

Digital Integrated Circuit Graph Data

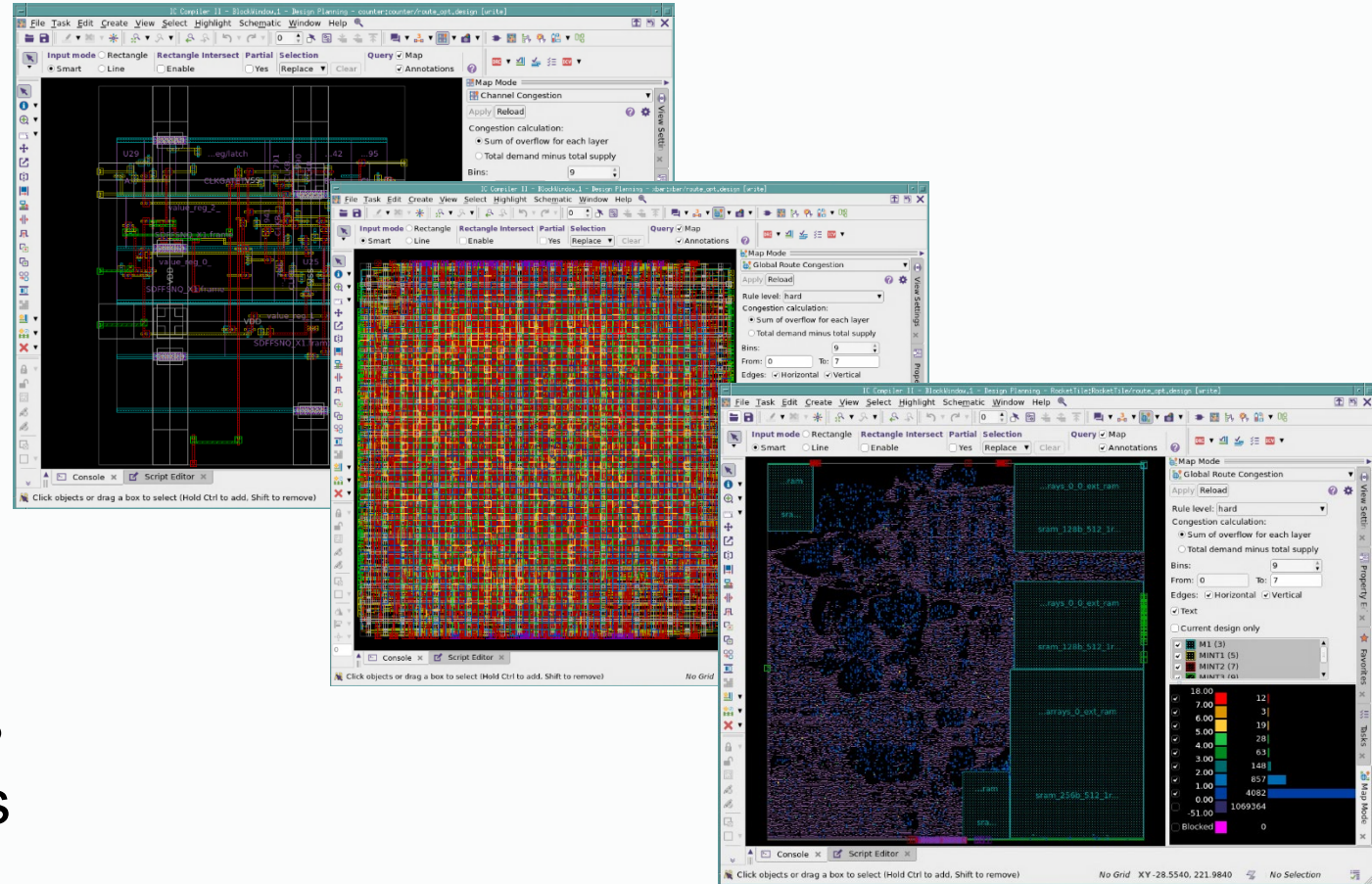
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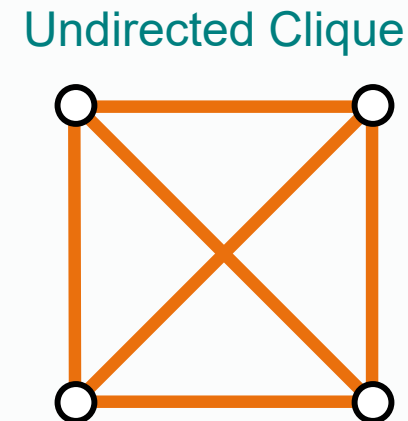
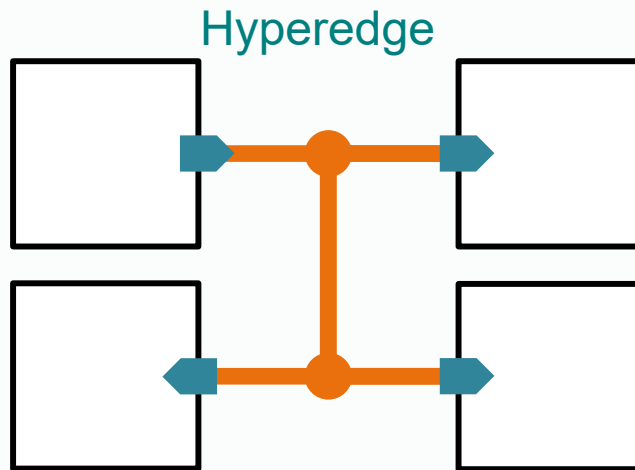
Example Designs

