

Lab 7: Networks

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Computational Topics

- Build and visualize food webs
- Write functions to implement mathematical equations

Conservation topics

- Paleofood webs
- Species extinction

In this lab we will practice our network visualization and manipulation skills using the paleo food web data from Yeakel et al. 2014 (<https://doi.org/10.1073/pnas.1408471111>).



Paleoweb

See the beautiful, animated version of the graphic above here
(<https://infograficos.estadao.com.br/public/cidades/extincoes-egito/>)

With some interaction networks we can observe the interactions, for example plant-pollinator networks, seed-dispersal networks, human social networks. In food webs sometimes feeding interactions are observed directly, through camera traps, people doing timed observations, and now molecular analysis of gut contents/scat. However, often with food webs people build probabilistic models of who interacts with who based on body size (as in the Yeakel et al. 2014), especially with paleowebs. Thus the data from Yeakel et al. is 1) an occurrence matrix

(Figure 2 from the publication) and a matrix of body sizes (two columns, females then males). We will use these data to build the foodwebs for each time period. This lab is pretty challenging because it will use many of our core programming skills (for loops, writing functions, subsetting data) and our network skills.

First we will read in the data. The matrix we are reading in has no row or column names, we will have to set them.

```
sp_occ <- read.table(file="data/egypt_data.txt", header = FALSE)
str(sp_occ)
```

```
## 'data.frame': 39 obs. of 23 variables:
## $ V1 : int 1 1 1 1 1 1 1 1 0 1 ...
## $ V2 : int 1 1 1 1 1 1 1 1 0 1 ...
## $ V3 : int 1 1 1 1 0 1 1 1 0 1 ...
## $ V4 : int 1 1 1 1 0 1 1 1 0 1 ...
## $ V5 : int 1 1 1 1 0 1 1 1 0 1 ...
## $ V6 : int 1 1 1 1 0 1 1 1 0 1 ...
## $ V7 : int 1 1 1 1 0 1 1 1 0 1 ...
## $ V8 : int 1 1 1 1 0 0 1 1 0 1 ...
## $ V9 : int 1 1 1 1 0 0 1 1 0 0 ...
## $ V10: int 1 1 1 1 0 0 1 1 0 0 ...
## $ V11: int 1 1 1 1 0 0 1 1 0 0 ...
## $ V12: int 1 1 1 1 0 0 1 1 1 0 ...
## $ V13: int 1 1 1 1 0 0 1 1 1 0 ...
## $ V14: int 1 1 0 1 0 0 1 1 1 0 ...
## $ V15: int 1 1 0 1 0 0 1 1 1 0 ...
## $ V16: int 1 1 0 1 0 0 1 1 1 0 ...
## $ V17: int 1 1 0 1 0 0 1 1 1 0 ...
## $ V18: int 1 1 0 1 0 0 0 1 1 0 ...
## $ V19: int 1 1 0 1 0 0 0 1 1 0 ...
## $ V20: int 1 1 0 1 0 0 0 1 1 0 ...
## $ V21: int 1 1 0 1 0 0 0 1 0 0 ...
## $ V22: int 1 1 0 1 0 0 0 1 0 0 ...
## $ V23: int 1 1 0 1 0 0 0 0 0 0 ...
```

```
sp_mass <- read.table(file="data/egypt_mass.txt", header=FALSE)
str(sp_mass)
```

```
## 'data.frame': 39 obs. of 2 variables:
## $ V1: int 6 4 18 25 40 122 122 50 35 2200 ...
## $ V2: int 15 8 36 55 90 260 260 60 65 6300 ...
```

```
colnames(sp_mass) <- c("f","m")
```

