Running head: Learning dietary specialization

Ontogeny shapes individual specialization

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ABSTRACT

Individual dietary specialization, where individuals occupy a subset of a population's wider dietary niche, is a key factor determining a species resilience against environmental change. However, the ontogeny of individual specialization, as well as associated underlying social learning, genetic, and environmental drivers, remain poorly understood. Using a multigenerational dataset of female European brown bears (*Ursus arctos*) followed since birth, we discerned the relative contributions of environmental similarity, genetic heritability, maternal effects, and offspring social learning from the mother to individual specialization. Individual specialization accounted for 43% of phenotypic variation and spanned half a trophic position, with individual diets ranging from omnivorous to carnivorous. The main determinants of dietary specialization were social learning during rearing (13%), environmental similarity (9%), maternal effects (11%), and permanent between-individual effects (8%), whereas the contribution of genetic heritability was negligible. The trophic position of offspring closely resembled the trophic position of their mothers during the first 3-4 years of independence, but this relationship ceased with increasing time since separation. Our study shows that social learning and maternal effects are as important for individual dietary specialization as environmental composition. We propose a tighter integration of social effects into future studies of range expansion and habitat selection under global change that, to date, are mostly explained by environmental drivers.

KEYWORDS

Dietary specialization, heritability, maternal effects, social learning, trophic position, trophic niche, omnivore, stable isotopes, nitrogen-15, *Ursus arctos*

1. INTRODUCTION

Among individuals of the same species, ecological niche variation is common and may occur when availability of food resources or habitat structure change across the species' range. Individual variation is key for driving species resilience in response to shifting resource availabilities in a rapidly changing world, and may ultimately determine local persistence or extinction of species [1]. Ecological generalists, species with a wide ecological niche, also seem to exhibit more individual specialization (i.e. between-individual variation of niche) [2] and are likely particularly well adapted to persist under shifts in resource availability or composition, enabling them to occupy larger distributional ranges than ecological specialists [3]. Inter- and intraspecific competition, predation and ecological opportunity alter resource availability and have been identified as the main ecological drivers explaining variation in the degree of individual dietary specialization among populations [4]. However, how individual variation in dietary specialization emerges and is maintained within populations has, to our knowledge, not been quantified in the wild.

In the fields of behavioral and evolutionary biology, individual variation is measured as the variance attributed to permanent between-individual differences, while the sources of variation can be quantified using complex hierarchical models (e.g., "animal model")[5]. In principle, three sources of variation are commonly considered [5, 6]: variation in the environment [7], additive genetic effects from which trait heritability can be estimated [8, 9], and parental (especially maternal) effects [10, 11]. In addition, individual variation can be maintained through social learning during ontogeny[12], an aspect that, to our knowledge, has rarely been integrated into animal models (but see [13, 14]). Here, we provide the first study to attribute individual variation in dietary niche to its sources.

Differences in the environment, in terms of habitat composition and associated availability of particular food resources, are generally considered the main cause of

individual variation in dietary specialization [15]. This is particularly true in range-resident

species, where individuals occupy a subset of the population's range and individual home

ranges vary in resource availability [16]. However, beyond the environment, resource

preferences have been suggested to be genetically heritable and determined through genes

inherited from both mother and father, where more closely related individuals share more

similar diets than distantly related individuals [15, 17]. Additionally, parental phenotypes may

also affect offspring phenotypes in ways other than genetic heritability [10, 11]. Maternal effects

are more commonly studied because mothers often have unilateral control over offspring

development [10], especially in mammals, however, paternal effects are plausible in species

with paternal care. Maternal effects on offspring behavior have been suggested to be lifelong,

they have either a genetic or environmental basis and summarize the cumulative influence of

many different proximate maternal effects, include pathways such as provisioning rates, milk

production, in-utero hormone transfer, and epigenetics [10]. Statistically, maternal effects

account for similarities in dietary niche among offspring of the same mother (fitted as a

random intercept for mother identity) [13, 14], but not for the similarity of dietary niche between

mother and offspring. The latter would be an example where the maternal trait affects the

offspring's trait, which statistically can be clearly differentiated from other maternal effects

[13, 14]. Similarities between the dietary phenotypes of mothers and their offspring indicate

social learning of resource preference or competence to secure a resource by the offspring

from the mother during early ontogeny [18, 19, 20, 21, 22]. Social learning is therefore an additional

pathway by which individual variation can be maintained. It is reasonable to assume that the

effects of social learning during rearing will weaken later in life through individual

experiential learning [23].

