage statistics between 2004 and 2019. All annual reports can be found at the webpage of SLU Viltskadecenter (2021).

Reimbursement, as presented in figure 4, includes both *compensation* for caused damage as well as *subsidies* for proactive measurements. The most common proactive measures that farmers were subsidized for in 2019 included (Frank et al. 2020):

- Diversionary feeding sites to lure birds away from economically sensitive crops (mainly by spreading of grains during certain time periods).
- Different scaring measures, e.g., liquified petroleum gas canons, flags and mirror scaring devices.

- Accommodation fields where birds may graze undisturbed.

Reimbursement to farmers don't necessarily reflect the abundance of different LGB species in Sweden, since reimbursement is paid differently depending on the species. Nonetheless, there seems to be a general correlation between crop damages and species abundance of T-geese in Sweden (Montràs-Janer 2021). Between 2000 and 2015, there was a positive correlation between population indices of T-geese and i) number of damage reports, ii) yield loss (biomass) and iii) reimbursement to farmers of T-geese in southern Sweden. The yield losses and reimbursement paid per reported damage increased for barnacle geese, but not for greylag geese between 2000 and 2015 (Montràs-Janer 2021). In other words: the number of reports on crop damage caused by greylag geese increased between 2000 and 2015.

2.2. Legislation

In response to declining bird and goose populations in the first half of the 20th century in the EU, The Birds Directive was implemented in 1979 (European Commission 2019). Together with the Habitats Directive, it provides the main framework for nature conservation and protection in the EU (ibid.). Even though the Birds Directive aims to prevent all kind of human-initiated direct disturbance and killing of wild birds within the EU, it also contains a specific article about birds causing damages directly to crops. Exemptions to disturb and kill wild birds may be given if it aims to prevent *serious* damage to crops, livestock, forests, fishery and water resources (The European Parliament and the Council of the European Union 2009/147). The Birds Directive also stipulates what member countries of the EU may decide nationally and what must be negotiated at multilateral level, for example what might be considered as *serious* crop damage.

In the Birds Directive and on Swedish national level as well, many amendments, that regulate how barnacle and greylag goose may be disturbed and hunted, have been implemented. Without making it too detailed it's still fair to say that barnacle geese are more protected from hunting and disturbance than greylag geese are (SFS 2001:724; The European Parliament and the Council of the European Union 2009/147).

In 1995, the Swedish government initiated a system where farmers can apply to the CABs for reimbursement due to damage from animal wildlife (Montràs-Janer 2021). The legislation is not specific for geese, but also includes large predators such as wolves and bears. The rules are outlined in Viltskadeförordningen (SFS 2001:724) and Naturvårdsverkets (NVV) rules about subsidies and reimbursements

(NFS 2008:16). Reported damages are inspected and analysed by inspectors from the CAB, and thereafter reimbursement may get paid to the affected applicant/farmer depending on the culprit species.

2.3. Diversification and agroecological practices

Since agroecology can be referred to as a movement, a practice and a scientific discipline, (Hazard et al. 2016; FAO 2021), it can be hard to define it. To make it more concise, Wezel et al. (2014) described 15 agroecological practices that were repeatedly found in agroecological scientific publications. To qualify as such, the practice had to contribute to at least one of the following aspects i) Efficiency increase ii) Substitution of inputs iii) Redesign of the agroecosystem or/and the surrounding landscape. Most of these practices rely on diversification of the agroecosystem and more complex food-webs compared to industrial monocultures. The redesign practices are exemplified with re-integration and incorporation of natural and semi-natural elements such as hedges and vegetation strips surrounding areas of more intense agricultural production. Agroforestry systems, where trees are more directly incorporated into areas of intense agricultural production are also identified as a redesign practice (Wezel et al. 2014). It's therefore quite obvious that agroecological practices, aiming for redesign, would create a different farming landscape with smaller field size and more physical objects between areas of intense agricultural production and ultimately decreased visibility range for geese, which is something that geese typically avoid when they forage (Rosin et al. 2012; Fox et al. 2017).

3. Materials and methods

The data collection for this thesis was based on a method already applied in the national goose project "From field and farm to flyway" which is run by my two supervisors: Elmberg and Månsson. In addition to the method of scaring trials they had already developed, I collected data for distance to physical objects surrounding LGB flocks during scaring trials.

In total, 164 scaring trials were performed where each scaring trial can be seen as a sample. Due to data collection errors, 13 scaring trials had to be removed, thus 151 valid ones remained. Sites of scaring trials are shown in figure 8.



Figure 8. Each red flag indicates a scaring trial site, 13 of the scaring trials were later removed due to data collection errors.

3.1. Species, scaring sites and limitations

Barnacle and greylag geese were targeted because they are two of the most common and main culprit species for crop damage in Sweden (Montràs-Janer 2021), see section 1.3.3. However, since T-geese often aggregate with other LGB species, the flocks that I scared could also contain other goose species and whooper swans (*Cygnus cygnus*) Further limitations defined in advance included:

- As specified by the hypothesis, scaring trials were only carried out on agricultural land, i.e., recreational parks, golf courses, home gardens, etc. were omitted.
- Scaring trials were not performed inside nature reserves, national parks or other wildlife refuges.
- My own visibility range had to be at least 500 m., this implied that scaring in darkness or foggy weather was excluded.
- If something obvious disturbed the flock during the scaring trial, it was cancelled and not included in the data set. Such unintentional disturbance included other humans, predators, or vehicles.

Through the species gateway Artportalen, run by SLU Artdatabanken, I could see where T-geese had been observed by the public in the last months and in the last years (Figure 9). As I was based in Malmö, I could then confirm that there should be enough T-geese within 70 km from my home during the data collection time frame. Note, however, that the observations through Artportalen in figure 9 only mirror observations, but not standardized absences.

To at least *avoid* scaring the same goose individuals too close to each other in time, I didn't visit the same area two days in a row, which was controlled with the help of my GPS. No guarantee can however be given that I didn't scare the same goose individuals two days in a row since geese naturally cross the borders between the areas I visited. This approach is further discussed in the methodological discussion.



Figure 9 (Used under publisher's permission). Observations by the public of T-geese between 2016.01.01 and 2021.02.06 in Scania. Darkness of squares indicate higher numbers of observations. (SLU Artdatabanken 2021a)

3.2. Scaring procedures

All WPs (waypoints) were recorded with a handheld GPS. A simplification of the method is shown in figure 10. Numbers of scaring trials per day varied between one and eleven (figure 12). During all scaring trials, I wore the same plain-coloured jacket of burgundy. Colours of trousers varied between grey and green. My own height is 191 cm and my weight is 83 kg.

The data collection proceded as follows:

- 1. While I was driving, I looked for T-geese through my car without binoculars or other optical aides. For geographical range of scaring sites, see figure 8 above.
- 2. Where the traffic situation allowed safe parking the car, a scaring trial could be conducted. Free sight between myself and the majority of the individuals in the flock was a prerequisite, too. When the sight requirement couldn't be met, I walked to a spot where there was free sight between me and the majority of the individuals in the flock. The walking direction to such a spot

was never directly towards the geese, i.e., less than 180° towards the flock.. The distance between the car and the flock varied and the only goal was to park the car so that its presence didn't initiate flight of the flock.

- 3. I counted the number of individuals of each LGB species. For flocks of more than roughly 100 individuals, individuals were not counted exactly, but rather in units of five or ten. If there were species identification uncertainties, I used a field scope to get sure.
- 4. Waypoint (WP) 1 was registered as soon as I had exited the car or at the spot where there was free sight between the majority of geese in the flock and myself (see step 2).
- 5. I then walked towards the approximated centre of the flock in typical walking pace in a calm manner. Walking speed wasn't registered but most likely varied between 3 and 6 km/h. I intended to walk as straight as possible towards the flock, but this had to be balanced against not getting too wet myself and if the crops growing in the field was likely to get seriously damaged by my footsteps.
- 6. When the first goose/geese in the flock, initiated flight, I stopped walking and registered WP2 with the GPS. If the flock consisted of different LGB species, it was registered in which order the species initiated flight.
- 7. I next walked to the spot where I assessed that the first goose/geese initiated flight and registered WP3 with the GPS. The accuracy of reaching this spot varied due to field and landscape factors, particularly muddiness and water saturation. This is a validity issue which is brought up in the methodological discussion. At WP3 I also collected data for an additional variable:
- Distance to nearest physical objects surrounding WP3. With help of the compass in the GPS, objects were sought after in the four cardinal directions: i.e., north, east, south, and west. For each cardinal direction, a laser gauge was used to detect and measure distance in meters to the nearest physical object in each cardinal direction. The laser gauge was held horizontally at 70 cm above the ground at WP3. The objects had to be stationary, i.e., not moving. For instance, cars passing by were omitted.
 - 8. After registering the observations at WP3, I walked back to the car. A new scaring/sampling could be done after a minimum of 2 km of driving, alternatively if the T-goose individuals were believed to not be the same ones that had just been scared in the previous scaring trial. This was determined by simply observing the flight direction of the previously scared flock.



Figure 10. A simplified illustration of the method. The red-marked bird shows that this is the first individual that initiated flight, and thus where WP3 was marked.

3.3. Materials

3.3.1. Field materials

To accomplish the method described above, the following materials were used:

- Car (Toyota Corolla, estate car specified since appearance and sound of the car might have an impact on the scared flocks)
- Field binoculars (Nikon ProStaff 7S, 8x42)
- Field scope on tripod (Lotus SP80)
- Laser distance gauge (Leica RangeMaster CRF 800)
- Handgeld GPS (Garmin ETrex 32X). The device utilized both GLONASS and GPS satellites.
- Wooden stick of 70 cm height to control for height at WP3.

3.3.2. Data treatment and software usage

Coordinates/WPs were transferred to the BaseCamp software issued by Garmin, using the Topoactive Europe 2020.20, North East map. Coordinates were also converted from WGS84 to RT90 coordinates in ArcGis. The Pythagorean theorem, using X and Y-coordinates of the RT90 system, was then utilized to obtain the FID values.

Since distance to objects was measured in four cardinal directions, distance to object/s can be displayed in two ways:

- 1. Distance in meters to the nearest object in one of the cardinal directions.
- 2. Average distance in meters to objects, i.e., sum of distances to objects in each cardinal direction divided by the number of measurements. Since objects couldn't always be measured in each cardinal direction, the number of measurements differed.

All data were then compiled in an Excel CSV spreadsheet which was subsequently imported into RStudio developed by the R Core Team (2020). The following packages (authors/developers in brackets) were also used for data analysis and creating the plots of this thesis.