- counter
 - Simple starting example
 - 23 cells & 29 nets
- xbar
 - Highly congested routing
 - 3952 cells & 4484 nets
- RocketTile
 - Contains 5 memory macros
 - 183,560 cells & 84,494 nets



Hypergraph Representations

- How to represent hyperedges (*i.e.* nets) in a graph data-structure?
- Undirected clique method – Graph with all vertices fully connected
- Modified incidence matrix method – Sparse matrix with instance rows and net columns, value represents connected terminal index (or 0 if no connection)



Files provided for each design

- [design]_route_opt.def – Original design data
- [design]_clique.igpz – design as igraph Graph saved in “picklez” format
- **[design].json.gz – Instance and Net data**
- **[design]_connectivity.npz – Incidence matrix data as NumPy arrays**
- [design]_congestion.rpt – Routing congestion report
- [design]_congestion.npz – congestion as NumPy arrays
- [design]-congestion.png – Screen-shot of layout and congestion
- **Also find the cell library data in cells.json.gz and original library data in the lef subdirectory**

New

New

Supplemental files

- `route_opt.check_routes` – routing violations and resources usage per layer
- `route_opt.report_utilization` – ratio of required area to available area
- `route_opt.report_clock_qor.structure` – detailed clock-tree information
- `route_opt.report_qor` – Summary of system timing and other details

igraph Design Data (Clique Method)

```
>>> import igraph
>>> g = igraph.read('counter.igpz','picklez')
>>> len(g.vs)
23
>>> g.vs.attributes()
['instname', 'xloc', 'yloc', 'cell', 'orient']
>>> len(g.es)
56
>>> g.es.attributes()
['weight', 'netnames']
```

- **Vertex Attributes**

- **instname** (string) – Instance name
- **cell** (int) – Master library cell ID (array index)
- **xloc, yloc** (int) – location of the instance
- **orient** (int) – orientation of the instance

- read graph file
- get number of vertices
- get attributes for each vertex
- get number of edges
- get attributes for each edge

- **Edge Attributes**

- **weight** (float) – edge weight, $4/[n^2 - \text{mod}(n,2)]$
(where n = hyperedge vertices)
- **netnames** (string) – Net name(s) associated with this edge

Design Feature Data (Nets)

- Because multi-dimensional sparse arrays are not yet widely available, we cannot easily encode feature data in the incidence matrix
- Therefore, feature data for instances and nets are provided in [design].json.gz
- Contains a JSON object with the 'nets' member, which is an array of net objects
- Each net object has the following members:
 - name (str) – name of the net
 - id (int) – index of the net in the array

```
>>> with gzip.open('counter.json.gz','rb') as f:  
...     design = json.loads(f.read().decode('utf-8'))  
>>> design['nets'][3]  
{ 'name': 'zero', 'id': 3 }
```