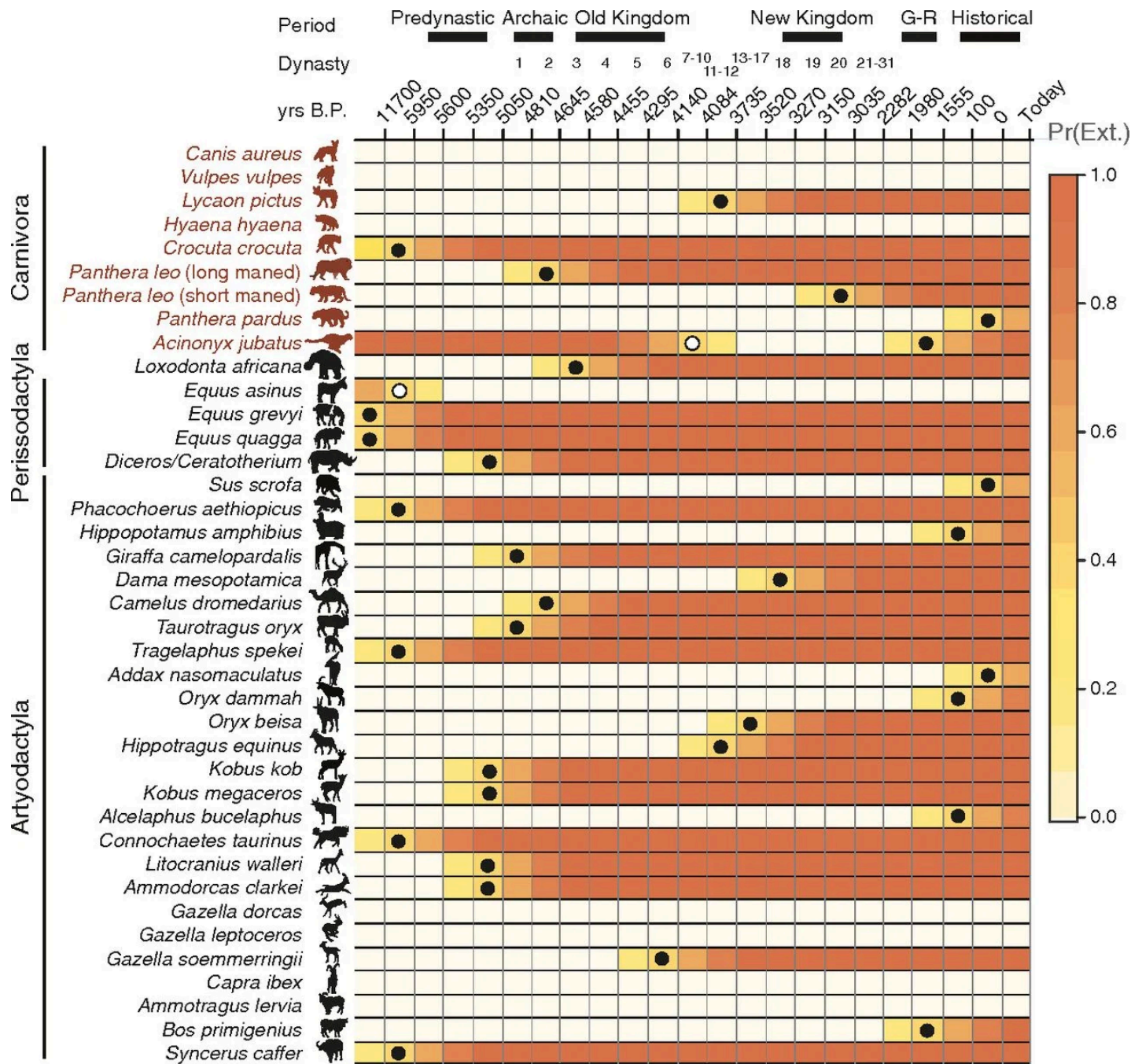


Figure 2

The rows are arranged in the order of Figure 2 of the manuscript. To set the rownames we can make a vector of the names then use the function 'rownames'. We also have to note which species are predators (all those in the species in the Carnivora clade in figure 2). Otherwise we will create a web where giraffes are voracious predators consuming all of the other species (I made this mistake when constructing the networks originally). I have transcribed the data from figure 2 for you:

```

row_labs_sp <- c("Canis aureus", "Vulpes vulpes", "Lycaon pictus", "Hyaena hyaena", "Crocuta crocuta", "Panthera leo (long maned)", "Panthera leo (short maned)", "Panthera pardus", "Acinonyx jubatus", "Loxodonta africana", "Equus asinus", "Equus grevyi", "Equus quagga", "Diceros/Ceratotherium", "Sus scrofa", "Phacochoerus aethiopicus", "Hippopotamus amphibius", "Giraffa camelopardalis", "Dama mesopotamica", "Camelus dromedarius", "Taurotragus oryx", "Tragelaphus spekei", "Adax nasomaculatus", "Oryx dammah", "Oryx beisa", "Hippotragus equinus", "Kobus kob", "Kobus megaceros", "Alcelaphus bucelaphus", "Connochaetes taurinus", "Litocranius walleri", "Ammodorcas clarki", "Gazella dorcas", "Gazella leptoceros", "Gazella soemmerringii", "Capra ibex", "Ammotragus lervia", "Bos primigenius", "Syncerus caffer")

rownames(sp_occ) <- row_labs_sp
rownames(sp_mass) <- row_labs_sp

## Set 1 for predators, 0 for prey
carnivores <- c(rep(1, 9), rep(0, length(row_labs_sp)- 9))
names(carnivores) <- row_labs_sp
sp_occ <- cbind(sp_occ, carnivores)
sp_mass <- cbind(sp_mass, carnivores)
#sp_occ$Species <- rownames(sp_occ)
sp_mass$Species <- rownames(sp_mass)

```

Lab part 1: Creating our foodwebs based on body sizes.

- Use the above vector of species names to label the row names of the species occurrence and the body size matrices. The columns of the species occurrence matrix are time points, so we can leave those as V1 etc., but we should set the column names of the mass matrix as “f”, “m” (female and male). Use ‘head’ to check each matrix to see if the names are displayed properly.

```
head(sp_mass)
```

##	f	m	carnivores	Species
## Canis aureus	6	15	1	Canis aureus
## Vulpes vulpes	4	8	1	Vulpes vulpes
## Lycaon pictus	18	36	1	Lycaon pictus
## Hyaena hyaena	25	55	1	Hyaena hyaena
## Crocuta crocuta	40	90	1	Crocuta crocuta
## Panthera leo (long maned)	122	260	1	Panthera leo (long maned)

```
head(sp_occ)
```

##	V1	V2	V3	V4	V5	V6	V7	V8	V9	V10	V11	V12	V13	V14	V15
## Canis aureus	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
## Vulpes vulpes	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
## Lycaon pictus	1	1	1	1	1	1	1	1	1	1	1	1	1	0	0
## Hyaena hyaena	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
## Crocuta crocuta	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0
## Panthera leo (long maned)	1	1	1	1	1	1	1	0	0	0	0	0	0	0	0
##	V16	V17	V18	V19	V20	V21	V22	V23	carnivores						
## Canis aureus	1	1	1	1	1	1	1	1					1		
## Vulpes vulpes	1	1	1	1	1	1	1	1					1		
## Lycaon pictus	0	0	0	0	0	0	0	0					1		
## Hyaena hyaena	1	1	1	1	1	1	1	1					1		
## Crocuta crocuta	0	0	0	0	0	0	0	0					1		
## Panthera leo (long maned)	0	0	0	0	0	0	0	0					1		

Yeakel recommended an updated equation to estimate the probability a predator consumed a prey based on their relative body masses from Rohr et al. 2010. (<https://doi.org/10.1086/653667>). The probability of existence of a trophic link between a predator of body-size m_i and a prey of body-size m_j is given by:

$$\text{logit}\left(P(A_{ij} = 1)\right) = \alpha + \beta \log\left(\frac{m_j}{m_i}\right) + \gamma \log^2\left(\frac{m_j}{m_i}\right). \quad (P(A_{1j} = 1))$$

is the probability predator i eats prey j).

- Write a function and call it 'probEat' to implement the equation above. Round the probability to two decimal places.

