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1                                         Appendix A1
2 Stan code to implement the Markov Chain Monte Carlo (MCMC) sampling for model 1
3
4
5 The function sparse_iar_lpdf() was devised by Joseph (2016)
6
7 functions {
8 /**
9 * Return the log probability of a proper intrinsic autoregressive (IAR) prior
10 * with a sparse representation for the adjacency matrix
11 *
12 * @param phi Vector containing the parameters with a IAR prior
13 * @param tau Precision parameter for the IAR prior (real)
14 * @param W_sparse Sparse representation of adjacency matrix (int array)
15 * @param n Length of phi (int)
16 * @param W_n Number of adjacent pairs (int)
17 * @param D_sparse Number of neighbors for each location (vector)
18 *
19 * @return Log probability density of IAR prior up to additive constant
20 */
21 real sparse_iar_lpdf(vector phi, real tau,
22 int[,] W_sparse, vector D_sparse, int n, int W_n) {
23     row_vector[n] phit_D; // phi' * D
24     row_vector[n] phit_W; // phi' * W
25     vector[n] ldet_terms;
26     phit_D = (phi .* D_sparse)';
27     phit_W = rep_row_vector(0, n);
28     for (i in 1:W_n) {
29         phit_W[W_sparse[i, 1]] = phit_W[W_sparse[i, 1]] + phi[W_sparse[i, 2]];
30         phit_W[W_sparse[i, 2]] = phit_W[W_sparse[i, 2]] + phi[W_sparse[i, 1]];

```

```

1      }
2      return 0.5 * ((n-1) * log(tau) - tau * (phit_D * phi - (phit_W * phi)));
3      }
4  }
5
6 data
7 {
8 int < lower = 1 > nData;
9 int < lower = 0 > damage[ nData ];
10 vector[ nData ] windSpeed;
11 int < lower = 1 > nForWindDir;
12 int < lower = 1, upper = nForWindDir > windDirIndex[ nData ];
13 matrix< lower = 0, upper = 1 >[ nForWindDir, nForWindDir ] WForWindDir; // adjacency
14 matrix
15 int < lower = 1 > W_nForWindDir;           // number of adjacent region pairs
16 int < lower = 1 > nForSeason;
17 int < lower = 1, upper = nForSeason > seasonIndex[ nData ];
18 matrix< lower = 0, upper = 1 >[ nForSeason, nForSeason ] WForSeason; // adjacency matrix
19 int < lower = 1 > W_nForSeason;           // number of adjacent region pairs
20 int < lower = 1 > nForDecade;
21 int < lower = 1, upper = nForDecade > decadeIndex[ nData ];
22 // real < lower = 0, upper = 100 > sigmaFirstRDecade;
23 int nObsMeanVol;
24 vector[ nObsMeanVol ] obsMeanVol;
25 int decadeWithMeanVol[ nObsMeanVol ];
26 int < lower = 0, upper = 100 > nWindSpeedForPred;
27 vector[ nWindSpeedForPred ] windSpeedForPred;
28 int < lower = 0, upper = 100 > nMeanVolForPred;
29 vector[ nMeanVolForPred ] meanVolForPred;
30 }

```

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1
2 transformed data
3 {
4     int W_sparseForWindDir[ W_nForWindDir, 2 ]; // adjacency pairs
5     vector[ nForWindDir ] D_sparseForWindDir; // diagonal of D (number of neigbors for each
6     site)
7     int W_sparseForSeason[ W_nForSeason, 2 ]; // adjacency pairs
8     vector[ nForSeason ] D_sparseForSeason; // diagonal of D (number of neigbors for each site)
9     real mWindSpeedForPred;
10    real mMeanVolForPred;
11
12    mWindSpeedForPred = mean( windSpeedForPred );
13    mMeanVolForPred = mean( meanVolForPred );
14
15    { // generate sparse representation for W
16        int counter;
17        counter = 1;
18        // loop over upper triangular part of W to identify neighbor pairs
19        for ( i in 1:( nForWindDir - 1 ) )
20        {
21            for ( j in ( i + 1 ):nForWindDir )
22            {
23                if ( WForWindDir[i, j] == 1 )
24                {
25                    W_sparseForWindDir[ counter, 1 ] = i;
26                    W_sparseForWindDir[ counter, 2 ] = j;
27                    counter = counter + 1;
28                }
29            }
30        }

