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Evaluation of Different Storage Formats for Geospatial Data Caches Including a Sample Implementation to Enable ArcGIS Portal to Export Caches

Introduction

Exporting cached maps might take days. Airbus intended to analyze the performance of: *Compact Cache, GeoPackage, MBTiles & LuciadFusion*, by developing an algorithm in a web application based on WPS.

Check if tile exists in cache and copy tiles: generate files on server, zip files, open file in browser by JavaScript & send link by email.

Each cache format contains a distinct data structure \rightarrow **GeoPackage & MBTiles:** store tiles in a list \rightarrow insert list into database per transaction. Locked database \rightarrow no parallelization possible

Compact Cache & LuciadFusion contain multiple files, while *GeoPackage & MBTiles* are made of single database file. It is tested whether the combination of multiprocessing & multithreading optimizes Compact Cache to achieve a similar speed to the database formats.

Algorithm

•If drew polygons, transform coordinates from EPSG 3857 to 4326: the geoportal's map is Web Mercator $3857 \rightarrow 4326$.

•Convert from 4326 into tiles mechanism and from tiles into 4326: create bbox (tile mechanism) around polygons (4326).

Inside bbox, check if tile (\rightarrow 4326) is intersected with polygon \rightarrow (saved as tiles), 3 loops are executed: Z (zoom) level, Y & X coordinates of each tile \rightarrow Higher Z \rightarrow More iterations of (X, Y) \rightarrow More tiles to be calculated, More intersection tests. Three polygons are processed with different LOD - level of detail: world - lowest **LuciadFusion:** Luciad provides special methods to store tiles faster \rightarrow more effective under high zoom \rightarrow employs Java \rightarrow wrapped for Python \rightarrow multiprocessing isn't possible through Pyjnius, which wraps the code.

Esri Compact Cache: Mapproxy blocks multiprocessing \rightarrow assign relevant tiles to each bundle file \rightarrow split bundles to be processed in multiple cores. Python's GIL blocks multithreading \rightarrow use Numba's nogil decorator \rightarrow in each core, chunks of data are processed in parallel by multiple threads.



Z, national - higher Z, regional - highest Z.



Fig. 1: Paralelization Techniques

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Fig. 2: Cache Formats Processing Time Comparison

Conclusion

When Compact Cache is optimized, it's at least as fast as the database formats - by combination of multiprocessing & multithreading, else it is the slowest \rightarrow one should consider employing database logic, or combination of multiprocessing & multithreading to export cache files.