Below are the values of alpha, beta, and gamma for the Serengeti. In addition, you will need a function to compute the inverse logit function because this equation is for the logit of the probability, so to calculate the 0-1 probability you will need to take the inverse logit of the other side of the equation. Also note, \log^2 is equivalent to $(\log(m_i/m_j))^2$

```

# Define probEat function
probEat <- function(alpha, beta, gamma, m_j, m_i) {
  log_mass_ratio <- log(m_j / m_i)
  square_log_mass_ratio <- log_mass_ratio^2
  logit_prob <- alpha + beta * log_mass_ratio + gamma * square_log_mass_ratio
  probability <- exp(logit_prob) / (1 + exp(logit_prob))
  return(probability)
}

# Constants
alpha <- 2.51
beta <- 0.79
gamma <- -0.37

# Filter predators and prey
predators <- sp_mass[sp_mass$carnivore == 1, ]
prey <- sp_mass[sp_mass$carnivore == 0, ]

# Initialize results as a data frame
probability_results <- data.frame(predator = character(),
                                   prey = character(),
                                   probability = numeric(),
                                   stringsAsFactors = FALSE)

# Nested loop for probabilities
for (i in 1:nrow(predators)) {
  for (j in 1:nrow(pre)) {
    m_j <- predators$f[i]
    m_i <- prey$f[j]
    probability <- probEat(alpha, beta, gamma, m_j, m_i)

    # Append to results
    probability_results <- rbind(probability_results, data.frame(
      predator = predators$Species[i],
      prey = prey$Species[j],
      probability = probability
    ))
  }
}

# Print results
print(probability_results)

```

##	predator	prey	probability
## 1	Canis aureus	Loxodonta africana	2.899179e-07
## 2	Canis aureus	Equus asinus	2.846500e-03
## 3	Canis aureus	Equus grevyi	1.090779e-03
## 4	Canis aureus	Equus quagga	1.256377e-02
## 5	Canis aureus	Diceros/Ceratotherium	6.567126e-05
## 6	Canis aureus	Sus scrofa	5.695797e-01
## 7	Canis aureus	Phacochoerus aethiopicus	3.580329e-01
## 8	Canis aureus	Hippopotamus amphibius	2.478497e-04
## 9	Canis aureus	Giraffa camelopardalis	4.104377e-04
## 10	Canis aureus	Dama mesopotamica	8.764233e-02
## 11	Canis aureus	Camelus dromedarius	6.527585e-04
## 12	Canis aureus	Taurotragus oryx	1.940340e-03
## 13	Canis aureus	Tragelaphus spekei	4.205835e-01
## 14	Canis aureus	Addax nasomaculatus	2.191298e-01
## 15	Canis aureus	Oryx dammah	2.829276e-02
## 16	Canis aureus	Oryx beisa	4.412294e-02
## 17	Canis aureus	Hippotragus equinus	5.582491e-03
## 18	Canis aureus	Kobus kob	2.191298e-01
## 19	Canis aureus	Kobus megaceros	2.191298e-01
## 20	Canis aureus	Alcelaphus bucelaphus	4.412294e-02
## 21	Canis aureus	Connochaetes taurinus	2.534605e-02
## 22	Canis aureus	Litocranius walleri	5.695797e-01
## 23	Canis aureus	Ammodorcas clarkei	7.024425e-01
## 24	Canis aureus	Gazella dorcas	8.138893e-01
## 25	Canis aureus	Gazella leptoceros	8.284962e-01
## 26	Canis aureus	Gazella soemmerringii	4.479836e-01
## 27	Canis aureus	Capra ibex	3.039941e-01
## 28	Canis aureus	Ammotragus lervia	4.205835e-01
## 29	Canis aureus	Bos primigenius	1.345967e-05
## 30	Canis aureus	Syncerus caffer	3.745629e-03
## 31	Vulpes vulpes	Loxodonta africana	3.367834e-08
## 32	Vulpes vulpes	Equus asinus	6.218884e-04
## 33	Vulpes vulpes	Equus grevyi	2.201570e-04
## 34	Vulpes vulpes	Equus quagga	3.149058e-03
## 35	Vulpes vulpes	Diceros/Ceratotherium	1.075707e-05
## 36	Vulpes vulpes	Sus scrofa	3.580329e-01
## 37	Vulpes vulpes	Phacochoerus aethiopicus	1.722708e-01
## 38	Vulpes vulpes	Hippopotamus amphibius	4.465151e-05
## 39	Vulpes vulpes	Giraffa camelopardalis	7.678233e-05
## 40	Vulpes vulpes	Dama mesopotamica	2.829276e-02
## 41	Vulpes vulpes	Camelus dromedarius	1.265314e-04
## 42	Vulpes vulpes	Taurotragus oryx	4.104377e-04
## 43	Vulpes vulpes	Tragelaphus spekei	2.191298e-01
## 44	Vulpes vulpes	Addax nasomaculatus	8.764233e-02
## 45	Vulpes vulpes	Oryx dammah	7.753809e-03
## 46	Vulpes vulpes	Oryx beisa	1.279939e-02
## 47	Vulpes vulpes	Hippotragus equinus	1.294350e-03
## 48	Vulpes vulpes	Kobus kob	8.764233e-02
## 49	Vulpes vulpes	Kobus megaceros	8.764233e-02
## 50	Vulpes vulpes	Alcelaphus bucelaphus	1.279939e-02
## 51	Vulpes vulpes	Connochaetes taurinus	6.856283e-03