- Tidyverse (Wickham et al. 2019)
- Corrplot (Wei & Simko 2017)
- Lubridate (Grolemund & Wickham 2011)

In total, 164 scaring trials were performed, but 13 were removed from the analysis due to data collection errors. This rendered 151 scaring trials included in the analysis. Flock constellations could have been defined in different ways, read further in the methodological discussion, but the chosen definitions of flock constellations were:

- 1. Flocks containing greylag geese but no barnacle geese.
- 2. Flocks containing barnacle geese but no greylag geese.
- 3. Flocks containing both greylag and barnacle geese, i.e., mixed.

It's the distinction above that is referred to when the simplified terms "greylag goose flocks" or "barnacle goose flocks" is mentioned in the continuation of this thesis. Keep in mind that these definitions of constellations imply that LGB species other than T-geese were sometimes also present in the flocks. Further on, this definition of flocks, doesn't take flight initiation order between species into account.



Figure 11. Species constellations of scared T-geese flocks. Barnacle goose flocks=10, greylag goose flocks=122, mixed flocks of barnacle and greylag goose=19

Scaring trials started on 2021-02-12 and finished on 2021-03-23. The number of scaring trials was relatively evenly distributed throughout the study period (Figure 12) Note that there are different time gaps between dates when scaring trials were performed. Dates of the scaring trials were transformed to of ordinal dates of 2021 with 2021.01.01 as the start value (1) in correlation tests and linear models.



Figure 12. Number of scaring trials per day.

3.3.3. Statistical tests

By looking at figure 13, one can conclude that FID was not normally distributed. However, FID-values became normally distributed by logarithmic transformation with the natural logarithm ($e\approx 2.72$), which is shown in figure 14. Also, in all other graphs and statistical tests, it's the natural logarithm that is used when logarithmic transformation is mentioned.

Constellation-wise, Shapiro-Wilk tests showed that FID was probably (p=0.92) only normally distributed in flocks containing barnacle geese but no greylag geese. But due to the small sample size of this constellation n=10, the probability of normality shouldn't be relied on. The overall unnormal distribution of FID among flocks, led to the decision to run the non-parametric Spearman's correlation coefficient test, since it doesn't require normally distributed observations.



Figure 13. Distribution of FID for all scaring trials. FID-values are clearly skewed.



Figure 14. Logarithmically transformed FID-values. Shapiro-Wilk normality test yielded a p-value of 0.432, which indicates a high probability of normal distribution in the population.

Since the alternative hypothesis required FID to correlate negatively with distance to object, the correlation tests for *distance to object* was decided in advance to be negatively one-tailed. Ties that appeared in Spearman's correlation tests rendered inexact p-values, which were treated with asymptotic t-approximation.

Also when grouped by constellation, FID-values became normally distributed by logarithmic transformation in each constellation. Consequently, the linear model in figure 28 and table 3 to 5 showing multiple regression of date, flock size and distance to nearest object, is based on logarithmically transformed FID-values. Untransformed FID-values should not be considered statistically valid in the linear modelling, but for comparative purposes these are shown in appendix 3.

Besides the variables presented in the results, data for several other variables were also collected during the scaring trials, but these were omitted in results and in the linear modelling.. The entire data set, which also includes variables that weren't analysed, is found in appendix 1.

4. Results

The results are mainly displayed through tables and diagrams with supplementary descriptions. The first section shows general aspects of the data set and FID features related to the independent variables one by one. The second section describes the results through linear modelling taking multiple independent variables into account. FID is always expressed in meters.

4.1. General features of the data set

Flight initiation order between species is shown in figure 15. Of all scaring trials, 118 flocks were single-species flocks, whereas the remaining 33 flocks were mixed, and it's the latter 33 that are shown in figure 15. Since FID was measured as the first LGB individual/s/ initiating flight, this created multiple flight initiation orders.



Figure 15. Flight initiation order by species in flocks containing more than one LGB species. "Unknown" refers to flocks where it was not possible to determine the flight initiation order. In some cases it was not possible to tell the order of a specific species, see brackets.

Most flocks were comparitevely small in numbers of LGB individuals (see figure 16a and 16b). Only 15 out of 151 flocks contained more than 100 LGB individuals.



Figure 16a. Histogram of all flocks. Bar width is 200 individuals.



Figure 16b. Histogram of flocks containing 0-200 individuals, i.e. a blow-up of data in the leftmost bar in figure 16a.

The FID median for all flocks was 95 meters. It was higher for greylag goose flocks compared to barnacle goose flocks (figure 17), (94m and 56m respectively). The FID median for mixed flocks was 117 meters, which is closer to greylag goose flocks than barnacle (figure 17). Mean FID for all scaring trials was 105 meters. Mean FID for greylag goose flocks was 106 meters. Mean FID for barnacle goose flocks was 68 meters. Mean FID for mixed flocks was 116 meters. Non-parametric

Wilcoxon rank sum test with continuity revealed a significant difference between barnacle goose flocks and mixed flocks: p=0.0087.



Figure 17. FID (median) by species constellation. Whiskers depict the median $\pm 1.5^*$ the interquartile range. Values outside whiskers are considered as outliers and depicted as dots.

Ordinal dates in correlation to FID turned out to show a negative rho of -0.2614 and to be significant (figure 18). The correlation of ordinal dates may however just be generalized to the actual data collection time frame, i.e., between 2021.02.12 and 2021.03.23. Flock size, expressed as number of individuals in the flock, returned a rho of 0.1364, but this was not significant (figure 19).



Figure 18. Scatterplot showing association between ordinal date and FID for all scaring trials. Two-tailed Spearman correlation test provided a rho of -0,2614 with p-value: 0.0012.



Figure 19. Scatterplot showing association between the number of goose individuals in a flock and FID for all scaring trials. Two-tailed Spearman correlation test provided a rho of 0.1364 with p-value: 0.0948.

4.1.1. Distance to nearest object

In direct contrast to the hypothesis, there was a slightly positive correlation between distance to nearest object and FID. When all scaring trials were included, rho was estimated at 0.1588 with a p-value of: 0.97. Also, when looking at greylag goose flocks and mixed flocks separately, rho was positive. But even constellation-wise, very high p-values were obtained, and these are shown in the captions to figure 20-23. The high p-values can be attributed to the correlation-tests being negatively one-tailed when looking at *distance to nearest object*. If the correlation-tests instead would have been two-tailed, the p-values would have been much lower. The motivation behind running negative one-tailed tests can be attributed to the hypothesis assuming a negative correlation.



Figure 20. Scatterplot showing distance to nearest object and FID for all scaring trials. Onetailed negative Spearman correlation test provided a rho of 0.1588 and a p-value of 0.97.



Figure 21. Flocks containing greylag geese but no barnacle geese. Other LGB species may be present in the flock. One-tailed negative Spearman test provided a Rho of 0.1722 and a p-value of 0.97.



Figure 22. Flocks containing barnacle geese but no greylag geese. Other LGB species may also be present in the flock. One-tailed negative Spearman test provided a Rho of -0.0502 and a p-value of 0.45. Note the small sample size, n=9, for observations that contained pairwise observations of both variables.



Figure 23. Flocks containing both greylag geese and barnacle geese, i.e., mixed. Other LGB species may also be present in the flock. One-tailed negative Spearman test provided a Rho of 0.4611 and a p-value of 0.958.

4.1.2. Correlation plots

The correlation matrix plots shown in figure 24 to 27 offer the possibility to compare several variables simultaneously, and it's also a way to detect multicollinearity between independent variables. In the correlation matrix plots, it's shown that *Average distance to nearest object* shows comparatively high correlation with *Distance to nearest object*. This is not surprising since *Average distance to object* takes objects in all cardinal directions into account and that one of them per se is *Distance to nearest object*, see section 2.3.2.

Size of the circles in figure 24 to 27, indicates, just as colour, the strength of the correlation.



Figure 24. Correlation between variables for all scaring trials. The black background is just to distinguish it from the constellation specific correlation plots below. The matrix is based on two-tailed Spearman correlation coefficient tests.



Figure 25. Correlation between variables in greylag goose flocks. The matrix is based on twotailed Spearman correlation coefficient tests.



Figure 26. Correlation between variables in barnacle goose flocks. Keep the small sample size in mind, n=9. The matrix is based on two-tailed Spearman correlation coefficient tests.


Figure 27. Correlation between variables in mixed flocks. The matrix is based on two-tailed Spearman correlation coefficient tests.

4.2. Linear models

Based on figures 20 to 23, the hypothesis that FID would show a negative correlation to nearest object, can be refuted. Even though the correlations for greylag goose and mixed flocks were positive, it's still interesting to inquire the coefficient *distance to nearest object* in a multiple regression with other variables, namely: ordinal date; flock size and flock constellation. P-values in the multiple regression are based on two-sided testing in contrast to the one-sided testing related to figure 20 to 23.

Stars next to p-values indicate levels of significance: <0.001=***, <0.01=**, <0.05=*. The coefficients that turned out to be significant were the intercepts for greylag goose flocks (table 3) and mixed T-goose (table 5) flocks and finally *Distance to nearest object* in greylag goose flocks (table 3).

Table 3. Coefficients of the multiple linear regression describing greylag goose flocks. FID-values have been logarithmically transformed in order to obtain normally distributed FID-values, see section 3.3.3.

Coefficient	Estimate	Std. error	P-value
Intercept	4.97	0.365	2e-16***
Distance to nearest	0.002	0.0008	0.012*
object			
Ordinal date	-0.008	0.004	0.08
Flock size (number	0.004	0.002	0.135
of individuals)			

Table 4. Coefficients of the multiple linear regression describing barnacle goose flocks. FID-values have been logarithmically transformed in order to obtain normally distributed FID-values, see section 3.3.3.

Coefficient	Estimate	Std. error	P-value
Intercept	4.006	2.348	0.149
Distance to nearest	-0.0007	0.003	0.831
object			
Ordinal date	0.0002	0.286	0.996
Flock size (number	0.0004	0.0004	0.446
of individuals)			

Table 5. Coefficients of the multiple linear regression describing mixed flocks. FID-values have been logarithmically transformed in order to obtain normally distributed FID-values, see section 3.3.3.

Coefficient	Estimate	Std. error	P-value
Intercept	4.536	0.664	2.82e-05***
Distance to nearest	0.003	0.001	0.052
object			
Ordinal date	0.0001	0.009	0.991
Flock size (number	-0.0001	0.0003	0.74
of individuals)			

The relative importance of *Distance to nearest object* can also be shown in linear multiple regression curves where all other variables are kept equal (figure 28), in this case at mean values for each variable in each constellation. Note that all variables from the tables above are included, also those that didn't show any significance. As expected from the coefficients in table four, barnacle goose flocks show a negative correlation with FID, but this result should not be relied on since the p-value was >0.05.



Figure 28. Linear model for all constellations using logarithmically transformed FID-values and the range of values for Distance to nearest object. Mean values for variables other than "Distance to nearest object" are used. The regression lines are expressed below.