Attributing variation in diet to the individual level, to isolate its sources and to identify

developmental drivers of diet preferences in the wild requires multigenerational datasets of

repeated measures of the diet of individuals throughout their life [5]. We used a unique 30-year

longitudinal dataset of 71 female Scandinavian brown bears (*Ursus arctos*) of known mothers

with repeated annual isotopic estimates of trophic position to study, for the first time, the

sources of individual dietary specialization in the wild. Brown bears are ecological

generalists with a distribution range spanning the northern hemisphere from tundra to deserts,

paralleled by extensive variation in diet. Populations range from tracking food resource

pulses, such as spawning fish [24], scavenging on ungulate carcasses or preying on ungulate

calves [25], or feeding extensively on invertebrates, to populations using primarily fruiting

plant-based diets [26, 27]. Given such extreme dietary plasticity, it is not surprising that great

dietary variation has been found within populations [28, 29]; however, the determinants and

ontogeny of this variation at the individual level remain largely unknown. In ecology,

differences in diet are often primarily attributed to differences in resource availability and

abundance. Even within populations inhabiting a continuous biome, home-range scale

variation in habitat composition [30] can lead to variation in resource availability. Brown bears

maintain non-territorial home ranges and the most parsimonious source of individual

specialization is, therefore, heterogeneity in the environment. It is further plausible that

individual specialization in brown bears is genetically heritable. For example, while body

size is determined largely by resource availability in the environment, it has also been shown

to be genetically heritable in our study population [31], suggesting greater similarity among

closely related individuals also in other linked traits, such as trophic position. In addition,

maternal effects could shape individual specialization in brown bear offspring. As a potential

pathway, milk quantity or quality [32] can vary among females due to genetic differences

and/or differences in their home range quality, leading to consistently larger or smaller

offspring from the same mother, which in turn could cause similarities in trophic position

among siblings. Last, brown bears live a solitary lifestyle except for the period of offspring

rearing involving up to three years of maternal care [33], after which female offspring often

settle close to their mother's home range [34]. In their first years of life, bear cubs accompany

their mother, and so it is reasonable to predict that brown bear offspring learn their dietary

niche from their mothers. If mothers differ in dietary niches, these differences may be

maintained in the population through offspring social learning from the mother (hereafter

"social learning").

Individual trophic position is one metric to assess individual specialization along a

continuum from a more plant-based to a more meat- or insect-based diet. Trophic position

can be estimated from the ratio of stable-nitrogen isotopes ($\delta^{15}N$) in growing tissue and

reflects cumulative diet intake during the period of tissue growth. Individuals with higher

trophic positions are specialized on more protein-rich diets, relative to individuals with lower

trophic positions which are increasingly more herbivorous. Trophic position rarely provides

information on specific dietary items[2, 18] or individual variation in niche breadth[35] but rather

quantifies the consumption of animal matter relative to other individuals in a population of

omnivores. We analyzed annual trophic positions from $\delta^{15}N$ values in brown bear hair keratin

[36]. Hair $\delta^{15}N$ represents a dietary integration of about a month (i.e., a growing hair in June

reflects the diet intake since May [37]). Bear hair is annually renewed through molting in June,

regrows over the summer and fall and stops growing during winter hibernation (**Fig 1A, [38,**

39]). Guard hair samples collected in spring therefore reflect annual estimates of the

cumulative protein intake of individuals during the previous active foraging season [38].

Using repeated samples of known mother-daughter pairs, we first estimated the extent

of dietary specialization as permanent individual variation and second fitted a spatially

explicit Bayesian hierarchical model (i.e., ´animal model´ [7, 40, 41]) to quantify its sources.

Specifically, we accounted for environmental similarity, with pairwise habitat similarity in

individual bear home ranges encompassing the proportion of mature habitat, disturbed

habitat, and habitat diversity. We further accounted for genetic heritability with a pedigree,

for maternal effects by incorporating the mother's ID as a random effect, and for social

learning as the fixed effect of a mother's trophic position on her offspring's trophic position.