## 52	Vulpes vulpes	Litocranius walleri	3.580329e-01
## 53	Vulpes vulpes	Ammodorcas clarkei	5.219777e-01
## 54	Vulpes vulpes	Gazella dorcas	6.941108e-01
## 55	Vulpes vulpes	Gazella leptoceros	7.190257e-01
## 56	Vulpes vulpes	Gazella soemmerringii	2.416239e-01
## 57	Vulpes vulpes	Capra ibex	1.363823e-01
## 58	Vulpes vulpes	Ammotragus lervia	2.191298e-01
## 59	Vulpes vulpes	Bos primigenius	1.980874e-06
## 60	Vulpes vulpes	Syncerus caffer	8.380164e-04
## 61	Lycaon pictus	Loxodonta africana	5.369303e-05
## 62	Lycaon pictus	Equus asinus	8.764233e-02
## 63	Lycaon pictus	Equus grevyi	4.340734e-02
## 64	Lycaon pictus	Equus quagga	2.313361e-01
## 65	Lycaon pictus	Diceros/Ceratotherium	4.771788e-03
## 66	Lycaon pictus	Sus scrofa	8.818325e-01
## 67	Lycaon pictus	Phacochoerus aethiopicus	8.138893e-01
## 68	Lycaon pictus	Hippopotamus amphibius	1.379781e-02
## 69	Lycaon pictus	Giraffa camelopardalis	2.050152e-02
## 70	Lycaon pictus	Dama mesopotamica	5.695797e-01
## 71	Lycaon pictus	Camelus dromedarius	2.936718e-02
## 72	Lycaon pictus	Taurotragus oryx	6.653049e-02
## 73	Lycaon pictus	Tragelaphus spekei	8.379782e-01
## 74	Lycaon pictus	Addax nasomaculatus	7.354647e-01
## 75	Lycaon pictus	Oryx dammah	3.580329e-01
## 76	Lycaon pictus	Oryx beisa	4.387028e-01
## 77	Lycaon pictus	Hippotragus equinus	1.391993e-01
## 78	Lycaon pictus	Kobus kob	7.354647e-01
## 79	Lycaon pictus	Kobus megaceros	7.354647e-01
## 80	Lycaon pictus	Alcelaphus bucelaphus	4.387028e-01
## 81	Lycaon pictus	Connochaetes taurinus	3.390889e-01
## 82	Lycaon pictus	Litocranius walleri	8.818325e-01
## 83	Lycaon pictus	Ammodorcas clarkei	9.118609e-01
## 84	Lycaon pictus	Gazella dorcas	9.334997e-01
## 85	Lycaon pictus	Gazella leptoceros	9.361458e-01
## 86	Lycaon pictus	Gazella soemmerringii	8.472404e-01
## 87	Lycaon pictus	Capra ibex	7.886322e-01
## 88	Lycaon pictus	Ammotragus lervia	8.379782e-01
## 89	Lycaon pictus	Bos primigenius	1.311463e-03
## 90	Lycaon pictus	Syncerus caffer	1.062213e-01
## 91	Hyaena hyaena	Loxodonta africana	2.150731e-04
## 92	Hyaena hyaena	Equus asinus	1.877236e-01
## 93	Hyaena hyaena	Equus grevyi	1.041663e-01
## 94	Hyaena hyaena	Equus quagga	3.945327e-01
## 95	Hyaena hyaena	Diceros/Ceratotherium	1.433279e-02
## 96	Hyaena hyaena	Sus scrofa	9.132252e-01
## 97	Hyaena hyaena	Phacochoerus aethiopicus	8.718947e-01
## 98	Hyaena hyaena	Hippopotamus amphibius	3.780232e-02
## 99	Hyaena hyaena	Giraffa camelopardalis	5.393836e-02
## 100	Hyaena hyaena	Dama mesopotamica	7.090902e-01
## 101	Hyaena hyaena	Camelus dromedarius	7.414967e-02
## 102	Hyaena hyaena	Taurotragus oryx	1.496005e-01
## 103	Hyaena hyaena	Tragelaphus spekei	8.866493e-01

## 104	Hyaena hyaena	Addax nasomaculatus	8.227068e-01
## 105	Hyaena hyaena	Oryx dammah	5.313305e-01
## 106	Hyaena hyaena	Oryx beisa	6.049412e-01
## 107	Hyaena hyaena	Hippotragus equinus	2.708034e-01
## 108	Hyaena hyaena	Kobus kob	8.227068e-01
## 109	Hyaena hyaena	Kobus megaceros	8.227068e-01
## 110	Hyaena hyaena	Alcelaphus bucelaphus	6.049412e-01
## 111	Hyaena hyaena	Connochaetes taurinus	5.127178e-01
## 112	Hyaena hyaena	Litocranius walleri	9.132252e-01
## 113	Hyaena hyaena	Ammodorcas clarkei	9.311790e-01
## 114	Hyaena hyaena	Gazella dorcas	9.435878e-01
## 115	Hyaena hyaena	Gazella leptoceros	9.449904e-01
## 116	Hyaena hyaena	Gazella soemmerringii	8.922902e-01
## 117	Hyaena hyaena	Capra ibex	8.562696e-01
## 118	Hyaena hyaena	Ammotragus lervia	8.866493e-01
## 119	Hyaena hyaena	Bos primigenius	4.324553e-03
## 120	Hyaena hyaena	Syncerus caffer	2.191298e-01
## 121	Crocota crocuta	Loxodonta africana	1.361890e-03
## 122	Crocota crocuta	Equus asinus	4.139429e-01
## 123	Crocota crocuta	Equus grevyi	2.800273e-01
## 124	Crocota crocuta	Equus quagga	6.313617e-01
## 125	Crocota crocuta	Diceros/Ceratotherium	5.829090e-02
## 126	Crocota crocuta	Sus scrofa	9.374177e-01
## 127	Crocota crocuta	Phacochoerus aethiopicus	9.177241e-01
## 128	Crocota crocuta	Hippopotamus amphibius	1.302818e-01
## 129	Crocota crocuta	Giraffa camelopardalis	1.722708e-01
## 130	Crocota crocuta	Dama mesopotamica	8.356278e-01
## 131	Crocota crocuta	Camelus dromedarius	2.191298e-01
## 132	Crocota crocuta	Taurotragus oryx	3.580329e-01
## 133	Crocota crocuta	Tragelaphus spekei	9.248399e-01
## 134	Crocota crocuta	Addax nasomaculatus	8.936762e-01
## 135	Crocota crocuta	Oryx dammah	7.313709e-01
## 136	Crocota crocuta	Oryx beisa	7.772035e-01
## 137	Crocota crocuta	Hippotragus equinus	5.150239e-01
## 138	Crocota crocuta	Kobus kob	8.936762e-01
## 139	Crocota crocuta	Kobus megaceros	8.936762e-01
## 140	Crocota crocuta	Alcelaphus bucelaphus	7.772035e-01
## 141	Crocota crocuta	Connochaetes taurinus	7.190257e-01
## 142	Crocota crocuta	Litocranius walleri	9.374177e-01
## 143	Crocota crocuta	Ammodorcas clarkei	9.453213e-01
## 144	Crocota crocuta	Gazella dorcas	9.492570e-01
## 145	Crocota crocuta	Gazella leptoceros	9.493853e-01
## 146	Crocota crocuta	Gazella soemmerringii	9.275432e-01
## 147	Crocota crocuta	Capra ibex	9.101357e-01
## 148	Crocota crocuta	Ammotragus lervia	9.248399e-01
## 149	Crocota crocuta	Bos primigenius	2.050152e-02
## 150	Crocota crocuta	Syncerus caffer	4.550391e-01
## 151	Panthera leo (long maned)	Loxodonta africana	5.366696e-02
## 152	Panthera leo (long maned)	Equus asinus	8.387459e-01
## 153	Panthera leo (long maned)	Equus grevyi	7.801326e-01
## 154	Panthera leo (long maned)	Equus quagga	8.981503e-01
## 155	Panthera leo (long maned)	Diceros/Ceratotherium	5.001258e-01