Barnacle: y(log(FID))=-0.0007036*Distance to nearest object + 0.0003696*Mean number of individuals + -0.0001520*Mean ordinal date + 4.0062753 (Intercept)

Greylag: y(log(FID))=0.0020880*Distance to nearest object*+ 0.0040289*Mean number of individuals + -0.0055520*Mean ordinal date+ 4.9709891 (Intercept)

Mixed: y(log(FID))= 0.0033715*Distance to nearest object + -0,0001154 *Mean number of individuals + 0.0001126*Mean ordinal date + 4.5359580 (Intercept)

5. Discussion

Since the hypothesis was clearly refuted, a fairly large part of the discussion is devoted to reproducibility and validity issues.

5.1. Discussion of results

The hypothesis was that proximity to physical objects would show a negative association with FID-values among T-geese. On the contrary, there was a positive correlation between distance to nearest object and FID for greylag goose flocks and mixed flocks. Among barnacle geese, there was still a negative correlation between distance to nearest object and FID. But the small sample size of barnacle goose flocks, couldn't bring any significance to such correlation, see figure 22. In the multiple linear regression model of logarithmically transformed FID-values (figure 28), where several variables were considered, the relative effect of *Distance to nearest object* on FID decreased. But as none of the other variables in the multiple regression were significant, no conclusion of the interplay between *Distance to nearest object* and other variables can be made.

Field size and visibility range correlate positively with forage selection among geese in several studies. The general trend is that geese prefer agricultural fields of larger size over smaller fields (Rosin et al. 2012; Fox et al. 2017). The guidelines from NVV and SLU Viltskadecenter, bring up that incorporation of physical objects such as cover hedges and unharvested stalks of crops can be utilized as a strategy to avoid damage on cropland from LGB (Månsson et al. 2015). The results from this thesis don't support that geese are more easily scared in such habitats. But even though the results showed that scaring isn't facilitated by proximity to physical objects, it could still be the case that T-geese cause less damage in fields with such features as they may still prefer to land on fields far from these kind of physical objects. The method applied in this thesis has only inquired where geese were already present, i.e., not inquired which fields that were avoided by T-geese. No efforts were made to inquire whether T-geese significantly avoid fields as a function of distance to physical objects from the very beginning.

Even though it doesn't answer the hypothesis, it was interesting to see that barnacle goose flocks showed a significantly lower FID than mixed flocks (figure 17). This is logical if one considers that barnacle geese are far more protected from hunting than greylag geese in the EU and Sweden (The European Parliament and the Council of the European Union 2009/147). It's therefore anticipated that barnacle show less sensitivity to humans than do greylag geese. Even though the difference between barnacle goose flocks and mixed flocks turned out to be significant, one should keep the small sample size of barnacle goose flocks in mind (n=10). If it holds true that mixed flocks of T-geese and other LGB are more easily scared compared to single species flocks, this can have obvious scaring management implications. If farmers start preferring mixed flocks rather than single species flocks on their fields, due to the facilitated scaring potential, it could also mean that the task for inspectors working at CAB becomes harder. This is since reimbursement from CAB is paid differently by species and a part of the CAB inspector's job is to determine which species that has caused the damage (Månsson et al. 2018; SFS 2001:724). Montràs-Janer (2021) also points out that it's desirable that culprit species are always specified in the damage reports. Then it'd be easier to predict in which areas that conflicts between farming interests and T-geese may arise (ibid.). However, FID may also vary between regions, for example did similar scaring trials around Kristianstad show that barnacle goose flocks didn't turn out to exhibit significantly lower FID-values than other constellations (appendix 2).

A free reflection during scaring trials of barnacle goose flocks, was that these flocks were more restless than greylag flocks/individuals. This wasn't measured, but during scaring trials of barnacle goose flocks, the scared individuals often just started to circulate around me after flight initiation. This contrasts with greylag geese which I found much more determined in their flight direction after flight initiation. No general conclusion can however be based on this unsystematic observation. But given the problem of moving around geese between crop fields causing conflicts between farmers (Månsson et al. 2015; Månsson et al. 2018), it stresses the needs to understand what they do and where they head after being scared. Related to my own research questions, it would be interesting to see how far T-geese fly after scaring as a function of distance to physical objects and patchiness of the landscape.

Habituation to scaring devices among LGB, is one of the main issues in scaring strategies and management of such populations (Månsson et al. 2015; Fox et al. 2017). Habituation can be described as an animal learning to not respond to a stimulus (Raven et al. 2011). Even if the animal initially has genetically adapted instincts to respond to a specific stimulus, the response may decrease if the stimulus repeatedly doesn't affect fitness positively or negatively (Raven et al. 2011). Thus, when stimuli, e.g., scarecrows, or physical objects such as trees or topography, are

exposed to a goose, the initial response of the goose might be strong, but as they get repeatedly exposed to it, the weaker the response becomes, which is the process of habituation. This could be elaborated much further, but it emphasizes the need that visual stimuli such as physical objects, eventually must imply fitness-reducing/lethal attacks if geese are not to habituate to physical objects. A potential explanation to why the hypothesis was refuted might be that T-geese in the study area have encountered too few attacks from predators such as foxes that depend on physical objects and patchiness of the landscape. Thereby visibility range isn't as important as it is for other goose populations that have been more frequently exposed to physical objects as an agent of natural selection.

5.1.1. Agroecological approach

So far, the discussion might seem a bit reluctant to acknowledge that proximity to objects couldn't be associated with higher FID-values. This might stem from agroecology's encouraging view of patchiness and more physical objects in the agricultural landscape (Thies & Tscharntke 1999; Wezel et al. 2014). The belief in agroecology is that such landscape features increase the interactions and food webs between biotopes of intense agricultural productivity, biotopes of reduced human disturbance, and natural ecosystems. Eventually such an approach is believed to contribute to ecological intensification that depends less on external inputs and fossil fuels (Tittonell 2014; Wezel et al. 2014; Gliessman 2015).

Gliessman (2015) suggests five steps of conversion to agroecological food systems. And as the name "food system" implies, this includes looking beyond agroecosystems themselves, and to also consider social and economic issues. The first three levels of agroecological conversion consider sustainability within agroecosystems (Gliessman 2015). The hypothesis of this thesis is typically an issue that's regarded within conversion level one to three. Level four, on the other hand, moves beyond farm level, as its goal is to: "Re-establish a more direct connection between those who grow the food and those who consume it." Gliessman (2015, p. 348). Although the data collected for this thesis didn't aim to reveal any possibilities of reaching level four, it's still interesting to consider goose management on level four. One way to do so, is to look at how the public (consumers in Gliessman's definition of level four) view geese. This has been done by Eriksson et al. (2020). By random sampling, a survey was e-mailed to adult citizens living in the goose rich municipalities of Örebro and Kristianstad. The aim was to reveal attitudes towards geese and to find predictors of acceptance towards geese. One of the results was that 71 % of the respondents in Kristianstad and 60 % of the respondents in Örebro were positive to have geese in Sweden (Eriksson et al. 2020). However, 36 % of the respondents in Kristianstad and 48 % of the respondents in Örebro thought that the numbers of geese were too high in their home municipality (ibid.). Eriksson

et al. (2020) wonder if this may be a result of geese being part of the culinary traditions of Scania, and that people are more used to geese in Kristianstad than in Örebro. In areas where geese are more profoundly considered a local pest and a problem to the public on beaches, parks, golf courses etc., farmers could connect with consumers by telling them how they work to hamper damage and disturbance to humans from geese on a landscape level. But as Gliessman points out, this requires short links between producers and consumers (2015). In areas where geese are considered more problematic by the public, consumers should be more willing to pay attention to farming practices that aim to reduce goose populations on a regional scale. As Eriksson et al. (2020) show, acceptance towards geese differ depending on geographical area. Hence, the communication and connection between farmers and consumers could be differently successful depending on the region and the magnitude of how the public perceive geese as a nuisance.

Finally - even if an agroecological approach of patchiness and ecological intensification of agroecosystems, would be successful to combat geese as a pest, there would still be challenges. Notably because the general prediction by the European Environment Agency (EEA) (2019) is that agricultural productivity in southern Europe will be worse affected by climate change than in northern Europe including Sweden. EEA believes that this might increase the relative importance of northern European countries, e.g., Sweden, for the food supply of entire Europe. In that perspective, there will likely be stronger incentives to use more external inputs and relying less on agricultural diversification to combat geese as a pest in Sweden. Wezel et al. (2014) also disclose a pattern of agroecological practices being less disseminated in naturally fertile agricultural areas with high productivity. If Sweden's relative importance for Europe's food security increases, it could lead to short term productivity intensification of Swedish agriculture. In such case, agroecological practices are less likely to be favoured in Sweden. Instead, it's more likely that more direct methods such as direct killing of geese and scaring measures based on intensive supply of external inputs, e.g., liquified petroleum gas canons or drones, are favoured. Such rather direct measures would not demand the redesign of conventional agroecosystems that agroecological practices imply. Direct killing or scaring through external inputs may not be environmentally harmful in themselves, but with agroecological practices many other sustainability benefits are enhanced too. Not at least since habitat destruction has been the main driver of biodiversity loss in the last century (IPBES, 2019). For farmland birds, the trends have been particularly adverse since 1980 (Pan-European Common Bird Monitoring Scheme, 2021), see figure 29. Negative side effects on farmland bird species/populations from increased hunting or use of external inputs such as liquified petroleum gas canons to regulate goose populations, must be carefully evaluated. Hopefully, the Birds Directive should probably serve to not let such side effects become too adverse on farmland birds other than geese.



Figure 29 (Used under publisher's permission). Species abundance indicator for birds grouped by habitat. (Pan-European Common Bird Monitoring Scheme 2021).

5.2. Methodological discussion

5.2.1. Reproducibility

Since the method used for this thesis utilized an observational design with uncontrolled variables, it's not possible to repeat the sampling procedure and obtain the same results. Mainly because it cannot be determined exactly where in the landscape T-geese will be present, i.e., which agricultural/crop fields they would be in and how closely they would be situated to physical objects.

However, some methodological facets could easily be repeated in other scaring trials:

- The area where scaring trials took place. Even though it's not possible to use exactly the same scaring sites, one could set up rules to carry out scaring trials in a certain vicinity of scaring trial sites. All coordinates of scaring sites are found in appendix 4.

- The seasonal and diurnal time frame when scaring trials were performed.
- The appearance and walking pace of the person scaring the flocks.
- To only conduct scaring trials when the visibility range is minimum 500 meters for the human naked eye.
- To only conduct scaring trials on agricultural land.