```

```

1      }
2      for (i in 1:nForWindDir)
3          D_sparseForWindDir[ i ] = sum( WForWindDir[ i ] );
4      {
5          vector[ nForWindDir ] invsqrtDForWindDir;
6          for ( i in 1:nForWindDir )
7          {
8              invsqrtDForWindDir[ i ] = 1 / sqrt( D_sparseForWindDir[ i ] );
9          }
10     }
11
12 { // generate sparse representation for W
13     int counter;
14     counter = 1;
15     // loop over upper triangular part of W to identify neighbor pairs
16     for ( i in 1:( nForSeason - 1 ) )
17     {
18         for ( j in ( i + 1 ):nForSeason )
19         {
20             if ( WForSeason[i, j] == 1 )
21             {
22                 W_sparseForSeason[ counter, 1 ] = i;
23                 W_sparseForSeason[ counter, 2 ] = j;
24                 counter = counter + 1;
25             }
26         }
27     }
28 }
29 for (i in 1:nForSeason)
30     D_sparseForSeason[ i ] = sum( WForSeason[ i ] );

```

```

1   {
2     vector[ nForSeason ] invsqrtDForSeason;
3     for ( i in 1:nForSeason )
4     {
5       invsqrtDForSeason[ i ] = 1 / sqrt( D_sparseForSeason[ i ] );
6     }
7   }
8 }
9
10 parameters
11 {
12   real < lower = -100, upper = 100 > b0;
13   real < lower = -100, upper = 100 > bForWindDir;
14   vector[ nForWindDir ] phi_unscaledForWindDir;
15   real < lower = 0, upper = 100 > sigmaForWindDir;
16   vector[ nForSeason ] phi_unscaledForSeason;
17   real < lower = 0, upper = 100 > sigmaForSeason;
18   real bForMeanVol;
19   vector < lower = 0, upper = 500 > [ nForDecade ] estMeanVol;
20   real < lower = 0, upper = 100 > sigmaForDecade;
21   real < lower = 0, upper = 100 > sigmaForObsMeanVol;
22 }
23
24 transformed parameters
25 {
26   vector[ nForWindDir ] phiForWindDir; // brute force centering
27   vector[ nForSeason ] phiForSeason; // brute force centering
28   // vector[ nData ] linearPredictor;
29   vector[ nData ] p;
30   real tauForWindDir;

```

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*Supplementary Appendices S1 and S2 to the article “A model for longitudinal data sets relating wind-damage probability to biotic and abiotic factors: a Bayesian approach”, by Kiyoshi Umeki, Marc D. Abrams, Keisuke Toyama and Eri Nabeshima. Forest Systems Vol. 28 No. 3, December 2019 (<https://doi.org/10.5424/fs/2019283-15200>)*

```

1 real tauForSeason;
2
3 phiForWindDir = phi_unscaledForWindDir - mean( phi_unscaledForWindDir );
4 phiForSeason = phi_unscaledForSeason - mean( phi_unscaledForSeason );
5 // linearPredictor = b1 * windSpeed + phiForWindDir[ windDirIndex ] +
6 phiForSeason[ seasonIndex ] + rDecade[ decadeIndex ];
7 p = inv_logit( b0 + bForWindDir * windSpeed + bForMeanVol * estMeanVol[ decadeIndex ] +
8 phiForWindDir[ windDirIndex ] + phiForSeason[ seasonIndex ] );
9 tauForWindDir = 1 / ( sigmaForWindDir * sigmaForWindDir );
10 tauForSeason = 1 / ( sigmaForSeason * sigmaForSeason );
11 }
12
13 model
14 {
15 phi_unscaledForWindDir ~
16 sparse_iar( tauForWindDir, W_sparseForWindDir, D_sparseForWindDir, nForWindDir,
17 W_nForWindDir );
18 phi_unscaledForSeason ~
19 sparse_iar( tauForSeason, W_sparseForSeason, D_sparseForSeason, nForSeason,
20 W_nForSeason );
21
22 estMeanVol[ 1 ] ~ normal( 0, 100 );
23 estMeanVol[ 2:nForDecade ] ~ normal( estMeanVol[ 1:( nForDecade - 1 ) ], sigmaForDecade );
24 obsMeanVol ~ normal( estMeanVol[ decadeWithMeanVol ], sigmaForObsMeanVol );
25 damage ~ bernoulli( p );
26 }
27
28 generated quantities
29 {
30 vector [ nWindSpeedForPred ] predPForWindSpeed;