We allowed the effect of social learning to shift with time since offspring gained

independence to account for individual learning later in life. We determined maternal trophic

positions from a population-wide model accounting for sexual dimorphism, age, and

permanent between-individual variation in diet (**Supplement S3**). We validated that a

similarity between offspring and mother trophic position reflected social learning during

rearing, and found that their trophic positions were highly correlated when together in the

first year of life (**S4**). As we were interested in lifelong variation of dietary niche, and male

offspring were only monitored for a short period after family breakup, we focused on

individual specialization of female offspring. In the supplementary material we provide an

additional analysis of the relationship between maternal and male offspring trophic position

in the first four years after family breakup (**S8**) and of the relationship between paternal

trophic position and offspring trophic position (**S9**). We also provide an alternative analysis

accounting for spatial correlation via a spatial distance instead of an environmental similarity

matrix (**S6**), as well as a reduced model excluding the effect of environmental similarity to

test whether spatial and genetic effects were confounded in philopatric female bears (**S7**).We

validated our effect of social learning by refitting the model to a reduced dataset with

observed maternal trophic positions during rearing, instead of modeled-averaged maternal

trophic positions (**S10**).

2. RESULTS

We analyzed annual trophic positions from 213 hair samples collected from 71 female brown

bears born to 33 unique mothers (median 2 daughters; range 1-7 daughters per mother).

Repeated sampling (median 3 years; range 1-11 years) showed that female trophic position

was unaffected by age (median [mean, 89% equal tails credible interval] explained variance =

1% [1%, 0 – 4%]; see also **Fig S3A** and **Table S1**) and that individuals showed long-term

individual specialization, accounting for 48% [47%, 31 – 61%] of the total variance in

trophic position (**Fig 2,** *Basic model*). Individual variability in trophic position spanned half a

trophic position ranging from 2.7 to 3.1 for individual females (**Fig 1B**), which is equivalent

to the difference between an omnivore feeding on a mix of plants and animal prey and a

carnivore feeding predominantly on animal prey.

Individual specialization was primarily driven by social learning, environmental

similarity, and maternal effects. Maternal trophic position dynamic over time since separation

accounted for 13% [13%, 5% - 23%] of the total phenotypic variation in trophic position,

while environmental similarity accounted for 9% [15%, 0.1 - 49%]. Additionally, maternal

effects accounted for 11% [13%, 0.5% - 30%] of variation in trophic position, indicating that

siblings (full and half) of the same mother were more similar in trophic position throughout

life compared to non-siblings. A remaining 8% [10%, 0.3 – 26%] of variance in trophic

position was attributed to permanent between-individual effects (**Fig 2**). Genetically more

closely related individuals (including paternal half-siblings, aunts, uncles or cousins) did not

share a more similar trophic position (2% [5%, <0.1% – 17%] of variance explained),

providing no evidence that dietary specialization could be heritable in this population (**Fig 2,**

see **S11** for the full summary table).

After family breakup, female offspring initially maintained a similar trophic position

to their mother (Pearson's $r = 0.66$ in the first two years after separation), which gradually

became more dissimilar over time (Pearson's $r = 0.31$ in year 3 - 4 after separation, **Fig 3A**).

In the first years, offspring of more carnivorous mothers also had a high trophic position

while offspring of less carnivorous mothers had a lower trophic position. About five years

after the separation from the mother, this correlation ceased to exist. Additionally, daughters

of the same mother shared similar dietary niches, with siblings occupying consistently lower

or higher trophic positions (**Fig 3B**). Bears inhabiting home ranges with a similar

composition of mature and disturbed forest, as well as a similar habitat diversity in the home

range, also had more similar trophic positions. The distance between pairwise home range

centroids ranged from 0.7 to 172 km with a median pairwise distance of 48 km and

individuals living in closer proximity had a more similar trophic position than individuals

living farther apart (**S6**). However, spatial distance and maternal effects seemed to be

confounded in this female philopatric species (**S7**): after excluding spatial distance, social

learning and maternal effects, but not heritability, explained more variance in trophic

position, indicating that spatial proximity and maternal effects are confounded because

settlement home ranges of philopatric daughters are often close in space to their mothers

forming so called matrilineal assemblages. In a separate analysis (**S8**) we also showed that

the relationship between maternal and offspring trophic position in the first years after family

breakup was not sex-specific. Both male (n = 31, Pearson correlation coefficient = 0.4) and

female (n = 69, Pearson correlation coefficient = 0.45) offspring trophic positions resembled