5.2.2. Validity due to practical issues

Practical validity issues of obtaining correct FID-values concern whether actual methods and materials could measure FID and distance to object in a precise way. During the data collection phase, I thought about many different validity issues. Below are the ones that I personally found most striking:

- The distance between the car/WP1 and the flock differed quite a lot. The longer distance from the flock I had to park the car, the longer time it usually required to count the geese. The time I was visible to individuals in the flock before I approached the flock might have had an impact on FID. But this was not measured.
- Due to muddiness and water saturation in some fields it was hard to walk straight between WP1 and WP3 (figure 30). The longer detour I had to make to reach WP3, the less precise could I be in reaching the *de facto* spot of flight initiation. Fields that were easier to walk on should therefore show better precision for reaching WP3 and eventually FID.
- Additionally: In fields that were extremely muddy (figure 30), walking pace towards the geese was slowed down. Walking speed per scaring trial was not measured.
- At WP3 the slope of the ground varied. This made it sometimes hard to horizontally stabilize the 70 cm stick that the laser gauge was put on. Scaring trials at sites with less slope, probably show better precision for measuring distance to objects.
- When the conditions were wet and muddy at WP3, the 70 cm stick tended to slightly penetrate the ground during measurement of distance to objects. This probably yielded shorter distances to objects as the measuring height to objects was lower than 70 cm. Ultimately measurements in wet and muddy fields yielded lower values of distance to objects than was the case.
- I couldn't be sure how many times I scared the same goose individuals. Although I had a method for avoiding scaring the same individuals the same day or two days in a row (see section 2.1 and 2.2), this might still have occurred. A result of this could be habituation among the scared T-

geese towards the scaring trials themselves. A way to test for possible habituation and the risk of scaring the same goose individuals the same day, could have been to test if there were significant differences in FID depending on the time of the day. No such tests were however run. The comparatively large geographical range of scaring sites (figure 8) should at least serve to diminish the risk of scaring the same goose individuals too frequently.



Figure 30. This is what the boots looked like just after some of the scaring trials. The muddiness of the field and the weight of the boots had an obvious impact on walking speed towards the flocks.

To summarize, some validity issues regard the accuracy of FID-values, whereas some regard accuracy of distance to physical objects from WP3. Better precision of FID-values could be obtained by harnessing goose individuals equipped with GPS-collars. Better precision of measuring distance to objects from WP3 and general landscape features could be solved with GIS or other mapping techniques. Of course, it would also have been possible to measure distance to objects in other directions than the four cardinal directions. But to measure which was actually *the closest* object in 360° around WP3 would have been very time-consuming given the materials at hand. The number of scaring trials performed, i.e., sample size, would have suffered dramatically.

Finally, it's worth to reflect on whether FID is a good indicator of how easily Tgeese are to scare and prevent from causing damage on crops. Many other options are theoretically possible to measure how T-geese react to scaring or disturbance. For instance, how they fly away after a conducted scaring trial, or how and in how big groups they regroup after a scaring trial. If a scared flock is split into many smaller subgroups, the damage on crops is probably smaller and not as locally adverse. Not surprisingly, farmers are more inclined to report crop damage to CABs when they encounter larger rather than smaller flocks on their cropland. (Montras-Janer 2021).

5.2.3. Data treatment and potential improvement

Division and definition of flock constellations could have been done differently. Flocks constellations could also have been divided by:

- The first species that initiated flight could have defined the constellation. But this would also imply that LGB species other than greylag or barnacle would constitute flock constellations (figure 15).
- Flocks could have been defined by the numerically dominant species in the flock.

Even though FID in my case only caught the first flight initiating individual/s, it doesn't reveal the flight initiation distance for the subsequent goose/LGB individuals in the flock. It might be that the first flight initiating individuals/s, initiated flight much earlier than the subsequent individual/s. but the data collected here, doesn't reveal such gradual FID of individual/s within the flock. Such an approach would require spending far more time on each scaring trial. Probably, it'd also require to be more than one person in field: One who performs the actual scaring trials/walking, and another person who counts the species and their numbers of individuals for the gradually yielded FID-values. Large flock size should imply that it's more likely to encounter some individual/s in the flock that exhibit high FID-values. Simply because there is higher potential of variation of individual FIDvalues in a flock with many individuals compared to flocks of fewer individuals. . There was a positive correlation between flock size and FID, but it was not significant (figure 19). Perhaps this shows that geese are more confident and not as easily disturbed when they occur in larger groups, and that's why there's not a stronger correlation and effect of flock size on FID.

Finally, many more statistical tests and graphs could have been run and shown. For instance, the correlation tests were only run with the untransformed FID-values, whereas the linear modelling, uses the logarithmically transformed values. The correlation tests, could of course too, had been run and shown with the logarithmically transformed FID-values, too. The data presented in the results don't pay any respect to qualitative features (e.g., hedge or tree or car embankment) of the physical objects. But actually, this was registered during the data collection, too.

Parametric Anova or Ancova could then have been applied on the logarithmically transformed FID-values to see if there were any qualitative differences between objects. For instance, to see how physical objects of agroecological characteristics, e.g., perennials differed from other physical objects, e.g., road embankments The data set with qualitative descriptions of the physical objects is however included in appendix 1. I do happily share information about how the categorization of objects was done.

6. Conclusion

Geese are known to prefer large agricultural fields and search for fields that provide good visibility. Based on these adaptations in geese, this thesis set out to test if distance to physical objects had an influence on how easily geese are scared on agricultural land.

In February and March 2021, 151 scaring trials were performed on flocks of greylag and barnacle geese in Scania. The hypothesis was that barnacle and greylag geese would be more sensitive to scaring/disturbance if they were closer to physical objects such as trees, road embankments and topographical slope. The hypothesis was clearly refuted. Flocks that contained greylag geese but no barnacle geese, and mixed flocks of barnacle and greylag geese were even less sensitive to scaring/disturbance when they were closer positioned to physical objects. Flocks containing barnacle geese, but no greylag geese, were more easily scared when they were closer to physical objects, but this wasn't significant, and no inference can be drawn to the population.

Validity issues during data collection were present as the accuracy of measuring FID and distance to objects could sometimes be questioned. For instance, it was harder to measure these variables when it was wet and muddy. A more accurate way to measure distance to objects would probably be with GIS. All coordinates of the scaring trials are found in appendix 4 for possible GIS analysis.

Since agroecology aims to balance different pest problems against each other, the idea of a patchier agricultural landscape should still not be dismissed in a holistic analysis of different pest problems. As such, these landscapes may provide many other ecosystem services and can boost ecological intensification in agriculture. But given how barnacle geese and greylag geese behave in Scania during the data collection time frame, there's no evidence to say that a patchy landscape with more physical objects facilitate scaring of geese.

Potential future research questions from an agroecological approach towards the issue of geese as a pest in agriculture include:

- Do geese, that are part of different food webs, exhibit different sensitivity to distance to physical objects? I.e., How important are predator attacks on suspiciousness/sensitivity to physical objects and patchiness of agricultural landscapes among geese?
- In regions where geese are perceived as a nuisance in parks, golf courses etc. by the public: What's the potential of farmers communicating to the public/consumers how they work to decrease local goose pressure. Can such communication contribute to level four of agroecological conversion? I.e., can such communication contribute to connect those who produce the food and those who consume it?

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Appendix 1

Below is the entire dataset except coordinates that instead are found in appendix 4. Remember that just a few of the variables were finally used in the analysis. Please zoom in to fullest extent to read the observations in the table.

