```
1.Functia frepcomgen(n, m):
frepcomgen <- function(n, m)</pre>
 joint_distribution = generate_random_joint_distribution(n, m)
 print("Repartitia comuna completa")
 print(joint distribution)
 partial joint distribution = erase some values(joint distribution = joint distribution)
 print("Repartitia comuna incompleta")
 print(partial_joint_distribution)
 return(partial_joint_distribution)
a) generate_random_joint_distribution
  x = c(sample(1:(10 * n), n, rep=FALSE))
  y = c(sample(1:(10 * m), m, rep=FALSE))
  x = sort(x)
  y = sort(y)
  aux = matrix(sample(1:(n*m), n * m, replace = TRUE), nrow = n, ncol = m)
  joint distribution = matrix(0, nrow = n + 2, ncol = m + 2)
      -imi generez in x si y valorile pentru cele doua variabile discrete
      -in aux voi avea o matrice de NxM elemente pe care le voi insuma si apoi
       le voi normaliza si trece in joint_distribution astfel incat ele sa poata
       reprezenta o repartitie comuna(impart la suma lor initiala)
        joint_distribution[i, j] = aux[i - 1,j - 1]
        joint distribution[i, j] = joint distribution[i, j] / sum(aux)
      -in joint_distribution voi avea pe prima linie si pe prima coloana,
       valorile lui y, respectiv valorile lui x; pe ultima linie si ultima
       coloana voi avea probabilitatile valorilor lui y, respectiv x
              joint distribution[i + 1, 1] = x[i]
              joint_distribution[i + 1, m + 2] = rowSums(aux)[i]
              joint distribution[1, i + 1] = y[i]
              joint_distribution[n + 2, i + 1] = colSums(aux)[i]
b) erase_some_values
      - sterg elementele matricei de pe diagonala principala si le marchez cu -1
                     for (p in 3:mini) {
                       joint distribution[p, p] = -1
2. Functia fcomplrepcom
```

- -cat timp la pasul anterior au fost gasite noi elemente in matrice se face o noua iteratie prin matrice si se incearca aflarea unui nou element al matricei
- se gaseste un nou element al matricei atunci cand este singurul necunoscut de pe aceeasi linie sau coloana functia solve-element

```
if (A[r, col] != -1) {
                             no elem line = 1 + no elem line
                             sum_line = sum_line + A[r, col]
                        if (no elem line == C - 3) {
                           ans = A[r, C] - sum line
                           return(ans)
                        for (row in 2:(R - 1)) {
                           if (A[row, c] != -1) {
                             no elem col = 1 + no elem col
                             sum col = sum col + A[row, c]
                           }
                        if (no elem col == R - 3) {
                           ans = A[R, c] - sum col
                           return(ans)
3. Functia care calculeaza Cov(5x, -3y)
      - mai intai calculez covarianta variabilelor:
mean xv = 0
for (row in 2:(nrow(joint_distribution) - 1)) {
 for (col in 2:(ncol(joint_distribution) - 1)) {
   mean_xy = (mean_xy + (joint_distribution[row, col] * joint_distribution[1, col] * joint_distribution[row, 1]))
mean_x = 0
mean_y = 0
for (row in 2:(nrow(joint distribution) - 1)) {
 mean_x = (joint_distribution[row, 1] * joint_distribution[row, ncol(joint_distribution)] + mean_x)
for (col in 2:(ncol(joint_distribution) - 1)) {
 mean_y = (joint_distribution[1, col] * joint_distribution[nrow(joint_distribution), col] + mean_y)
res = (mean_xy - (mean_x * mean_y) )
return(res)
       - apoi in functie inmultesc pe res cu (-15)
4. Functia care calculeaza P(0<X<3/Y>2) si P(X>6,Y<7)
       - P(0<X<3/Y>2) = P(0<X<3 \cap Y>2) / P(Y > 2)
for (row in 2:(nrow(joint_distribution) - 1)){
  for (col in 2:(ncol(joint_distribution) - 1)) {
    if (joint distribution[1, col] > 2) {
       s2 = s2 + joint_distribution[row, col]
       if (joint_distribution[row, 1] > 0 && joint_distribution[row, 1] < 3) {</pre>
         s1 = s1 + joint distribution[row, col]
    }
  }
res1 = s1 / s2
```

for (col in 2:(C - 1)) {

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-P(X>6,Y<7) = P(X>6 \cap Y<7)
for (row in 2:(nrow(joint distribution) - 1)) {
   for (col in 2:(ncol(joint_distribution) - 1)) {
     if (joint distribution[row, 1] > 6 && joint distribution[1, col] < 7) {</pre>
       res2 = res2 + joint distribution[row, col]
     }
  }
}
5. Functia fverind:
      - pentru fiecare element al matricei ex: a; verific sa fie egal cu
      produsul dintre probabilitatea lui x_i si probabilitatea lui y_i daca gasesc
      un element care nu indeplineste aceasta conditie atunci cele doua
      variabilele sunt dependente
for (row in 2:(R - 1)){
  for (col in 2:(C - 1)) {
   if (joint distribution[row, col] %!=% (joint distribution[row, C] * joint distribution[R, col])) {
       print("Variabilele sunt dependente")
     }
   }
print("Variabilele sunt independente")
6. Functia fvernecor:
      -calculez indicele \varphi = cov(x,y)/sqrt(var(x) * var(y))
      -covarianta o calculez folosind functia definita si la subpunctul anterior
      -pentru varianta folosesc functia:
variance discrete variable <- function(discrete var) {</pre>
  mean var = 0
  mean var2 = 0
  for (col in ncol(discrete var)) {
    mean var2 = (mean var2 + (discrete var[1, col] ^ 2) * discrete var[2, col])
    mean_var = (mean_var + discrete_var[1, col] * discrete_var[2, col])
  }
  res = mean var2 - (mean var ^ 2)
  return(res)
}
```