their mother's trophic position in the first 4 years of independence, corroborating our findings

that initial social learning determines foraging behavior in the early years after family

breakup. Conversely, paternal trophic position had no effect on offspring trophic position in

the first four years of independence (Pearson correlation coefficient = 0.13, **S9**). While the

modelled maternal trophic position correlated strongly with the observed trophic position in

any given year (**S3, Fig S3B**), social learning explained even more of the phenotypic variance

in daughter trophic position (22% [8% - 27%] instead of 13 %, **S10**) when fitting the

observed maternal trophic position during rearing, instead of the modelled posterior average

maternal trophic position to a reduced dataset (62 hair samples collected from 38 daughters).

Our estimates of offspring social learning from the mother are therefore likely conservative

and may underestimate the true effect of social learning on individual specialization.

Figure 1. A) Bear hair generally grows from June until October and stable-nitrogen isotopes
(δ15N) reflects cumulative diet intake during the period of hair growth. The quiescent phase,
when hair ceases growing, lasts through hibernation, followed by emergence from the winter
den and molting in late May-early June. Hair samples were taken during bear captures in
April - June and reflect the bears' diet in the previous year; **B**) Posterior distribution of the
population trophic niche (bold line) and individual dietary niches indicated by each
individual's posterior trophic position (modelled distribution with individual posterior
medians indicated by black dots). Scientific illustration by Juliana D. Spahr, SciVisuals.com.

Figure 2. Individual specialization accounted for 48% of the phenotypic variation in trophic position of female brown bears in Central Sweden. Trophic position did not change with age. We determined the proportion of variance (mean of the posterior distribution) explained by different sources of individual specialization: Offspring social learning from the mother, environmental similarity, genetic heritability, maternal effects, permanent between-individual effects, and residual within-individual components.

Figure 3. A) Relationship between female brown bear trophic position and their mother's
trophic position over number of years since separation, which occurs at 1.5 – 2.5 years of age

in our population. The daughter's trophic position resembled their mothers' trophic position

in the first years after separation but this similarity ceased after 4 years. Lines indicate

predicted posterior mean estimates with ribbons corresponding to the estimated standard

error, raw data are shown as points. **B)** Additional maternal effects (e.g., maternal genotype

or maternal environment) explained further similarities in trophic position among daughters

of the same mother (and differences between daughters of different mothers).

3. DISCUSSION

Our multigenerational dataset reveals unique insights into the ontogeny of individual

dietary specialization along a continuum from a more herbivorous to a more carnivorous diet

in a long-lived omnivore. Specifically, the foraging strategy of offspring was intimately tied

to the foraging strategy of their mother, a relationship that lasted up to four years after

independence. We interpret this relationship as evidence that social learning plays an

important role in shaping an individual's dietary specialization. Five years into independence,

the similarity between the trophic position of mothers and daughters slowly faded, likely due

to individual learning and experience during solitary life. In addition, offspring of the same

mother also shared similarities in their trophic position, potentially mediated through

maternal genetic or environmental effects on body size [31]. Additive genetic effects on the

other hand were insignificant providing no evidence for heritability of dietary specialization

in this population. Similar to the effect of social learning from the mother fading over time,

additive genetic and maternal effects could be life-stage specific with maternal effects being

more influential in juveniles[13], although evidence for this is mixed [10]. We were not able to

quantify life-stage specific heritability and maternal effects due to sample size limitations. In

general, previous ecological studies have mainly concentrated on resource availability as the

main driver of resource selection [42] and individual specialization [4]. However, our results

show that, within populations, the environment is only one of several components shaping

individual variation in dietary niche. We conclude that social learning of maternal dietary

preferences during early-life ontogeny and maternal effects (i.e., maternal genotype and

environment), which together explained about 24% of the variation in trophic position, play a

pivotal role in spreading and maintaining feeding strategies within populations, even in

species with otherwise solitary lifestyles. In addition, variation linked to between-individual

effects (in our study 8 %) could be associated to either uncontrolled variation in resource

availability in the environment (i.e., ecological opportunity [4, 35]) or individual differences in

resource preference. The latter could, for example, be caused by differential individual

learning and demonstrates the potential for behavioral innovation in this population.

Ultimately, between-individual variation in dietary specialization allows populations to adapt

to changes in resource availability, such as, new invasive prey or declines in food items due

to climate change.