## 156	Panthera leo (long maned)	Sus scrofa	9.473550e-01
## 157	Panthera leo (long maned)	Phacochoerus aethiopicus	9.493032e-01
## 158	Panthera leo (long maned)	Hippopotamus amphibius	6.508930e-01
## 159	Panthera leo (long maned)	Giraffa camelopardalis	7.002593e-01
## 160	Panthera leo (long maned)	Dama mesopotamica	9.379692e-01
## 161	Panthera leo (long maned)	Camelus dromedarius	7.408155e-01
## 162	Panthera leo (long maned)	Taurotragus oryx	8.175219e-01
## 163	Panthera leo (long maned)	Tragelaphus spekei	9.493506e-01
## 164	Panthera leo (long maned)	Addax nasomaculatus	9.470647e-01
## 165	Panthera leo (long maned)	Oryx dammah	9.188047e-01
## 166	Panthera leo (long maned)	Oryx beisa	9.274995e-01
## 167	Panthera leo (long maned)	Hippotragus equinus	8.697679e-01
## 168	Panthera leo (long maned)	Kobus kob	9.470647e-01
## 169	Panthera leo (long maned)	Kobus megaceros	9.470647e-01
## 170	Panthera leo (long maned)	Alcelaphus bucelaphus	9.274995e-01
## 171	Panthera leo (long maned)	Connochaetes taurinus	9.163894e-01
## 172	Panthera leo (long maned)	Litocranius walleri	9.473550e-01
## 173	Panthera leo (long maned)	Ammodorcas clarkei	9.414513e-01
## 174	Panthera leo (long maned)	Gazella dorcas	9.269240e-01
## 175	Panthera leo (long maned)	Gazella leptoceros	9.231621e-01
## 176	Panthera leo (long maned)	Gazella soemmerringii	9.492168e-01
## 177	Panthera leo (long maned)	Capra ibex	9.488401e-01
## 178	Panthera leo (long maned)	Ammotragus lervia	9.493506e-01
## 179	Panthera leo (long maned)	Bos primigenius	3.122850e-01
## 180	Panthera leo (long maned)	Syncerus caffer	8.523026e-01
## 181	Panthera leo (short maned)	Loxodonta africana	5.366696e-02
## 182	Panthera leo (short maned)	Equus asinus	8.387459e-01
## 183	Panthera leo (short maned)	Equus grevyi	7.801326e-01
## 184	Panthera leo (short maned)	Equus quagga	8.981503e-01
## 185	Panthera leo (short maned)	Diceros/Ceratotherium	5.001258e-01
## 186	Panthera leo (short maned)	Sus scrofa	9.473550e-01
## 187	Panthera leo (short maned)	Phacochoerus aethiopicus	9.493032e-01
## 188	Panthera leo (short maned)	Hippopotamus amphibius	6.508930e-01
## 189	Panthera leo (short maned)	Giraffa camelopardalis	7.002593e-01
## 190	Panthera leo (short maned)	Dama mesopotamica	9.379692e-01
## 191	Panthera leo (short maned)	Camelus dromedarius	7.408155e-01
## 192	Panthera leo (short maned)	Taurotragus oryx	8.175219e-01
## 193	Panthera leo (short maned)	Tragelaphus spekei	9.493506e-01
## 194	Panthera leo (short maned)	Addax nasomaculatus	9.470647e-01
## 195	Panthera leo (short maned)	Oryx dammah	9.188047e-01
## 196	Panthera leo (short maned)	Oryx beisa	9.274995e-01
## 197	Panthera leo (short maned)	Hippotragus equinus	8.697679e-01
## 198	Panthera leo (short maned)	Kobus kob	9.470647e-01
## 199	Panthera leo (short maned)	Kobus megaceros	9.470647e-01
## 200	Panthera leo (short maned)	Alcelaphus bucelaphus	9.274995e-01
## 201	Panthera leo (short maned)	Connochaetes taurinus	9.163894e-01
## 202	Panthera leo (short maned)	Litocranius walleri	9.473550e-01
## 203	Panthera leo (short maned)	Ammodorcas clarkei	9.414513e-01
## 204	Panthera leo (short maned)	Gazella dorcas	9.269240e-01
## 205	Panthera leo (short maned)	Gazella leptoceros	9.231621e-01
## 206	Panthera leo (short maned)	Gazella soemmerringii	9.492168e-01
## 207	Panthera leo (short maned)	Capra ibex	9.488401e-01

## 208	Panthera leo (short maned)	Ammotragus lervia	9.493506e-01
## 209	Panthera leo (short maned)	Bos primigenius	3.122850e-01
## 210	Panthera leo (short maned)	Syncerus caffer	8.523026e-01
## 211	Panthera pardus	Loxodonta africana	3.085524e-03
## 212	Panthera pardus	Equus asinus	5.313305e-01
## 213	Panthera pardus	Equus grevyi	3.945327e-01
## 214	Panthera pardus	Equus quagga	7.190257e-01
## 215	Panthera pardus	Diceros/Ceratotherium	1.041663e-01
## 216	Panthera pardus	Sus scrofa	9.435878e-01
## 217	Panthera pardus	Phacochoerus aethiopicus	9.301586e-01
## 218	Panthera pardus	Hippopotamus amphibius	2.107724e-01
## 219	Panthera pardus	Giraffa camelopardalis	2.665707e-01
## 220	Panthera pardus	Dama mesopotamica	8.718947e-01
## 221	Panthera pardus	Camelus dromedarius	3.246097e-01
## 222	Panthera pardus	Taurotragus oryx	4.766822e-01
## 223	Panthera pardus	Tragelaphus spekei	9.351033e-01
## 224	Panthera pardus	Addax nasomaculatus	9.132252e-01
## 225	Panthera pardus	Oryx dammah	7.958097e-01
## 226	Panthera pardus	Oryx beisa	8.296452e-01
## 227	Panthera pardus	Hippotragus equinus	6.228608e-01
## 228	Panthera pardus	Kobus kob	9.132252e-01
## 229	Panthera pardus	Kobus megaceros	9.132252e-01
## 230	Panthera pardus	Alcelaphus bucelaphus	8.296452e-01
## 231	Panthera pardus	Connochaetes taurinus	7.865675e-01
## 232	Panthera pardus	Litocranius walleri	9.435878e-01
## 233	Panthera pardus	Ammodorcas clarkei	9.482989e-01
## 234	Panthera pardus	Gazella dorcas	9.490591e-01
## 235	Panthera pardus	Gazella leptoceros	9.486356e-01
## 236	Panthera pardus	Gazella soemmerringii	9.369640e-01
## 237	Panthera pardus	Capra ibex	9.248399e-01
## 238	Panthera pardus	Ammotragus lervia	9.351033e-01
## 239	Panthera pardus	Bos primigenius	4.003430e-02
## 240	Panthera pardus	Syncerus caffer	5.695797e-01
## 241	Acinonyx jubatus	Loxodonta africana	8.198377e-04
## 242	Acinonyx jubatus	Equus asinus	3.433357e-01
## 243	Acinonyx jubatus	Equus grevyi	2.191298e-01
## 244	Acinonyx jubatus	Equus quagga	5.695797e-01
## 245	Acinonyx jubatus	Diceros/Ceratotherium	4.003430e-02
## 246	Acinonyx jubatus	Sus scrofa	9.323264e-01
## 247	Acinonyx jubatus	Phacochoerus aethiopicus	9.078861e-01
## 248	Acinonyx jubatus	Hippopotamus amphibius	9.431179e-02
## 249	Acinonyx jubatus	Giraffa camelopardalis	1.277646e-01
## 250	Acinonyx jubatus	Dama mesopotamica	8.074978e-01
## 251	Acinonyx jubatus	Camelus dromedarius	1.665390e-01
## 252	Acinonyx jubatus	Taurotragus oryx	2.900589e-01
## 253	Acinonyx jubatus	Tragelaphus spekei	9.166680e-01
## 254	Acinonyx jubatus	Addax nasomaculatus	8.783291e-01
## 255	Acinonyx jubatus	Oryx dammah	6.833708e-01
## 256	Acinonyx jubatus	Oryx beisa	7.373322e-01
## 257	Acinonyx jubatus	Hippotragus equinus	4.447852e-01
## 258	Acinonyx jubatus	Kobus kob	8.783291e-01
## 259	Acinonyx jubatus	Kobus megaceros	8.783291e-01