				Scating	Number						Number		Number								accordin	mellan WR Loc								
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133-135 2021 136-138 2021	1-02-27 10:00:00 Eastern 1-02-27 10:50:00 Eastern	Gardetinge Ellas	Sunny	walking			70							70 NA 15 NA	173.76746	7.5 V 5.5 V		315	176 ley growing 118 Kloverley growing	0-15	NA 150	26	13 103 17 193	2 topograp? 7 tree	107	tree tree	148	NA vindkraftvi	75) 512)	tar emban tree
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145-147 2021 146-150 2021	1-02-27 15:12:00 Eastern 1-02-27 15:29:00 Eastern	Harlösa Ellas Revioerby Ellas	Sunny Sunny	walking			15	1880		-				1880 NA 17 unkrown	116.68872	5.8 V 4.5 V	A	Multiple 243	226 winter cer growing 180 Clover ley growing	0-15	90	2 33	14 214	tree topograph	180	/ topograph 6 topograph	178	tree	213	tree topperaph
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265-262 2021	1-03-10 14:39:00 Souther	m Bröddarp Ellas	Sunny	walking			11	250		-				261 simultane	149.46830	5.4 SC	0 4	nultiple 48	202 Utgräud bi NA 137 likerbolme emaine	0-15	NA 01	NA 15	1 7	topograph	17	f topograph	41	topograph	4	topography
266-268 2021 269-271 2021	1-03-10 15:55:00 Souther	m Steglarp Ellas	Sunny	walking			36	3	_	-		8	-	47 1. greylag 2 NA	101.97708	5.4 50	0 8	110 to 140	34 winter cereals 100 winter cereals		100	2 17	19 14	topograph	11	I topograph	45 Without m	topograph Without or	85	topography
272-274 2021	1-03-10 17:11:00 Souther	m Vellinge vi Ellas	Sunny	walking			9							9 NA	114.00637	3.8 50	0	153	61 winter wheat	0.17	8	9 16	18 144	topograph	285	house	Without m	Without ry	30	topography
278-280 2021	1-03-12 09:45:00 Northe	m Norra Lorr Ellas	Cloudy	watking			9							9 NA	145.35615	5.8 V 5.7 V		281	247 winter wheat	0-15	131	S NA	54	s busnes S bree	100	topograph topograph	65	topograph	81	car emban
281-283 2021 284-286 2021	1-03-12 10:22:00 Northe 1-03-12 10:59:00 Northe	m Lila Lonve Ellas	Cloudy Cloudy	walking			14							14 NA 2 NA	94.972533	2.2 V 5 SA	v	222	41 winter cer growing 175 winter cereals	0-15	NA SI	5 15	1 141 77 Without r	Without n	176	5 house	162 NA	NA	147)	car emban
288-290 2021 291-293 2021	1-03-12 11:33:00 Northe 1-03-12 13:26:00 Northe	m Borgeby Ellas m Stenbocks/Ellas	Cloudy Cloudy	walking walking			2							2 NA 7 NA	146.47965 194.03693	5.1 V 6.5 SN	v	294	18 bare soil NA 190 winter rap growing	0-15	0 NA 187	17	14 53	1 topograph 1 tree	NA SS	NA	273	tree topograph	262	car emban
294-296 2021 297-299 2021	1-03-12 14:20:00 Northe 1-03-12 14:40:00 Northe	m Barseblick Ellas m Löddeköpi Ellas	Cloudy at Cloudy	n walking walking	-		46			-		-	-	46 NA 2 NA	104.35626	1.5 V 3.1 V		nultiple 282	10 winter cer growing 353 winter wh growing	0-15	91	1 17	16 85	previous y car embas	24	s previous y 8 car embar	67 126	car emban	45 Without re	topography Without re
300-302 2021 303-305 2021	1-03-12 15:36:00 Northe 1-03-12 16:25:00 Northe	m Löddeköpi Ellas	Cloudy Cloudy	walking			9			-	6			9 NA 44 simultan	78.870394	4.4 SA 4.1 SA	v	231 265	323 winter wheat 302 winter wheat		82	1 20 5 25	25 164 10 158	topograph tree	224	tree topperaph	113	previous y topograph	173	reed aggre topperaph
306-308 2021	1-03-12 16:53:00 Northe	m Stängby ky Ellas	Cloudy Cloudy #	walking			20			-				20 NA	124 13253	2.9 S		145 multiple	25 winter whi growing 318 winter whi ensuing	0-15	NA 111	15	18 193	topograph	Without r	wWithout re	213	tree	748	house
313-315 2021	1-03-15 10:56:00 Free	Köpingebr Ellas	Cloudy	walking			2							2 NA	45.918386	1.9 5		150	61 Hästbete growing	0-15	21		16 5	s reed aggre	60) topograph	64	reed aggre	5	reed aggre
319-318 2021	1-03-15 12-25:00 Free	Balkiles Clas	Cloudy	watking			2							2 NA	71.173380	2.1 5		326	163 winter whi growing	0-15	6	2 5	11 33	2 topograph	154	topograph topograph	12	bushes	42	topo@raph)
322-324 2021 325-327 2021	1-03-15 12:51:00 Free 1-03-15 13:17:00 Free	Splerups ky Ellas	Cloudy Cloudy	walking			2							2 NA 7 NA	103.96790	1.1 S 1.4 S		274	263 Mkt klent (growing 169 winter wh growing	0-15	93	2 13	15 21 11 114	l tree I topograpi	54	i topograph J topograph	18	topograph topograph	190)	topography
328-330 2021 331-333 2021	1-03-15 14:03:00 Free 1-03-15 14:20:00 Free	Skivarp Ellas Skivarp Ellas	Cloudy Cloudy	walking walking			5 37							5 NA 37 NA	164.73848 116.91853	2.6 S	c	Circulating 280	67 winter whi growing 300 winter cer growing	0-15	171	5 26 6 25	14 Without n 21 388	wWithout n I house	496	i house i housebil	115	car emban car emban v	226 Without re	dree Without re
334-336 2021 337-329 2021	1-03-15 15:35:00 Free 1-03-16 10:00:00 Eastern	Klörup Ellas YddingesjöEllas	Cloudy Cloudy	walking walking	-		22	32		-		-	-	22 NA 39 1. greylag	98.907635	1.5 S 3.1 N		180	196 winter whi growing 310 winter whi growing	0-15	90	0 14 5 NA	55	car embar 5 topograpi	209	 topograph topograph 	84 32	topograph tree	68) 132)	topography topography
340-342 2021	1-03-16 10:30:00 Eastern	Hyby Ellas	Cloudy	walking			2							2 NA	89.277243 95.131477	3.6 N		132	114 winter whi growing 159 winter whi entwine	0-15	71	5 13	13 Without r	Without n	Without r	w Without re	79	topograph tree	349	topography
346-348 2021	1-03-16 11:32:00 Eastern	Assartorp Elias	Cloudy	walking			16							16 NA	44.695165	3.2 N	0	14 30	1/290 without re growing	0-15	31	1 13	16 51	topograph	87	topograph	55	topograph	99	topography
352-354 2021	1-03-16 12:27:00 Eastern	Genarp Ellas	Cloudy	walking			2							2 NA	77.231165	1.1 N	N	303	1 winter wh growing	0-15	61	5 20	33 13	7 property 7	199	topograph	100	car emban	250	steenwith
2021	1-03-16 13:50:00 Eastern	Alberta Elias	Cloudy	walking			12							7 NA	133 52654	2.9 N	-	172	122 winter whi growing	0-15	110	1 26	160	house	83 Without r	v Without re	204	topograph	148	car emban
2021 266-368 2021	1-03-17 07:12:00 Souther 1-03-17 10:58:00 Souther	m Västra Kär Ellas	Cloudy	walking walking			2	20				1		20/NA 2 NA	au 007785 129 35080	1.5 N	N 1	186	241 dare soll NA 165 winter whi growing	0-15	122	2 15	627	topograph	140	bush	106	topograph	111	bush
372-374 2021	1-03-17 12:15:00 Souther	m Treleborg Elas	Cloudy	warking walking			63 17							64 NA 17 NA	273.58486	2.2 Ö		118 Multiple. t	221 ley growing	0-15	177	9 NA	183	topograpi	25	tree	230	house	78	topography
375-377 2021 378-380 2021	1-03-17 14:14:00 Souther 1-03-17 15:20:00 Souther	m Västra Vär Ellas m Gessle villi Ellas	Sunny Cloudy	walking walking			4	1040						4 NA 1010 NA	118 76064 53 311507	0.5 SA 2.6 Ni	v a	nultiple 298	108 winter whi growing 236 without re growing	0-15	94	4 15 0 15	15 353 16 35	7 house 9 buffert zo	14	buffert apr	107	buffert and reed aggre	28	buffert zon
381-383 2021 384-386 2021	1-03-17 15:58:00 Souther 1-03-17 17:04:00 Souther	m Gessie vilu Elias m Klagshame Elias	Cloudy Cloudy	walking walking			14	5						14 NA 5 NA	76.587158 43.724062	0 No 1.8 SA	A V	351 165	104 winter wheat 110 winter whi growing	0-15	53	7 13	16 253 15 330	previous y house	247	JNA Vilastaket	197	horse ferc reed aggre	157 Without re	previous ye Without re
287-389 2021 393-395 2071	1-03-18 09:54:00 Northe 1-03-18 12:51:00 Northe	m Alnarp Ellas m Svalöv nor Ellas	Sunny Sunny	walking walking			6							6 NA 23 NA	167.81201 50.413361	2.2 N 2.3 N	N	106 multiple. b	179 winter whigrowing 207 winter wheat & have	0-15	167	NA	48	tree house	60 Without •	/ topograph without ~	622 155	house topograph	134	house topograph-
296-298 2021 200-401 2021	1-03-18 15:07:00 Northe	n Stenbocks/Ellas	Sunny	walking			5							S NA	72 511685	4.3 N	N I	184	61 winter whi growing	0-15	NA	11	17 320) tree	72	t topograph	34	haimtäckn	Without re	Without re
402-404 2021	1-03-18 16:05:00 Northe	m Barseblick Ellas	Sunny	walking			17	(1)	anns i flocken i	ntialt men	d	2		19 1. greylag	76.947957	2.1 N		141	118 winter whi growing	0-15	7	10 10	40	l car embar	25	reed aggre	25	topograph Witherste	21	topography
410-412 2021	1-03-19 09:18:00 Free	Kristineber Ellas	Sunny	watking			12							4 NA	st 404./21 58.014596	NA N	A I	85	156 winter cer growing	0-15	93	7 11	~ 41 11 115	car embar	- 65 r 100 topog	e topograph	129	järnvägssa	286	topographi
413-415 2021 416-418 2021	1-03-19 09:46:00 Free 1-03-19 10:25:00 Free	Arrie Elias Västra Ing Elias	Sunny	walking			2							2 NA 2 NA	193.46422 165.97426	2 N 4.1 N	0	100	203 winter whi growing 229 winter whi growing	0-15	100	S NA 9 NA	91	L topograph 5 topograph	1 30	i topograph J previous y	111	tree topograph	21	topography
419-421 2021 422-424 2021	1-03-19 12:08:00 Free 1-03-19 13:35:00 Free	Smygehart Ellas Näsbyholn Ellas	Sunny Sunny	walking walking			6				(1) Står å	xar på 56 m 8	n från WP3	6 NA 76 simultane	55.134620 92.215028	1.8 N 2.1 N	0	20 85 NA	39 without re growing A without re growing	0-15	60	2 NA	13 538	35 I tree	reed aggro 24	e 80 4 reed aggre	topograph 62	236 house car emban	47	topography topography
425-427 2021 428-430 2021	1-03-19 14:15:00 Free 1-03-19 15:13:00 Free	SkWithout Ellas Svaneholm Ellas	Sunny Sunny	walking walking			2					2		4 1. greate 5 NA	84.651490 57.681633	2.9 Ni 2.9 Ni	0	315	210 winter whi growing 185 without reach # ~dw	0-15 finierat ~	NA 53	NA 37	40	topograph tree	1	topograph topograph	62 #1	topograph previous v	58) 91	topography car embary
431-433 2021	1-03-19 15:46:00 Free 1-03-19 16:30:00 Even	Janstorp Elias	Sunny	walking	-		5			-			-	S NA 2 NA	73 923892	1.8 N	0 2	210-280	285 lay growing 61 winter who array	0-15	61	S NA	91	topograph	51 16 town	topograph	14	topograph topograph	22	previous ye
437-439 2021	1-03-21 10-38:00 Eastern 1-03-21 11:38:00 Eastern	Gårdstång Ellas Harlösa Ellas	Sunny	walking			14							14 NA 2 NA	158 58100	2.8 N	N	235	233 winter whi growing 272 without na array	0-15	153	7 16	11 88	Car embar	117	topograph E topograph	289	tree	324	tree topogrand
442-445 2021	1-03-21 13-20:00 Eastern	Slydkra Ellas	Sunny	walking			2	0						2 NA	44.947825	3 1	N	275	345 without re growing	0-15	4	2 13	10 171	i vilastaket	17	f car emban	Without IN	Without re	125	tree
450-452 2021	1-03-21 14:58:00 Eastern	i leyby Ellas	Sunny	watking			30	54				2		32 1. greylag	178.60748	SN SN	N	2.84	6 winter whi growing	0-15	145	2 23	12 264	topograpi	41	r topograph	129	topograph	AL D	NA
454-455 2021 456-458 2021	1-03-21 15:27:00 Eastern 1-03-22 10:02:00 Souther	n Alberta Ellas m Klagshamt Ellas	Sunny Cloudy	walking walking			7							7 NA 14 NA	65.872622 76.790005	6 N	0 1	214 E 197	109 winter whi growing 88 winter cer growing	0-15	63	a S DNA	90	topograpi topograpi	1 NA 1 34	NA I bush	68 Without n	träskjul Without re	60) 140	salmbal vilastaket
459-461 2021 462-464 2021	1-03-22 11:15:00 Souther 1-03-22 12:37:00 Souther	m Maglarp Ellas m Amie Ellas	Cloudy Cloudy	walking walking			4							4 NA S NA	42 815002	2.5 S 2.2 S		194 148	132 bare soll, v NA 158 winter cer growine	0-15	20	9 NA 5 17	63 19 164	l reed aggre	107	topograph tree	417	car emban topograph	30	topography stener/With
465-468 2021 469-471 2021	1-03-22 12:40:00 Souther	m Östra Grev Ellas m Arrie Ella*	Cloudy	walking walking			2							2 NA 4 NA	53 646215 54 978805	2.4 S		61 210	40 winter rap growing 159 winter rap growing	0-15	ST NA	7 NA	60	topograph	40	/ reed aggre	167	topograph car ember	66	ceed aggre
472-474 2021	1-03-22 15:30:00 Souther	m Lockarp Ellas	Cloudy	walking			2							2 NA 2 NA	49 360248	NA N.	A	214	15 winter whi growing 222 historia anari	0-15	4	7 10	10 m	topowers -	31 Without -	reed aggre	134	previous y	261	property he
478-480 2021	1-03-22 15:57:00 Souther	m Gessie ville Ellas	Cloudy	walking				9						9 NA	43 232943	1.6 SA	v a	NA	129 lay med at growing	15-30	44	4 20	N 61	i jordhög	225	1 bush	90	property h	126	tree
494-486 2021	1-03-22 09:31:00 Souther	n Bernstorp Ellas	Cloudy	watking			s	390						385 1. barnac 290 NA	28.466366	1.7 V		multiple b	343 winter whi growing	0-15	31	1 16	15 122	ceed aggre	avishout r	house	302	reed aggre	494	reed aggre
eau-489 2021 490-492 2021	1-03-23 11:29:00 Northe 1-03-23 12:21:00 Northe	m Nybo Ellas m Flädle Ellas	Cloudy Cloudy	walking walking			5	185						5 NA 185 NA	62.455168 162.08566	1.5 V 0.5 V		214 multiple. r	261 without re growing 132 winter cer growing	0-15	55	/ 13 I NA	11	villastaket I tree	1 25	topograph	64 103	tree	55	sopography vass
492-495 2021 497-499 2021	1-03-23 13:31:00 Northe 1-03-23 14:34:00 Northe	m Bjärred sö Ellas m Bjärred ös Ellas	Cloudy Cloudy	walking walking			4							4 NA 4 NA	45 902585	2.2 SA 0 No	V A	168 53	170 winter rap growing 294 winter whi growing	15-30 15-30	40	2 S	12 503 15 100	2 house 5 höbel	Without r 248	#Without re a topograph	123	previous y car emban	205	vilastaket previous ye
500-502 2021	1-03-23 15:29:00 Northe	n Alnarp Ellas	Cloudy	walking			5	430		-			-	435 1. barnac	26.578492	2.9 SA	v a v	nultiple. b	258 winter whi growing 143 winter whi arrow	15-30	60	2 16	13 31	topograph	100	buffert and	75	topograph topograph	236	tree tree
508-508 2021 509-511 2021	1-03-23 17:00:00 Northe 1-03-23 17:47:00 Morthe	m Bjärred Ellas	Cloudy	walking			6				9	-		6 NA 20 sim-base	84.856252	1.7 SA	v s v	190-280	241 bare soil NA 82 bare soil 8 NA	0-15	0 80	2 23	12 208	I property 7	171	i tree 2 vietiche	45	tree tooorach	36	tree topperar**