Our findings are particularly relevant for species in which dietary specialization impacts

individual fitness [20, 35, 43]. For example, protein-rich diets may promote greater offspring

survival or mass gain [44]. Social learning in general, therefore, presents an important, yet

understudied, pathway by which alternative behavioral strategies can establish and spread

more rapidly within populations than by genetic evolution alone [45]. Species more adept in

social learning of dietary strategies may therefore show greater behavioral variability at the

population level, which could give them an advantage when adapting to changing

environments due to landscape modification or urbanization, climatic variations or global

change in general. Moreover, there is evidence that the strength of social learning in shaping

individual phenotypes is not only species-specific, but can also vary among populations or

individuals of the same species [12, 46].

Our research also points to several aspects of social learning that warrant future

research. First, there is little information on whether maternal care and social learning tend to

be more prevalent in species or populations with greater dietary specialization. There is some

evidence that within populations, dietary generalists (i.e. those with a wider dietary niche)

seem to provide more intense parental care [47] than their conspecific dietary specialists (i.e.

ones with a narrower dietary niche), but the link to social learning of foraging preferences

remains unclear. Second, while generalist species with a wide ecological niche have been

frequently shown to be more successful under changing environmental conditions, such as

urban environments or fragmented landscapes, than specialist species [48, 49, 50], it is currently

unknown whether this success could be partially mediated by social learning. Finally, social

learning could alternatively limit behavioral innovation and adaptation due to adherence to

social traditions [51]. We therefore suggest that alternative hypotheses should be evaluated that

consider how social learning impacts individual specialization and in turn the adaptability of

species under global change.

Our findings that dietary specialization can be socially learned and transmitted are

particularly relevant for species where individual specialization is related to human-wildlife

conflict [52]. For example, the removal of single individuals which are known to cause conflict

is an effective strategy to halt the spread of problematic behavior and mitigate the conflict,

while minimizing the impact on species conservation goals [52]. Foraging behavior that causes

conflict with humans has also been shown to change in ursids over their lifetimes, remarking

the crucial role of individuality and plasticity in behavior [53]. Social learning from the mother

of behavior [54], including individual specialization and foraging on anthropogenic food

resources, has been previously observed in ursids [55, 56, 57, 58]. However, none of these studies

tracked offspring diet over their lifetimes or were able to simultaneously account for the

mother's diet, genetics, the environment, and other maternal effects that could explain similar

patterns of individual specialization. While some of the aforementioned studies suggest either

the environment or social learning as primary drivers of individual specialization, we suggest

using caution in assigning causality in dietary specialization, when potentially confounding

alternative sources cannot be accounted for. Specifically, in female-biased philopatric

species, spatial proximity does not only encode for spatial variation in resource abundance

but is also conflated with heritability and with maternal effects. In brown bears, some

daughters settle close to their mother's home range [34] creating spatial clusters of closely

related females, so called matrilinear assemblages [59]. Due to spatial dependence of these

assemblages, it can therefore be difficult to untangle social learning from the mother from

other maternal effects (i.e., maternal genotype or maternal environment) or the ambient

environment. Our study population spanned over 170 km with spatial proximity explaining

50% of the total phenotypic variation in trophic position of female bears: individuals further

apart tended to have more different diets. However, when replacing spatial proximity with

environmental similarity among home ranges, the explanatory power was attributed to social

learning and maternal effects along with the environment. Our results therefore demonstrate

that individual dietary specialization is not caused by a single driver in isolation but the

product of many factors, namely social learning, maternal effects, and the environment.

Our finding that social learning has a similar impact on resource selection as the

environment provides important insights for a range of studies on habitat selection, dispersal,

and range expansion. For example, a popular theory known as "natal habitat preference

induction" suggests that dispersing animals select areas for settlement that resemble their

natal habitat, even at fine habitat scales [30]. Our results challenge the notion that habitat

similarity alone drives natal settlement strategies and rather suggest that socially learned diet

preferences, and hence the selection for food resources themselves, could play an important

role in producing similar patterns of settlement selection like induced natal habitat

preferences. Recent studies of migration and short stopover behavior in whooping cranes

(*Grus americana*) have also observed that social learning rather than environmental

conditions [60] or genetic heritability [61] led to the emergence and establishment of alternative

migratory behavior. Similar to what our study shows with respect to individual specialization,

social learning of migration strategies primarily determined behavior in early life whereas

individual-experiential learning shaped behavior later in life [62].