## 260	Acinonyx jubatus	Alcelaphus bucelaphus	7.373322e-01
## 261	Acinonyx jubatus	Connochaetes taurinus	6.690202e-01
## 262	Acinonyx jubatus	Litocranius walleri	9.323264e-01
## 263	Acinonyx jubatus	Ammodorcas clarkei	9.425155e-01
## 264	Acinonyx jubatus	Gazella dorcas	9.485214e-01
## 265	Acinonyx jubatus	Gazella leptoceros	9.489825e-01
## 266	Acinonyx jubatus	Gazella soemmerringii	9.200134e-01
## 267	Acinonyx jubatus	Capra ibex	8.985445e-01
## 268	Acinonyx jubatus	Ammotragus lervia	9.166680e-01
## 269	Acinonyx jubatus	Bos primigenius	1.343062e-02
## 270	Acinonyx jubatus	Syncerus caffer	3.837879e-01

c. Now create networks of who eats whom. We will start with adjacency matrices. We will assume all of our species are the size of females. For this step, don't worry about predators vs. prey yet, just calculate all of the feeding probabilities based on body sizes.

Hint: if you start with a square matrix of all zeros (one row and one column for each species), you can use a for loop to fill in that matrix with probabilities calculated from your function above.

```
#Empty matrix
species_list <- sp_mass$Species
num_species <- length(species_list)
feeding_matrix <- matrix(0, nrow = num_species, ncol = num_species, dimnames = list(species_list, species_list))

# Fill in the matrix with feeding probabilities based on body sizes
for (i in 1:num_species) {
  for (j in 1:num_species) {
    if (i != j) { # Ensure we're not calculating a species eating itself
      m_j <- sp_mass$f[i]
      m_i <- sp_mass$f[j]
      probability <- probEat(alpha, beta, gamma, m_j, m_i)
      feeding_matrix[i, j] <- probability
    }
  }
}
```

d. Now that you have your matrix of potential feeding interactions based on body size, use the 'carnivores' vector created above to set all of the feeding interactions of herbivores (0s in that vector) to zero. In food webs the columns are the higher trophic level and the rows are the lower. HINT: the function 'sweep' may be useful, though there are many approaches to do the needed matrix multiplication. Print the row and column sums.

```
# Now use the carnivores vector to zero out interactions involving herbivores
feeding_matrix <- sweep(feeding_matrix, 2, carnivores, "*") # Zero out columns of herbivores

# Print the row and column sums
row_sums <- rowSums(feeding_matrix)
col_sums <- colSums(feeding_matrix)

# Print the sums to see feeding interactions
print("Row sums (predator feeding counts):") #number of species each predator feeds on
```

```
## [1] "Row sums (predator feeding counts):"
```

```
# How likely a predator (species i) is to feed on any of the prey species (species j), summed across all prey species.
print(row_sums)
```

##	Canis aureus	Vulpes vulpes
##	3.653397	2.624839
##	Lycaon pictus	Hyaena hyaena
##	6.107261	6.629648
##	Crocuta crocuta	Panthera leo (long maned)
##	7.108173	7.182140
##	Panthera leo (short maned)	Panthera pardus
##	7.182140	7.230153
##	Acinonyx jubatus	Loxodonta africana
##	7.004501	3.340770
##	Equus asinus	Equus grevyi
##	7.350024	6.991301
##	Equus quagga	Diceros/Ceratotherium
##	7.845624	5.931349
##	Sus scrofa	Phacochoerus aethiopicus
##	7.778380	8.104993
##	Hippopotamus amphibius	Giraffa camelopardalis
##	6.436817	6.624679
##	Dama mesopotamica	Camelus dromedarius
##	8.215087	6.798155
##	Taurotragus oryx	Tragelaphus spekei
##	7.207914	8.033013
##	Addax nasomaculatus	Oryx dammah
##	8.211863	8.048611
##	Oryx beisa	Hippotragus equinus
##	8.131776	7.589451
##	Kobus kob	Kobus megaceros
##	8.211863	8.211863
##	Alcelaphus bucelaphus	Connochaetes taurinus
##	8.131776	8.024807
##	Litocranius walleri	Ammodorcas clarkei
##	7.778380	7.368397
##	Gazella dorcas	Gazella leptoceros
##	6.687505	6.549228
##	Gazella soemmerringii	Capra ibex
##	7.996055	8.154993
##	Ammotragus lervia	Bos primigenius
##	8.033013	5.260322
##	Syncerus caffer	
##	7.449649	

```
print("Column sums (prey being fed upon counts):") #number of predators feeding on each prey species
```

```
## [1] "Column sums (prey being fed upon counts):"
```

```
# The total feeding probability for that species as prey (across all predators).
print(col_sums)
```