Appendix 2

The most recently systematically collected data of FID among LGB in Scania, were collected between 2020-11-11 and 2021-03-14. These data were collected on the initiative by my supervisors in the Kristianstad municipality. The data are yet unpublished, but with their permission, they can be used for quick comparisons to the data collected for this thesis. An advantage in the comparison is that the methods are the same for both datasets. In the discussion, the results from Kristianstad will be contrasted to the results of this thesis.

In total 201 scaring trials were conducted in the Kristianstad dataset, and 143 of these could be categorized as T-geese constellations, see figure 1. The overall mean for T-geese was 127.80 meters and the median was 118.34 meters. As discerned by figure 2, FID-values were not normally distributed.



Figure 1. Flock constellations during scaring trials in the Kristianstad municipality. Since these scaring trials included all LGB, the constellation "other" is also included. In total, 201 scaring trials were performed. 143 of these regarded T-geese (i.e., barnacle, greylag and mixed flocks).



Figure 2. Histogram when T-geese are selected in the dataset of scaring trials in Kristianstad.

Comparisons between constellations are shown in figure 3, and since greylag and mixed constellations exhibited unnormal distribution, a Kruskal-Wallis test was run to test differences in FID between the constellations. The Kruskal-Wallis test produced a p-value of 0.4985, which refuted the potential of significant inferential differences between the constellations.



Figure 3. FID compared between constellations. Whiskers are set as 1.5*Inter quartile range. Values outside whiskers are considered as outliers and depicted as dots.

Appendix 3

Below are the linear models of the multiple regression showed with untransformed FID-values.

Table 1. Coefficients of the multiple linear regression describing greylag goose flocks. Unransformed FID-values.

Coefficient	Estimate	Std. error
Intercept	131.663	39.531
Distance to nearest	0.287	0.088
object		
Ordinal date	-0.713	0.527
Individuals	0.553	0.29

Table 2. Coefficients of the multiple linear regression describing barnacle goose flocks. Untransformed FID-values.

Coefficient	Estimate	Std. error
Intercept	25.589	193.116
Distance to nearest object	-0.023	0.257
Ordinal date	0.451	2.354
Individuals	0.028	0.037

Table 3. Coefficients of the multiple linear regression describing mixed flocks. UntransformedFID-values.

Coefficient	Estimate	Std. error
Intercept	101.81	70.504
Distance to nearest	0.408	0.165
object		
Ordinal date	-0.056	1.015
Individuals	-0.022	0.036



Figure 1. Linear model for all species using untransformed FID-values. Regression lines are based on the equations below which in turn are taken from table one to three.

Barnacle: y(FID)= -0.0232*Distance to nearest object + 0.02786 *Mean number of individuals + 0.45155*Mean ordinal date + 25.58930 (Intercept)

Greylag: y(FID) = 0.28669*Distance to nearest object + 0.55359*Mean number of individuals + -0.71264*Mean ordinal date + (131.66291) (Intercept)

Mixed: y(FID): 0.40783*Distance to nearest object + -0.02244*Mean number of individuals + 0.05646*Mean ordinal date + (101.80599) (Intercept)

Appendix 4

Below are the coordinates for all waypoints, WP1 are also included, and thus it's usually just each third ID that constitutes an FID value. Sometimes cancelled scaring trials or data collection errors imply longer jumps between FIDs to obtain the FIDs, compare with the FID column in appendix 1. For illustrative purposes, the first FID 11 FID-values are written out explicitly.

ID	Lat	Lon	ele	time	POINT_X	POINT_Y	Name	Pythagora's	FID
	(WGS84)	(WGS84)			(RT90)	(RT90)		theorem	
1	55,4567	13,00165	2,890549	2021-02-	1322485	6151019	1		
				12T09:41:18Z					
2	55,45649	13,00149	4,604938	2021-02-	1322474	6150995	2		
				12T09:44:39Z					
3	55,45559	13,00103	4,483433	2021-02-	1322441	6150897	3	103,8517	103,8517
				12T09:46:19Z					
4	55,51114	12,93719	1,447363	2021-02-	1318660	6157243	4	7386,742	
				12T11:06:06Z					
5	55,51072	12,93543	0,489635	2021-02-	1318547	6157201	5	120,779	
				12T11:09:19Z					
6	55,51035	12,93469	-0,40465	2021-02-	1318498	6157161	6	62,73842	62,73842
				12T11:10:35Z					
7	55,5153	12,93546	1,11306	2021-02-	1318570	6157710	7	553,4468	
				12T11:30:17Z					
8	55,51589	12,93572	1,298123	2021-02-	1318589	6157776	8	68,435	
				12T11:33:00Z					
9	55,51697	12,93658	1,123736	2021-02-	1318648	6157893	9	131,5699	131,5699
				12T11:38:53Z					
10	55,73709	13,34719	17,41227	2021-02-	1345442	6181394	10	35639,42	
				12T12:32:25Z					
11	55,7373	13,3461	16,32608	2021-02-	1345374	6181420	11	72,52994	
				12T12:35:04Z					
12	55,73785	13,34461	18,6179	2021-02-	1345283	6181484	12	111,5269	111,5269
				12T12:37:45Z					

13	55,72995	13,34043	21,67617	2021-02- 12T13:04:08Z	1344989	6180614	13	918,0171	
14	55,73002	13,34019	23,33881	2021-02- 12T13:07:08Z	1344974	6180623	14	17,0208	
15	55,73026	13,33923	20,60979	2021-02- 12T13:09:25Z	1344914	6180652	15	66,51753	66,51753
16	55,73535	13,3697	20,65637	2021-02- 12T13:32:51Z	1346848	6181150	16	1996,869	
17	55,73484	13,36973	19,71408	2021-02- 12T13:35:55Z	1346848	6181093	17	56,7053	
18	55,73433	13,36974	17,27806	2021-02- 12T13:38:02Z	1346847	6181037	18	56,69253	56,69253
19	55,65065	13,05591	9,152039	2021-02- 18T09:38:51Z	1326772	6172463	25	21829,07	
20	55,65079	13,05632	7,73597	2021-02- 18T09:44:10Z	1326798	6172478	26	30,20731	
21	55,65117	13,05698	8,697367	2021-02- 18T09:46:16Z	1326841	6172519	27	59,55539	59,55539
22	55,73063	13,04194	13,73142	2021-02- 18T10:34:43Z	1326248	6181398	28	8899,434	
23	55,73028	13,04175	13,51948	2021-02- 18T10:40:07Z	1326235	6181359	29	41,38508	
24	55,72866	13,04162	14,12552	2021-02- 18T10:43:48Z	1326219	6181180	30	180,1463	180,1463
25	55,74231	13,03021	14,53399	2021-02- 18T12:09:00Z	1325564	6182728	31	1680,897	
26	55,74231	13,03022	12,59155	2021-02- 18T12:14:18Z	1325564	6182727	32	0,706752	
27	55,74261	13,03064	10,24581	2021-02- 18T12:15:42Z	1325592	6182759	33	42,51999	42,51999
28	55,74232	13,03016	10,78886	2021-02- 18T12:23:33Z	1325561	6182729	34	43,57692	
29	55,74233	13,03017	11,9274	2021-02- 18T12:25:40Z	1325561	6182729	35	0,550956	
30	55,74242	13,03423	10,52817	2021-02- 18T12:30:58Z	1325816	6182730	36	255,1645	255,1645
31	55,7546	13,05286	13,08635	2021-02- 18T15:11:27Z	1327040	6184038	37	1791,447	
32	55,7546	13,05286	14,54801	2021-02- 18T15:13:11Z	1327039	6184038	38	0,54831	