Conclusion

Drivers of individual dietary specialization are well documented among populations of the

same species. However, systematic studies delineating the sources of individual

specialization within populations are lacking, likely because suitable datasets including

multigenerational, genetic, environmental, and life-history information are rare. We show

that, in addition to the environment, social learning and maternal effects can be important

sources of dietary specialization.

4. METHODS

Bear hair sample collection

We collected brown bear hair samples in south-central Sweden (~N61°, E15°) as part of a

long-term, individual-based monitoring project (Scandinavian Brown Bear Research Project;

www.bearproject.info). Hair samples were collected from known individuals and their

offspring during captures in spring (April - June) 1993 – 2016. Bears were immobilized from

a helicopter (Arnemo & Fahlman, 2011). A vestigial premolar tooth was collected from all

bears not captured as a yearling to estimate age based on the cementum annuli in the root [63].

Tissue samples (stored in 95% alcohol) were taken for DNA extraction to assign parentage

and construct a genetic pedigree [59]. Guard hairs with follicles were plucked with pliers from a

standardized spot between the shoulder blades and archived at the Swedish National

Veterinary Institute. All animal captures and handling were performed in accordance with

relevant guidelines and regulations and were approved by the Swedish authorities and ethical

committee (Uppsala Djurförsöksetiska Nämnd: C40/3, C212/9, C47/9, C210/10, C7/12,

C268/12, C18/ 15. Statens Veterinärmediciniska Anstalt, Jordbruksverket, Naturvårdsverket:

Dnr 35-846/03, Dnr 412-7093-08 NV, Dnr 412-7327-09 Nv, Dnr 31-11102/12, NV-01758-

14). We used data of offspring after separation from their mother (main text) and of adult

male and female bears (**Supplementary analysis S3**). Bear cubs are born in January or

February during winter hibernation and are typically first captured together with their mother

as yearlings at the age of ~ 15 months. Cubs in this population separate from their mother

during the mating season in May or June after 1.5 or 2.5 years [64]. A hair sample taken in

spring reflects the summer-fall diet of the bear in the previous active season (**Fig 1A**). Only

hair samples of solitary, independent offspring taken in spring at least 10 months after

separation from the mother were included in this study.

Moose sample collection

We collected samples of the natural foods most important for brown bear in the study area

(**Fig S1**), including 21 samples of moose hair (*Alces alces*), the most common meat source in

the diet of brown bears in our study area [65] in the spring-autumn field season of 2014.

Samples were placed in a paper envelope and dried at ambient temperature.

Stable isotope analyses

Hair samples were rinsed with a 2:1 mixture of chloroform:methanol or washed with pure

methanol to remove surface oils [66]. Dried samples were ground with a ball grinder (Retsch

model MM-301, Haan, Germany). We weighed 1 mg of ground hair into pre-combusted tin

capsules and combusted at 1030°C in a Carlo Erba NA1500 elemental analyser. N_2 and CO_2

were separated chromatographically and introduced to an Elementar Isoprime isotope ratio

mass spectrometer (Langenselbold, Germany). Two reference materials were used to

normalize the results to VPDB and AIR for δ^{13}C and δ^{15}N measurement, respectively: BWB

III keratin (δ^{13}C =- 20.18‰, δ^{15}N = 14.31‰, respectively) and PRC gel (δ^{13}C =-13.64‰,

δ^{15}N = 5.07‰, respectively). Measurement precisions as determined from both reference and

sample duplicate analyses were ± 0.1‰ for both δ^{13}C and δ^{15}N.

Bear trophic position

We calculated the trophic position of each bear hair sample relative to the average δ^{15}N value

of moose hair representing trophic level 2 (mean ± sd = 1.8 ± 1.26 ‰, n = 21, **Fig S1**). Bears

consume most of a moose carcass, including meat, skin, and hair. Soft tissue samples of

moose carcasses could not be obtained but according to the literature the ratio of δ^{15}N in

ungulate hair to meat ranges between 0.77‰ – 1.0‰. (see S3.1 in [67]). We consider trophic

positions calculated from moose hair representative and a correction of the δ^{15}N moose hair

signature would only add an arithmetic correction but not change the distribution of bear

trophic positions. Trophic position is calculated as the discrepancy of δ^{15}N in a secondary

consumer and its food source divided by the enrichment of δ^{15}N per trophic level, plus

lambda, the trophic position of the food source (e.g. 1 for primary producers, 2 for primary

consumers, 3 for secondary consumer, 4 for tertiary consumers) [68]. We used an average

trophic enrichment factor of 3.4‰ [68] and added a lambda of 2 given the moose baseline

trophic position as a strict herbivore.