##	Canis aureus	Vulpes vulpes
##	27.83918	24.97777
##	Lycaon pictus	Hyaena hyaena
##	32.65088	33.08111
##	Crocuta crocuta	Panthera leo (long maned)
##	32.96384	29.00279
##	Panthera leo (short maned)	Panthera pardus
##	29.00279	32.62155
##	Acinonyx jubatus	Loxodonta africana
##	33.08000	0.00000
##	Equus asinus	Equus grevyi
##	0.00000	0.00000
##	Equus quagga	Diceros/Ceratotherium
##	0.00000	0.00000
##	Sus scrofa	Phacochoerus aethiopicus
##	0.00000	0.00000
##	Hippopotamus amphibius	Giraffa camelopardalis
##	0.00000	0.00000
##	Dama mesopotamica	Camelus dromedarius
##	0.00000	0.00000
##	Taurotragus oryx	Tragelaphus spekei
##	0.00000	0.00000
##	Addax nasomaculatus	Oryx dammah
##	0.00000	0.00000
##	Oryx beisa	Hippotragus equinus
##	0.00000	0.00000
##	Kobus kob	Kobus megaceros
##	0.00000	0.00000
##	Alcelaphus bucelaphus	Connochaetes taurinus
##	0.00000	0.00000
##	Litocranius walleri	Ammodorcas clarkei
##	0.00000	0.00000
##	Gazella dorcas	Gazella leptoceros
##	0.00000	0.00000
##	Gazella soemmerringii	Capra ibex
##	0.00000	0.00000
##	Ammotragus lervia	Bos primigenius
##	0.00000	0.00000
##	Syncerus caffer	
##	0.00000	

Lab part 2: Breaking the networks into time periods

- With our matrix of feeding interaction we can create a web for each time period, including only the species that were not extinct in the period. Try first just using the second time period (the second column of 'sp_occ').

Use the function 'empty' from the bipartite package to empty the matrix of rows and columns with no interactions. The number of species in the second time period is 36 'sum(sp_occ[,2])'. Check to see that the number of rows in your network with probabilities > 0 is 36.

HINT: You will need to zero out the rows where a species is not present in that time period and the columns. The function 'sweep' may be useful again.

```
# Extract the presence vector for the second time period
presence <- sp_occ[, 2] # Second column of species occurrence matrix

# Zero out rows and columns of extinct species in the feeding matrix
filtered_matrix <- sweep(feeding_matrix, 1, presence, "*") # Zero out rows
filtered_matrix <- sweep(filtered_matrix, 2, presence, "*") # Zero out columns

active_species <- rowSums(filtered_matrix) > 0 # Rows with interaction probabilities > 0
print(paste("Number of species in the second time period:", sum(active_species))) # Should be 36
```

```
## [1] "Number of species in the second time period: 36"
```

```
#cleaned_matrix <- empty(filtered_matrix)
print(nrow(filtered_matrix))
```

```
## [1] 39
```

```
print(sum(sp_occ[,2]))
```

```
## [1] 36
```

- b. Now create a network for all of the time points by creating a list where each element is a network. You will need to use a for loop, or an 'lapply' if you feel like experimenting with apply functions. Print the first 5 columns and rows of the 5th time period.

HINT: If choosing the for loop route, remember to create an empty list of a specific length use the function 'vector'. To access a specific element of a list, use [[]], for example cool_list[[1]] accesses the first element of the list.


```

num_time_periods <- ncol(sp_occ)
time_period_networks <- vector("list", length = num_time_periods)

# Loop through each time period and create a filtered network
for (i in 1:num_time_periods) {
  # Extract the presence vector for the current time period
  presence <- sp_occ[, i]

  # Zero out rows and columns for extinct species
  filtered_matrix <- sweep(feeding_matrix, 1, presence, "*") # Zero out rows
  filtered_matrix <- sweep(filtered_matrix, 2, presence, "*") # Zero out columns

  # Clean the matrix to remove empty rows and columns
  time_period_networks[[i]] <- filtered_matrix
}

# Extract the matrix for the 5th time period
matrix_time_5 <- time_period_networks[[5]]

# Print the first 5 rows and columns
print(matrix_time_5[1:5, 1:5])

```

```

##              Canis aureus Vulpes vulpes Lycaon pictus Hyaena hyaena
## Canis aureus      0.0000000      0.9410046      0.7677278      0.6522657
## Vulpes vulpes     0.8936762      0.0000000      0.6188681      0.4550391
## Lycaon pictus     0.9493738      0.9458942      0.0000000      0.9011932
## Hyaena hyaena     0.9470427      0.9379146      0.9387502      0.0000000
## Crocuta crocuta   0.0000000      0.0000000      0.0000000      0.0000000
##              Crocuta crocuta
## Canis aureus              0
## Vulpes vulpes              0
## Lycaon pictus              0
## Hyaena hyaena              0
## Crocuta crocuta            0

```

Lab part 3: Visualize the networks

- Convert the adjacency matrices to igraph class objects using the function 'graph_from_adjacency_matrix'. You can use a for loop or an lapply. Because these are food webs, set the argument mode to "directed" and the argument diag to FALSE (this means a species cannot consumer members of its own species, i.e., no cannibalism/self-loops). Also remember that these interactions are weighted.

```
# Convert adjacency matrices to igraph objects
igraph_networks <- lapply(time_period_networks, function(matrix) {
  graph_from_adjacency_matrix(
    matrix,
    mode = "directed",
    weighted = TRUE,
    diag = FALSE
  )
})
```