33	55,75479	13,05434	14,07591	2021-02- 18T15:15:35Z	1327133	6184055	39	95,3756	95,3756
34	55,73413	13,34468	23,93232	2021-02- 19T09:03:26Z	1345272	6181070	40	18382,9	
35	55,73413	13,34468	23,82601	2021-02- 19T09:06:36Z	1345272	6181070	41	0,448763	
36	55,73434	13,34203	0	2021-02- 19T09:13:59Z	1345107	6181100	42	168,0579	
37	55,73175	13,37694	17,22404	2021-02- 19T09:54:45Z	1347289	6180733	43	2212,41	
38	55,73175	13,37683	17,41389	2021-02- 19T10:02:06Z	1347282	6180734	44	6,947574	
39	55,73257	13,38068	16,00521	2021-02- 19T10:09:19Z	1347526	6180817	45	258,3565	
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46	55,68512	13,42426	35,45119	2021-02- 19T13:24:24Z	1350082	6175441	52	6456,721	
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50	55,51634	12,93508	20,72131	2021-02- 22T11:23:15Z	1318550	6157827	56	8,937145	
51	55,51686	12,9372	18,91805	2021-02- 22T11:27:14Z	1318687	6157880	57	146,2484	
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57	55,50788	12,93488	0,254358	2021-02- 22T12:12:36Z	1318499	6156887	63	80,21738	
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68	55,40252	13,02013	-3,74203	2021-02- 22T14:52:10Z	1323412	6144943	74	31,26372	
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71	55,57588	13,17221	17,16873	2021-02- 23T09:00:14Z	1333773	6163858	83	65,24946	
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73	55,58311	13,29339	33,30159	2021-02- 23T09:29:55Z	1341442	6164380	85	7615,098	
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76	55,60467	13,38365	29,20879	2021-02- 23T10:05:01Z	1347215	6166577	88	6136,054	
77	55,60477	13,38367	28,20224	2021-02- 23T10:06:08Z	1347217	6166587	89	10,81081	
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79	55,60434	13,38692	30,84442	2021-02- 23T10:22:09Z	1347420	6166533	91	251,7603	
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81	55,60583	13,38721	30,86178	2021-02- 23T10:29:55Z	1347445	6166698	93	108,9419	
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86	55,44985	13,7275	32,81448	2021-02- 23T13:36:30Z	1368363	6148643	98	93,40907	
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93	55,42996	13,55532	13,35725	2021-02- 23T14:43:20Z	1357400	6146768	105	127,1703	
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95	55,44404	13,46651	54,55748	2021-02- 23T15:19:39Z	1351833	6148522	107	127,2192	
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98	55,4299	13,43456	52,4895	2021-02- 23T15:49:08Z	1349758	6147017	110	76,61002	
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176	55,38552	13,45999	8,57732	2021-03- 04T14:32:24Z	1351201	6142023	188	93,54472	
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186	55,4328	13,36876	49,07674	2021-03- 04T16:06:58Z	1345606	6147483	198	4015,442	
187	55,43209	13,37014	45,18679	2021-03- 04T16:10:10Z	1345690	6147401	199	117,5347	
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198	55,72467	13,03697	6,322813	2021-03- 05T13:19:57Z	1325910	6180747	210	2030,674	
199	55,72591	13,03658	5,418687	2021-03- 05T13:24:07Z	1325890	6180887	2111	140,7156	
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202	55,72157	13,06856	0,366501	2021-03- 05T14:34:32Z	1327880	6180324	214	141,4826	
203	55,72253	13,0701	0,51335	2021-03- 05T14:37:35Z	1327981	6180426	215	143,6904	
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205	55,68296	13,09354	3,031807	2021-03- 05T15:50:48Z	1329280	6175965	217	10,74911	
206	55,68359	13,09447	1,781769	2021-03- 05T15:53:48Z	1329342	6176033	218	91,61777	
207	55,68376	13,09474	0,700844	2021-03- 05T15:54:36Z	1329359	6176051	219	25,39313	
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217	55,74377	13,41092	25,81076	2021-03- 09T09:33:55Z	1349469	6181997	229	52,16226	
218	55,78968	13,36491	53,71539	2021-03- 09T10:11:10Z	1346761	6187207	230	5872,094	
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222	55,7796	13,36222	55,72376	2021-03- 09T11:00:24Z	1346552	6186092	234	83,32364	
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224	55,68722	13,23886	24,69626	2021-03- 09T12:39:39Z	1338434	6176091	236	12858,27	
225	55,68731	13,23681	22,57012	2021-03- 09T12:43:39Z	1338306	6176106	237	129,5378	
226	55,68732	13,23607	23,10187	2021-03- 09T12:45:30Z	1338259	6176108	238	46,80783	
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228	55,76921	13,31973	28,06417	2021-03- 09T14:07:44Z	1343846	6185030	240	29,90678	
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230	55,6876	13,10522	2,695009	2021-03- 09T17:03:57Z	1330035	6176453	242	16067,2	
231	55,68695	13,10648	3,665304	2021-03- 09T17:06:56Z	1330111	6176378	243	107,6943	
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238	55,50308	12,94447	6,981837	2021-03- 10T09:40:45Z	1319083	6156327	250	93,34122	
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240	55,52063	12,94144	8,249506	2021-03- 10T10:04:49Z	1318971	6158288	252	36,14364	
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255	55,4512	13,12369	37,69982	2021-03- 10T15:01:41Z	1330178	6150102	267	77,81162	
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266	55,66312	13,07341	9,133387	2021-03- 12T08:45:58Z	1327927	6173807	278	1625,608	
267	55,66296	13,07275	10,5332	2021-03- 12T08:47:22Z	1327885	6173791	279	44,91544	
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282	55,76679	12,95882	4,039732	2021-03- 12T13:21:35Z	1321194	6185633	295	24,07211	
283	55,76749	12,95992	3,932522	2021-03- 12T13:23:49Z	1321267	6185709	296	104,3563	
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303	55,4469	13,80487	37,92692	2021-03- 15T10:55:52Z	1373248	6148171	317	22,34175	
304	55,4467	13,80602	30,53236	2021-03- 15T10:57:48Z	1373319	6148147	318	75,49602	
305	55,44872	13,72913	34,29351	2021-03- 15T11:25:26Z	1368462	6148514	319	4871,063	
306	55,44865	13,72908	32,87208	2021-03- 15T11:27:02Z	1368459	6148506	320	8,003971	
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309	55,42834	13,66292	25,50002	2021-03- 15T11:52:56Z	1364204	6146373	323	34,58568	
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313	55,43881	13,62871	37,39048	2021-03- 15T12:20:34Z	1362075	6147606	327	111,8053	
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316	55,4231	13,55009	16,73498	2021-03- 15T13:08:44Z	1357045	6146017	330	164,7385	
317	55,42208	13,54621	18,22368	2021-03- 15T13:20:41Z	1356795	6145910	331	271,131	
318	55,42224	13,545	17,18327	2021-03- 15T13:22:55Z	1356719	6145931	332	78,82174	
319	55,42271	13,54335	15,82754	2021-03- 15T13:24:54Z	1356616	6145986	333	116,9185	
320	55,45625	13,17015	31,98522	2021-03- 15T14:35:52Z	1333138	6150551	334	23918,23	
321	55,45582	13,17038	30,39816	2021-03- 15T14:37:51Z	1333151	6150503	335	49,63313	
322	55,45499	13,17093	29,12004	2021-03- 15T14:39:33Z	1333182	6150410	336	98,90764	
323	55,55639	13,25546	44,70738	2021-03- 16T09:00:35Z	1338942	6161493	337	12490,99	
324	55,55656	13,25513	46,34395	2021-03- 16T09:03:31Z	1338922	6161513	338	28,49875	
325	55,55727	13,25329	47,01405	2021-03- 16T09:06:10Z	1338809	6161596	339	140,6684	
326	55,58156	13,26768	35,78939	2021-03- 16T09:30:01Z	1339815	6164267	340	2853,66	
327	55,58152	13,26841	35,44519	2021-03- 16T09:31:21Z	1339861	6164260	341	46,72381	
328	55,58135	13,2698	37,61789	2021-03- 16T09:33:22Z	1339948	6164238	342	89,27724	
329	55,57468	13,2989	39,19137	2021-03- 16T09:53:49Z	1341755	6163430	343	1980,242	
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331	55,57316	13,29921	36,61025	2021-03- 16T09:57:16Z	1341769	6163259	345	96,13147	
332	55,58341	13,3472	40,63937	2021-03- 16T10:32:43Z	1344835	6164292	346	3235,722	
333	55,58363	13,3457	38,46458	2021-03- 16T10:36:53Z	1344741	6164319	347	97,9044	
334	55,58387	13,34513	36,54791	2021-03- 16T10:38:51Z	1344707	6164347	348	44,69517	

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336	55,57729	13,35933	48,57607	2021-03- 16T11:03:31Z	1345575	6163583	350	45,19312	
337	55,57685	13,35917	46,80976	2021-03- 16T11:04:33Z	1345564	6163535	351	49,60258	
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372	55,53197	12,92527	1,936439	2021-03- 17T16:08:01Z	1318004	6159592	386	43,72406	
373	55,6564	13,06192	11,17529	2021-03- 18T08:54:44Z	1327175	6173088	387	16317,84	
374	55,65529	13,06167	11,56152	2021-03- 18T08:57:04Z	1327155	6172965	388	124,8925	

375	55,65379	13,06151	11,61173	2021-03- 18T08:59:24Z	1327138	6172798	389	167,812	
376	55,75633	13,01975	2,639942	2021-03- 18T10:10:35Z	1324970	6184314	390	11718,2	
377	55,75617	13,02171	3,738016	2021-03- 18T10:13:17Z	1325092	6184291	391	124,6238	
378	55,75612	13,02237	1,705357	2021-03- 18T10:15:03Z	1325133	6184284	392	41,41552	
379	55,93341	13,11006	68,67542	2021-03- 18T11:51:47Z	1331405	6203795	393	20493,73	
380	55,93308	13,10977	70,7884	2021-03- 18T11:53:08Z	1331386	6203759	394	40,87895	
381	55,93267	13,10943	66,53445	2021-03- 18T11:54:08Z	1331363	6203714	395	50,41336	
382	55,77967	12,93478	0,384895	2021-03- 18T14:07:26Z	1319746	6187129	396	20248,79	
383	55,7799	12,93538	2,025318	2021-03- 18T14:08:43Z	1319785	6187153	397	45,56479	
384	55,78026	12,93634	5,381477	2021-03- 18T14:10:01Z	1319846	6187191	398	72,51169	
385	55,76274	12,92092	5,044388	2021-03- 18T14:27:54Z	1318798	6185282	399	2177,778	
386	55,76401	12,91805	3,062504	2021-03- 18T14:31:31Z	1318624	6185430	400	228,6272	
387	55,76455	12,91675	3,019287	2021-03- 18T14:35:05Z	1318545	6185494	401	102,1891	
388	55,7666	12,95867	4,074963	2021-03- 18T15:05:30Z	1321184	6185613	402	2642,37	
389	55,76644	12,95904	4,887672	2021-03- 18T15:07:57Z	1321207	6185593	403	29,55802	
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391	55,77072	13,05658	9,560856	2021-03- 18T16:01:05Z	1327344	6185823	405	6086,264	
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393	55,76072	13,09729	27,69247	2021-03- 18T16:27:07Z	1329854	6184609	407	2767,433	
394	55,76002	13,09717	25,061	2021-03- 18T16:30:28Z	1329844	6184531	408	78,53237	