429   Bear trophic position = (δ^{15}N$_{Ursus\ arctos}$ – average(δ^{15}N$_{Alces\ alces.hair}$)) / 3.4 + 2

430   Under an omnivorous diet including the consumption of herbivores (in particular moose but

431   also ants such as *Formica* spp*., Camponotus herculeanus* with average δ^{15}N indistinguishable

432   from moose, **Fig S1**), bear trophic position values were expected to fall between 2 and 3.

433   Values approaching 4 indicate a trophic enrichment through consumption of other

434   omnivorous or carnivorous animals.

Sources of individual variation in trophic position

Environmental similarity

Resources may not be distributed evenly in space. For moose, population density and hunting

quotas (which determine availability of slaughter remains) vary across the study area. For

ants, the availability of old forests and clearcuts determine their abundance [69]. Furthermore,

brown bear daughters are often philopatric with limited dispersal and settle close to their

mother's home range [34]. The median dispersal distance of daughters, namely the distance

between natal and settlement home range centroids in this study was 8.56 km (range 1.4 –

28.8 km). Genetic, spatial, and social learning effects may therefore be confounded with

related bears occupying adjacent ranges with similar resource availability. Elsewhere,

accounting for environmental similarity through spatial autocorrelation in animal models has

revealed that a major portion of variance may be attributed to environmental similarity rather

than genetic heritability [7, 40, 70,] but see also [71]. Here, we accounted for environmental

similarity by extracting habitat composition in each bear's lifetime home range (n = 71). We

constructed home ranges using a 95% kernel density estimator (kernelUD function) in the R

package adehabitatHR [72] using the reference bandwidth as smoothing parameter (option

"href"). Bears were monitored for a minimum of 1000 GPS locations (n = 47) or were located

via VHF on at least 25 days (n = 24). Median lifetime home range size was 256 km^2, which is

comparable to a circle with an 18 km diameter. We used a Corine landcover map (25 m

resolution) which we updated annually with polygons of newly emerged clearcuts (data

obtained from the Swedish Forest Agency). We extracted home range habitat composition in

the year when diet was assessed. When individuals were monitored for multiple years, we

extracted the home range composition for the median year. Annual changes in home range

habitat composition were negligible (**Fig S5**). We calculated the proportion of mid-aged and

old forest and proportion of disturbed forest (clearcuts and regenerating young forest) within

the 95% utilization distribution. Additionally, we calculated habitat diversity using the

Simpson diversity index from the R package landscapemetrics [73]. Following Thomson et al.

[40] we calculated the Euclidean distance between scaled and centered habitat composition and

habitat diversity in multivariate space, assuming equal importance of each component.

Pairwise distances were scaled between 0 and 1, where increasing values indicated more

similar habitat composition. In the supplementary material we provide an alternative analysis

accounting for spatial autocorrelation of individual dietary niches with a pairwise spatial

distance matrix (S matrix; **Supplementary analysis 6**, **Fig S6**). Spatial autocorrelation of

home range habitat composition seized after 10 – 15 km (**Fig S6**).

Genetic pedigree

A genetic pedigree based on 16 microsatellite loci was available for the population including

1614 individual genotypes, spanning six generation [59]. All female offspring and mothers in

this study were genotyped and included in the population's genetic pedigree. We used Cervus

3.0 [74] for assignment of fathers and COLONY [75] for creating putative unknown mother or

father genotypes and sibship reconstruction (see for details [59]).

Maternal identity

The study was based on a population of marked females and their offspring. Therefore, all

mothers included in this study were known from observations of mother-offspring

associations.

Maternal trophic position

Based on repeated hair samples of 115 female (n_{female} = 335) and 98 male (n_{male} = 219) bears,

we fitted a linear mixed effects model for female and male bears respectively, to estimate

sex-specific among individual variation in trophic position (**Supplementary analysis 3**). We

modelled trophic position as a function of a quadratic relationship with age and we controlled

for individual random intercepts. We concentrate on the relationship between age and trophic

position as age is not confounded with among-individual effects, however we also perused

the relationship between mass and trophic position in the **Supplementary analysis 2, Fig S2.**

Female trophic position did not vary with age but was highly repeatable over multiple years.