```

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*Supplementary Appendices S1 and S2 to the article “A model for longitudinal data sets relating wind-damage probability to biotic and abiotic factors: a Bayesian approach”, by Kiyoshi Umeki, Marc D. Abrams, Keisuke Toyama and Eri Nabeshima. Forest Systems Vol. 28 No. 3, December 2019 (<https://doi.org/10.5424/fs/2019283-15200>)*

```

1 vector [ nForWindDir ] predPForWindDir;
2 vector [ nForSeason ] predPForSeason;
3 vector [ nMeanVolForPred ] predPForMeanVol;
4 vector [ nData ] log_lik;
5
6 for ( i in 1:nData )
7 {
8     log_lik[ i ] = bernoulli_lpmf( damage[ i ] | p[ i ] );
9 }
10
11 predPForWindSpeed = inv_logit( b0 + bForWindDir * windSpeedForPred + bForMeanVol *
12 mMeanVolForPred );
13 predPForWindDir = inv_logit( b0 + bForWindDir * mWindSpeedForPred + bForMeanVol *
14 mMeanVolForPred + phiForWindDir );
15 predPForSeason = inv_logit( b0 + bForWindDir * mWindSpeedForPred + bForMeanVol *
16 mMeanVolForPred + phiForSeason );
17 predPForMeanVol = inv_logit( b0 + bForWindDir * mWindSpeedForPred + bForMeanVol *
18 meanVolForPred );
19 }
20
21 References
22
23 Joseph M, 2016. Exact sparse CAR models in Stan. https://mc-
24 stan.org/users/documentation/case-studies/mbjoseph-CARStan.html. [1 May 2019]
25
26

```

```

1                                         Appendix A2
2   Stan code to implement the Markov Chain Monte Carlo (MCMC) sampling for model 2
3
4
5   The function sparse_iar_lpdf() was devised by Joseph (2016)
6
7   functions {
8     /**
9      * Return the log probability of a proper intrinsic autoregressive (IAR) prior
10     * with a sparse representation for the adjacency matrix
11     *
12     * @param phi Vector containing the parameters with a IAR prior
13     * @param tau Precision parameter for the IAR prior (real)
14     * @param W_sparse Sparse representation of adjacency matrix (int array)
15     * @param n Length of phi (int)
16     * @param W_n Number of adjacent pairs (int)
17     * @param D_sparse Number of neighbors for each location (vector)
18     *
19     * @return Log probability density of IAR prior up to additive constant
20   */
21   real sparse_iar_lpdf(vector phi, real tau,
22     int[,] W_sparse, vector D_sparse, int n, int W_n) {
23     row_vector[n] phit_D; // phi' * D
24     row_vector[n] phit_W; // phi' * W
25     vector[n] ldet_terms;
26     phit_D = (phi .* D_sparse)';
27     phit_W = rep_row_vector(0, n);
28     for (i in 1:W_n) {
29       phit_W[W_sparse[i, 1]] = phit_W[W_sparse[i, 1]] + phi[W_sparse[i, 2]];
30       phit_W[W_sparse[i, 2]] = phit_W[W_sparse[i, 2]] + phi[W_sparse[i, 1]];

```

```

1      }
2      return 0.5 * ((n-1) * log(tau) - tau * (phit_D * phi - (phit_W * phi)));
3      }
4  }
5
6  data
7  {
8  int < lower = 1 > nData;
9  int < lower = 0 > damage[ nData ];
10 vector[ nData ] windSpeed;
11 int < lower = 1 > nForWindDir;
12 int < lower = 1, upper = nForWindDir > windDirIndex[ nData ];
13 matrix< lower = 0, upper = 1 >[ nForWindDir, nForWindDir ] WForWindDir; // adjacency
14 matrix
15 int < lower = 1 > W_nForWindDir;           // number of adjacent region pairs
16 int < lower = 1 > nForSeason;
17 int < lower = 1, upper = nForSeason > seasonIndex[ nData ];
18 matrix< lower = 0, upper = 1 >[ nForSeason, nForSeason ] WForSeason; // adjacency matrix
19 int < lower = 1 > W_nForSeason;           // number of adjacent region pairs
20 int < lower = 1 > nForDecade;
21 int < lower = 1, upper = nForDecade > decadeIndex[ nData ];
22 // real < lower = 0, upper = 100 > sigmaFirstRDecade;
23 int < lower = 0, upper = 100 > nWindSpeedForPred;
24 vector[ nWindSpeedForPred ] windSpeedForPred;
25 }
26
27 transformed data
28 {
29  int W_sparseForWindDir[ W_nForWindDir, 2 ]; // adjacency pairs

```

```

1  vector[ nForWindDir ] D_sparseForWindDir; // diagonal of D (number of neigbors for each
2  site)
3  int W_sparseForSeason[ W_nForSeason, 2 ]; // adjacency pairs
4  vector[ nForSeason ] D_sparseForSeason; // diagonal of D (number of neigbors for each site)
5  real mWindSpeedForPred;
6
7  mWindSpeedForPred = mean( windSpeedForPred );
8
9  { // generate sparse representation for W
10    int counter;
11    counter = 1;
12    // loop over upper triangular part of W to identify neighbor pairs
13    for ( i in 1:( nForWindDir - 1 ) )
14    {
15      for ( j in ( i + 1 ):nForWindDir )
16      {
17        if ( WForWindDir[i, j] == 1 )
18        {
19          W_sparseForWindDir[ counter, 1 ] = i;
20          W_sparseForWindDir[ counter, 2 ] = j;
21          counter = counter + 1;
22        }
23      }
24    }
25  }
26  for (i in 1:nForWindDir)
27    D_sparseForWindDir[ i ] = sum( WForWindDir[ i ] );
28  {
29    vector[ nForWindDir ] invsqrtDForWindDir;
30    for ( i in 1:nForWindDir )