395	55,75921	13,09707	23,79665	2021-03- 18T16:32:08Z	1329834	6184442	409	90,40432	
396	55,5491	13,07087	44,67797	2021-03- 19T08:18:30Z	1327267	6161126	410	23456,51	
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398	55,54807	13,07131	44,33549	2021-03- 19T08:23:11Z	1327291	6161010	412	58,0146	
399	55,52187	13,08295	45,48644	2021-03- 19T08:46:01Z	1327910	6158066	413	3009,121	
400	55,52157	13,08262	41,15174	2021-03- 19T08:48:14Z	1327888	6158034	414	38,60371	
401	55,52027	13,0806	39,71286	2021-03- 19T08:51:01Z	1327755	6157894	415	193,4442	
402	55,48545	13,12384	40,01007	2021-03- 19T09:26:40Z	1330335	6153913	416	4744,22	
403	55,48525	13,12301	38,22682	2021-03- 19T09:28:12Z	1330282	6153892	417	56,66686	
404	55,48468	13,12059	32,99753	2021-03- 19T09:30:31Z	1330126	6153835	418	165,9743	
405	55,34997	13,38531	1,65882	2021-03- 19T11:08:31Z	1346332	6138229	419	22498,42	
406	55,35101	13,38704	-0,42658	2021-03- 19T11:11:37Z	1346446	6138341	420	159,9517	
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408	55,45921	13,47166	42,00013	2021-03- 19T12:35:34Z	1352215	6150199	422	13133,51	
409	55,45924	13,47168	42,48385	2021-03- 19T12:39:10Z	1352217	6150202	423	3,671989	I
410	55,46002	13,47119	40,08896	2021-03- 19T12:41:33Z	1352189	6150290	424	92,21503	
411	55,48652	13,47579	48,18353	2021-03- 19T13:15:22Z	1352579	6153229	425	2965,41	
412	55,48645	13,47554	46,75659	2021-03- 19T13:16:45Z	1352562	6153221	426	18,08989	
413	55,48592	13,47457	46,35252	2021-03- 19T13:18:16Z	1352500	6153165	427	84,65149	
414	55,50729	13,49185	55,45178	2021-03- 19T14:13:44Z	1353671	6155506	428	2618,35	

415	55,50685	13,49183	54,1338	2021-03- 19T14:15:13Z	1353667	6155458	429	49,03109	
416	55,50633	13,49182	53,04097	2021-03- 19T14:16:21Z	1353665	6155400	430	57,68163	
417	55,51755	13,51656	59,60595	2021-03- 19T14:46:07Z	1355269	6156597	431	2001,439	
418	55,51791	13,51554	56,58135	2021-03- 19T14:49:18Z	1355206	6156639	432	76,26804	
419	55,51821	13,51449	54,75571	2021-03- 19T14:51:45Z	1355141	6156674	433	73,92389	
420	55,52527	13,44283	72,70718	2021-03- 19T15:20:42Z	1350643	6157612	434	4594,852	
421	55,52574	13,44491	68,4007	2021-03- 19T15:23:12Z	1350776	6157660	435	141,3599	
422	55,52585	13,44596	68,59145	2021-03- 19T15:24:38Z	1350843	6157670	436	67,55019	
423	55,76965	13,31967	-4,93901	2021-03- 21T09:38:21Z	1343844	6185079	437	28288,76	
424	55,76961	13,3196	-6,38553	2021-03- 21T09:39:44Z	1343839	6185075	438	6,526876	
425	55,76924	13,31716	-13,0195	2021-03- 21T09:42:37Z	1343685	6185039	439	158,5811	
426	55,72248	13,50923	28,49767	2021-03- 21T10:38:11Z	1355562	6179418	440	13140,14	
427	55,72276	13,50805	29,21067	2021-03- 21T10:39:59Z	1355489	6179452	441	80,28909	
428	55,72296	13,50709	29,16967	2021-03- 21T10:41:17Z	1355429	6179477	442	64,51319	
429	55,67948	13,49834	19,01077	2021-03- 21T12:10:36Z	1354718	6174656	443	4873,133	
430	55,68025	13,49802	19,8106	2021-03- 21T12:12:35Z	1354701	6174742	444	88,3111	
431	55,68065	13,49794	18,49647	2021-03- 21T12:14:03Z	1354697	6174787	445	44,94783	
432	55,72324	13,5072	31,8952	2021-03- 21T13:04:58Z	1355437	6179507	446	4777,34	
433	55,72265	13,50773	30,73897	2021-03- 21T13:07:02Z	1355468	6179440	447	73,8314	
434	55,72207	13,50833	26,4206	2021-03- 21T13:11:27Z	1355503	6179375	449	73,95582	

435	55,64579	13,31632	21,13865	2021-03- 21T13:58:04Z	1343138	6171303	450	14766,96	
436	55,64751	13,31677	18,85397	2021-03- 21T14:01:12Z	1343173	6171493	451	193,1149	
437	55,64911	13,31697	20,67863	2021-03- 21T14:04:37Z	1343192	6171670	452	178,6075	
438	55,62862	13,33141	28,21481	2021-03- 21T14:27:36Z	1344019	6169358	453	2455,973	
439	55,6286	13,3319	28,09393	2021-03- 21T14:29:05Z	1344050	6169354	454	31,46769	
440	55,62856	13,33295	27,0866	2021-03- 21T14:30:47Z	1344116	6169348	455	65,87262	
441	55,52159	12,93791	54,24532	2021-03- 22T09:02:50Z	1318753	6158404	456	27623,69	
442	55,52154	12,93957	50,64675	2021-03- 22T09:05:17Z	1318858	6158393	457	105,1666	
443	55,52157	12,94078	49,76487	2021-03- 22T09:06:37Z	1318935	6158394	458	76,79001	
444	55,39209	13,08835	7,362122	2021-03- 22T10:15:56Z	1327686	6143611	459	17178,94	
445	55,39047	13,09141	4,735561	2021-03- 22T10:21:12Z	1327873	6143424	460	264,5239	
446	55,39011	13,09163	6,010231	2021-03- 22T10:23:13Z	1327886	6143383	461	42,815	
447	55,52025	13,09421	49,77419	2021-03- 22T11:37:42Z	1328614	6157858	462	14493,42	
448	55,51916	13,09483	46,16747	2021-03- 22T11:40:16Z	1328649	6157735	463	128,0778	
449	55,51875	13,09505	47,51655	2021-03- 22T11:41:35Z	1328661	6157689	464	47,63376	
450	55,46656	13,14196	35,94668	2021-03- 22T12:40:45Z	1331399	6151767	465	6524,115	
451	55,46629	13,14231	35,19114	2021-03- 22T12:42:10Z	1331421	6151736	466	38,0751	
452	55,4664	13,14279	34,77388	2021-03- 22T12:43:31Z	1331451	6151747	467	32,51932	
453	55,46675	13,14337	33,659	2021-03- 22T12:44:53Z	1331489	6151785	468	53,64622	
454	55,5202	13,08515	36,4001	2021-03- 22T13:07:49Z	1328042	6157874	469	6997,609	

455	55,52005	13,08524	36,38959	2021-03- 22T13:08:43Z	1328047	6157858	470	17,5643	
456	55,51958	13,0855	38,35184	2021-03- 22T13:09:35Z	1328061	6157805	471	54,97881	
457	55,53758	13,05471	31,09154	2021-03- 22T14:10:40Z	1326197	6159885	472	2793,296	
458	55,53811	13,05465	30,84057	2021-03- 22T14:12:38Z	1326196	6159943	473	58,70427	
459	55,53855	13,05457	32,53769	2021-03- 22T14:13:57Z	1326193	6159993	474	49,36025	
460	55,5268	13,04517	33,12975	2021-03- 22T14:29:20Z	1325547	6158708	475	1437,118	
461	55,52631	13,04478	30,43565	2021-03- 22T14:31:44Z	1325520	6158655	476	59,85107	
462	55,52587	13,04441	30,79541	2021-03- 22T14:32:55Z	1325495	6158607	477	54,02915	
463	55,50138	12,94531	3,251446	2021-03- 22T14:57:38Z	1319128	6156135	478	6830,485	
464	55,50138	12,9463	4,513065	2021-03- 22T14:59:19Z	1319190	6156133	479	62,77291	
465	55,50136	12,94699	4,101017	2021-03- 22T15:00:21Z	1319233	6156129	480	43,23294	
466	55,50638	12,94032	1,114475	2021-03- 22T15:18:24Z	1318836	6156705	481	699,7541	
467	55,50687	12,94187	2,999321	2021-03- 22T15:22:15Z	1318935	6156756	482	112,0235	
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471	55,6163	13,09967	7,975268	2021-03- 23T08:33:35Z	1329376	6168532	485	95,91648	
472	55,61663	13,09948	7,382671	2021-03- 23T08:34:48Z	1329365	6168569	486	38,46637	
473	55,7448	13,14076	22,97877	2021-03- 23T10:29:36Z	1332514	6182732	487	14508,17	
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475	55,74513	13,13868	22,3393	2021-03- 23T10:32:51Z	1332385	6182774	489	62,45517	
476	55,71901	13,06066	9,370071	2021-03- 23T11:21:49Z	1327372	6180059	490	5700,896	
477	55,7186	13,06172	6,377293	2021-03- 23T11:23:51Z	1327437	6180010	491	81,23662	
478	55,71767	13,06371	1,838721	2021-03- 23T11:33:24Z	1327558	6179902	492	162,0857	
479	55,70995	13,0539	2,026811	2021-03- 23T12:31:08Z	1326908	6179068	493	1057,526	
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				23T16:49:03Z					
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				23T16:50:31Z					
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				10T18:16:47Z					