For all daughters, we extracted their mother's (and father's) trophic position as the median of

the posterior distribution of their respective random intercept. The modelled posterior trophic

position and the observed trophic position in a given sampling year were strongly positively

correlated (Pearson correlation coefficient r = 0.78, t = 22.63, df = 336, p < 0.001).

Statistical analysis

We applied a two-step modelling approach. First, we fitted a *basic* linear mixed-effects

model to estimate individual dietary specialization as permanent between-individual variation

(V_I) in trophic position. For this we used repeated measures of the same individual and fitted

individual random intercepts. We accounted for a nonlinear effect of age (second-order

polynomial, scaled by the standard deviation). We extracted the variance in fitted values

(V_{Age}; variance explained by age), permanent between-individual (V_I), and residual within-

individual variance (V_R) and estimated each component's proportional contribution to the

total phenotypic variance ($V_P = V_{Age} + V_I + V_R$) through variance standardization [76].

Second, we used a spatially explicit Bayesian hierarchical model (i.e. 'animal model') [5, 40, 41]

to partition permanent between-individual variance (V_I) in trophic position into its sources;

the fixed effects age (V_{Age}) and social learning (V_{SL}), and the variance components permanent

between-individual variance (V_I), environmental similarity (V_E), additive genetic variance

(V_A), maternal effects (V_M), and residual within-individual variance (V_R). We followed the

"hybrid" strategy suggested by McAdam, Garant [14] and tested for social learning of trophic

position from the mother ("social learning") by incorporating maternal trophic position as a

fixed effect into the model. Thus, V_M pools the remaining phenotypic variation of offspring

trophic position that cannot be explained by maternal trophic position. We accounted for a

nonlinear effect of time since separation of mother and daughter (scaled by the standard

deviation), and for an interaction between maternal trophic position and time since separation

to account for a decrease of the social learning effect over time. Age and time since

separation were perfectly correlated: Pearson correlation coefficient > 0.99.

We calculated the total variance explained by the model ($V_P = V_{Age} + V_{SL} + V_I + V_E +$

$V_A + V_M + V_R$) and calculated the proportion of the total variance explained by each model

component. For the fixed effects, we partitioned the variance explained into social learning

over time (i.e. maternal trophic position and its interaction with time since separation) and

age (i.e. the main effect of time since separation), respectively, by calculating the

independent contribution of each component to the total variance explained by the fixed

effects, following the approach by Stoffel, Nakagawa [77] adapted to a Bayesian framework

(see code under [78]). For all parameters, we report the median and mean as measures of

centrality and 89% credible intervals, calculated as equal tail intervals, as measure of

uncertainty [79, 80]. We deemed explained variance proportions as inconclusive when the lower

credible interval limit was < 0.001 (i.e., $< 0.1\%$) [81].

All models were fitted using the R package "brms" [82] based on the Bayesian software

Stan [83, 84]. We ran four chains to evaluate convergence which were run for 6,000 iterations,

with a warmup of 3,000 iterations and a thinning interval of 10. All estimated model

coefficients and credible intervals were therefore based on 1200 posterior samples and had

satisfactory convergence diagnostics with $\hat{R} < 1.01$, and effective sample sizes > 400 [85].

Posterior predictive checks recreated the underlying Gaussian distribution of trophic position

well. All statistical analyses were performed in R 4.0.0 [86]. Primary data and code to

reproduce all analyses are provided under (https://doi.org/10.17605/OSF.IO/68B9U, [78]).

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AUTHOR'S CONTRIUTIONS

AH, JA, and TM developed the work. AZ, JK, KAH, NS and AS provided the data. AZ

managed the sample collection. AS managed the hair samples database and prepared samples

for stable isotope analyses by KAH. DJ provided laboratory space and resources and

supervised preparatory procedures. SF constructed and provided the genetic pedigree. JH

provided home range centroids. AM and JA advised to the analysis and interpretation of

stable isotope data. TM, NS and AZ secure project funding. AH performed the statistical

analyses with input from JA. AH wrote the manuscript with help from TM, JA, and input

from all authors.

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