```

```

1   {
2     invsqrtDForWindDir[ i ] = 1 / sqrt( D_sparseForWindDir[ i ] );
3   }
4 }
5
6 { // generate sparse representation for W
7   int counter;
8   counter = 1;
9   // loop over upper triangular part of W to identify neighbor pairs
10  for ( i in 1:( nForSeason - 1 ) )
11  {
12    for ( j in ( i + 1 ):nForSeason )
13    {
14      if ( WForSeason[i, j] == 1 )
15      {
16        W_sparseForSeason[ counter, 1 ] = i;
17        W_sparseForSeason[ counter, 2 ] = j;
18        counter = counter + 1;
19      }
20    }
21  }
22 }
23 for (i in 1:nForSeason)
24   D_sparseForSeason[ i ] = sum( WForSeason[ i ] );
25 {
26   vector[ nForSeason ] invsqrtDForSeason;
27   for ( i in 1:nForSeason )
28   {
29     invsqrtDForSeason[ i ] = 1 / sqrt( D_sparseForSeason[ i ] );
30   }

```

```

1      }
2  }
3
4 parameters
5 {
6   real < lower = -100, upper = 100 > b1;
7   vector[ nForWindDir ] phi_unscaledForWindDir;
8   real < lower = 0, upper = 100 > sigmaForWindDir;
9   vector[ nForSeason ] phi_unscaledForSeason;
10  real < lower = 0, upper = 100 > sigmaForSeason;
11  vector [ nForDecade ] rDecade;
12  real < lower = 0, upper = 100 > sigmaForDecade;
13 }
14
15 transformed parameters
16 {
17   vector[ nForWindDir ] phiForWindDir; // brute force centering
18   vector[ nForSeason ] phiForSeason; // brute force centering
19   // vector[ nData ] linearPredictor;
20   vector[ nData ] p;
21   real tauForWindDir;
22   real tauForSeason;
23
24   phiForWindDir = phi_unscaledForWindDir - mean( phi_unscaledForWindDir );
25   phiForSeason = phi_unscaledForSeason - mean( phi_unscaledForSeason );
26   // linearPredictor = b1 * windSpeed + phiForWindDir[ windDirIndex ] +
27   phiForSeason[ seasonIndex ] + rDecade[ decadeIndex ];
28   p = inv_logit( b1 * windSpeed + phiForWindDir[ windDirIndex ] +
29   phiForSeason[ seasonIndex ] + rDecade[ decadeIndex ] );
30   tauForWindDir = 1 / ( sigmaForWindDir * sigmaForWindDir );

```

```

1  tauForSeason = 1 / ( sigmaForSeason * sigmaForSeason );
2 }
3
4 model
5 {
6   phi_unscaledForWindDir ~
7     sparse_iar( tauForWindDir, W_sparseForWindDir, D_sparseForWindDir, nForWindDir,
8     W_nForWindDir );
9   phi_unscaledForSeason ~
10    sparse_iar( tauForSeason, W_sparseForSeason, D_sparseForSeason, nForSeason,
11    W_nForSeason );
12
13   rDecade[ 1 ] ~ normal( 0, 100 );
14   rDecade[ 2:nForDecade ] ~ normal( rDecade[ 1:( nForDecade - 1 ) ], sigmaForDecade );
15   damage ~ bernoulli( p );
16   // target += bernoulli_logit_lpmf( damage | linearPredictor );
17 }
18
19 generated quantities
20 {
21   vector [ nWindSpeedForPred ] predPForWindSpeed;
22   vector [ nForWindDir ] predPForWindDir;
23   vector [ nForSeason ] predPForSeason;
24   vector [ nForDecade ] predPForDecade;
25   vector [ nData ] log_lik;
26
27   for ( i in 1:nData )
28   {
29     log_lik[ i ] = bernoulli_lpmf( damage[ i ] | p[ i ] );
30   }

```

```
1 predPForWindSpeed = inv_logit( b1 * windSpeedForPred + mean( rDecade ) );
2 predPForWindDir = inv_logit( b1 * mWindSpeedForPred + mean( rDecade ) +
3 phiForWindDir );
4 predPForSeason = inv_logit( b1 * mWindSpeedForPred + mean( rDecade ) + phiForSeason );
5 predPForDecade = inv_logit( b1 * mWindSpeedForPred + rDecade );
6 }
7
```

## 8 **References**

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9
10 Joseph M, 2016. Exact sparse CAR models in Stan. https://mc-
11 stan.org/users/documentation/case-studies/mbjoseph-CARStan.html. [1 May 2019]
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12
13
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