№2

Вывод первоначальной версии программы:

```
Solution:

x(0) = 1.000000

x(1) = 2.000000

x(2) = 2.000000

x(3) = -0.000000
```

Проверим решение с помощью питру:

```
import numpy as np

A = np.array([
    [2,5,4,1],
    [1,3,2,1],
    [2,10,9,7],
    [3,8,9,2]
])
b = np.array([20,11,40,37])
print(np.linalg.solve(A, b))

[ 1.000000000e+00     2.00000000e+00     2.00000000e+00     -3.70074342e-15]
```

Решения совпадают.

Время работы программы для матрицы размером матrix_size:

```
Serial Gauss duration: 753154 microseconds
```

Nº3

Analysis Configuration Collection Log Summary Bottom-up Caller/Callee Top-down Tree Platform

CPU Time [®]: 5.295s
 Total Thread Count: 1
 Paused Time [®]: 0s

▼ Top Hotspots

This section lists the most active functions in your application. Optimizing these hotspot functions typically results in improving overall application performance.

| Function | Module | CPU Time® |
|-------------------|---------------|-----------|
| SerialGaussMethod | IPS1.exe | 5.076s |
| rand | ucrtbased.dll | 0.200s |
| free_dbg | ucrtbased.dll | 0.010s |
| malloc | ucrtbased.dll | 0.009s |

*N/A is applied to non-summable metrics.

Hotspots Insights

If you see significant hotspots in the Top Hotspots list, switch to the Bottom-up view for in-depth analysis per function.
Otherwise, use the Caller/Callee view to track critical paths for these hotspots.

Explore Additional Insights

Parallelism ②: 21.3% ►
Use ③ Threading to
explore more
opportunities to
increase parallelism in
your application.

Напишем заведомо неправильную параллелизацию метода Гаусса:

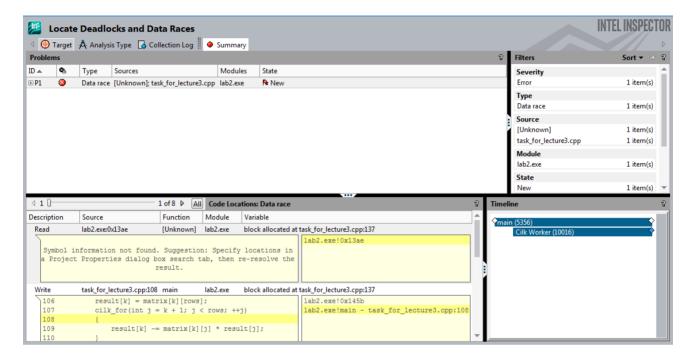
```
int ParallelGaussMethod(double** matrix, const int rows, double* result)
{
    int k;
    auto begin = std::chrono::high_resolution_clock::now();
    for (k = 0; k < rows; ++k)
        for(int i = k + 1; i < rows; ++i) {
            double koef = -matrix[i][k] / matrix[k][k];
            cilk_for (int j = k; j <= rows; ++j)</pre>
                matrix[i][j] += koef * matrix[k][j];
        }
    auto end = std::chrono::high_resolution_clock::now();
    result[rows - 1] = matrix[rows - 1][rows] / matrix[rows - 1][rows - 1];
    for (k = rows - 2; k \ge 0; --k) {
        cilk::reducer_opadd<double> res(matrix[k][rows]);
        cilk_for(int j = k + 1; j < rows; ++j)
            res -= matrix[k][j] * result[j];
        result[k] = res->get_value() / matrix[k][k];
    }
    return (int)std::chrono::duration_cast<std::chrono::microseconds>(end - be
}
```

Serial Gauss duration: 807899 microseconds
Parallel Gauss duration: 3799353 microseconds

Boost ratio: 0.212641

Скорость выполнения ожидаемо снизилась.

N₀4



Инспектор обнаружил гонку данных. Исправляем:

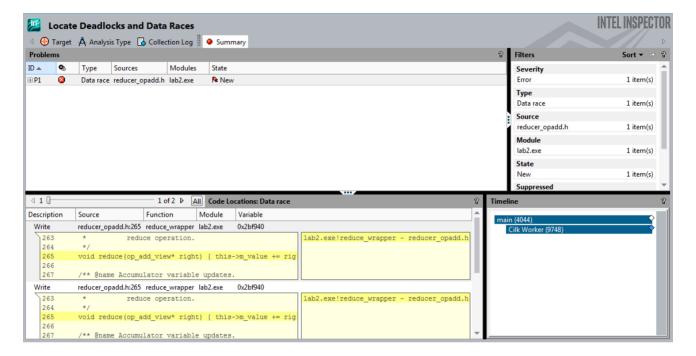
```
int ParallelGaussMethod(double** matrix, const int rows, double* result)
{
   int k;
   auto begin = std::chrono::high_resolution_clock::now();

   for (k = 0; k < rows; ++k)
        cilk_for(int i = k + 1; i < rows; ++i) {
            double koef = -matrix[i][k] / matrix[k][k];
            for (int j = k; j <= rows; ++j)
                matrix[i][j] += koef * matrix[k][j];
        }
   auto end = std::chrono::high_resolution_clock::now();

   result[rows - 1] = matrix[rows - 1][rows] / matrix[rows - 1][rows - 1];
   for (k = rows - 2; k >= 0; --k) {
```

```
cilk::reducer_opadd<double> res(matrix[k][rows]);
    cilk_for(int j = k + 1; j < rows; ++j)
        res -= matrix[k][j] * result[j];
    result[k] = res->get_value() / matrix[k][k];
}
return (int)std::chrono::duration_cast<std::chrono::microseconds>(end - be)
}
```

Повторный анализ:



N₂5

Решение для тестовой матрицы параллельным методом:

```
Solution:

x(0) = 1.000000

x(1) = 2.000000

x(2) = 2.000000

x(3) = -0.000000
```

Сравнение времени выполнения последовательного и параллельного методов:

Serial Gauss duration: 766317 microseconds Parallel Gauss duration: 616903 microseconds Boost ratio: 1.242200