b. Plot three networks of your choice, using different colors for the predators and prey.

```
# Define colors based on carnivore status
node_colors <- ifelse(carnivores == 1, "red", "green") # Red for predators, green for prey
groups <- ifelse(carnivores == 1, "Predator", "Prey")

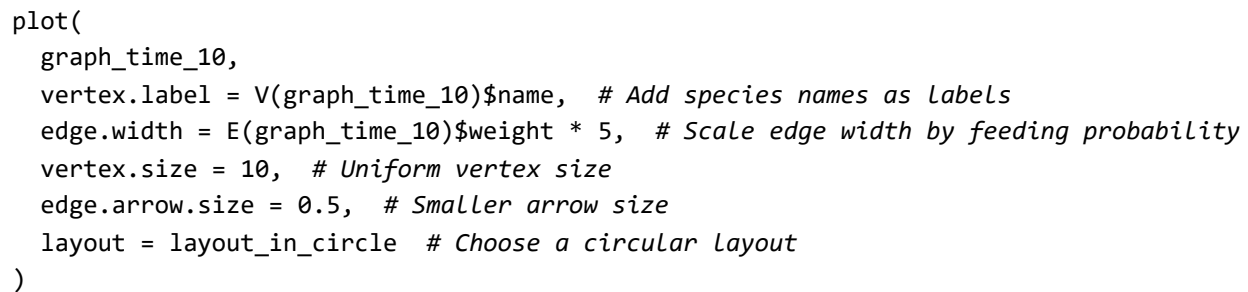
# Choose three networks to plot
graph_time_5 <- igraph_networks[[5]]
graph_time_10 <- igraph_networks[[10]]
graph_time_20 <- igraph_networks[[20]]

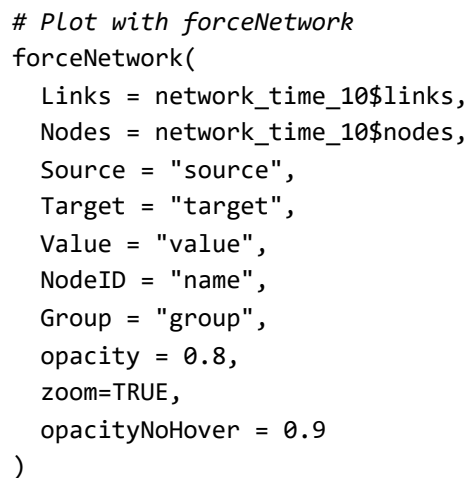
# Convert igraph objects to networkD3 objects with group information
network_time_5 <- igraph_to_networkD3(graph_time_5, group = groups)
network_time_10 <- igraph_to_networkD3(graph_time_10, group = groups)
network_time_20 <- igraph_to_networkD3(graph_time_20, group = groups)

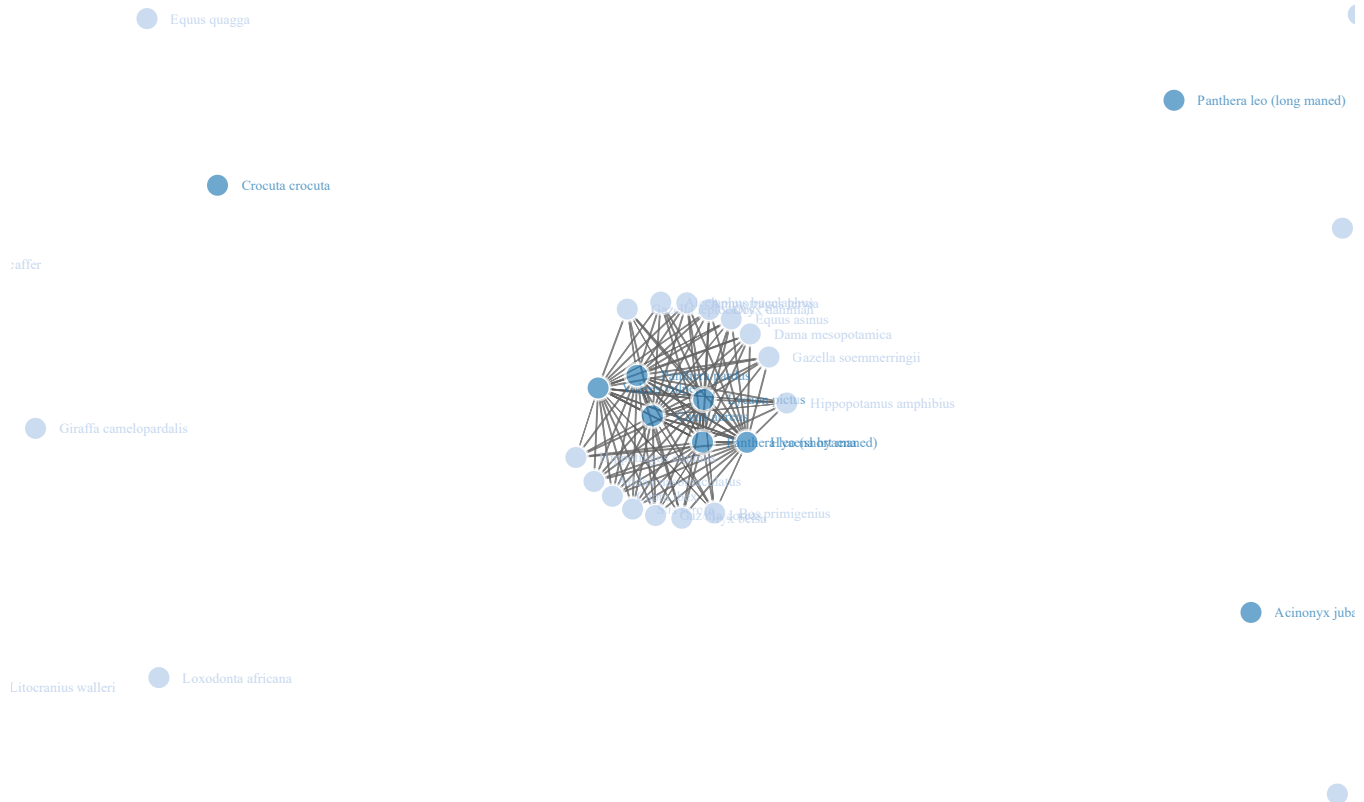
plot(
  graph_time_5,
  vertex.label = V(graph_time_5)$name, # Add species names as labels
  edge.width = E(graph_time_5)$weight * 5, # Scale edge width by feeding probability
  vertex.size = 10, # Uniform vertex size
  edge.arrow.size = 0.5, # Smaller arrow size
  layout = layout_in_circle # Choose a circular layout
)
```



```
# Plot with forceNetwork
forceNetwork(
  Links = network_time_5$links,
  Nodes = network_time_5$nodes,
  Source = "source",
  Target = "target",
  Value = "value",
  NodeID = "name",
  Group = "group",
  opacity = 0.8,
  zoom=TRUE,
  opacityNoHover = 0.9
)
```







```
plot(
  graph_time_20,
  vertex.label = V(graph_time_20)$name, # Add species names as labels
  edge.width = E(graph_time_20)$weight * 5, # Scale edge width by feeding probability
  vertex.size = 10, # Uniform vertex size
  edge.arrow.size = 0.5, # Smaller arrow size
  layout = layout_in_circle # Choose a circular layout
)
```