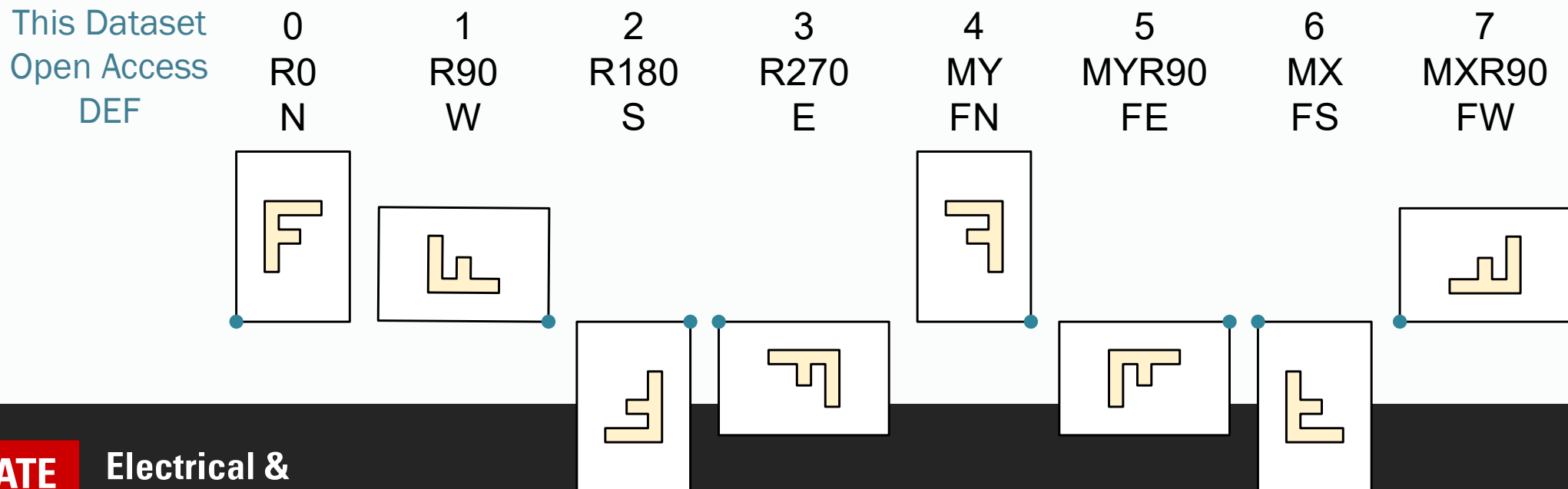
Design Feature Data (Instances)

- The JSON object in [design].json.gz also contains the 'instances' member, which is an array of instance objects
- Each instance object has the following members:
 - name (str) – name of the instance
 - id (int) – index of the instance in the array
 - cell (int) – Master library cell ID (array index)
 - xloc, yloc (int) – location of the instance in ½ nm units (divide by 2000 to get microns)
 - orient (int) – orientation of the instance

```
>>> design['instances'][6]
{'name': 'U22', 'id': 6, 'xloc': 6784, 'yloc': 7680, 'cell': 7, 'orient': 6}
```


Orientations

- Orientation of each instance within a design is given by an integer as shown below (with OpenAccess and DEF orientations for comparison)
- Blue dot indicates the location (xloc & yloc) for each orientation
- Note that DEF always restricts the location to lower-left corner (unlike the Open Access location, which is used in our data-set)



Cell Library Data

- Cell library data provided in cells.json.gz
- Contains an array of JSON objects with the following members
 - name (str) – name of the cell / module
 - id (int) – index of the cell in the array
 - width (int) – width of the cell in $\frac{1}{2}$ nm units (divide by 2000 to get microns)
 - height (int) – height of the cell in $\frac{1}{2}$ nm units (divide by 2000 to get microns)
 - terms – an array of terminal objects

```
>>> import json, gzip
>>> with gzip.open('cells.json.gz','rb') as f:
...     cells = json.loads(f.read().decode('utf-8'))
>>> cells[7]
{'name': 'AOI21_X1', 'id': 7, 'width': 768, 'height': 1536, 'terms': [{'name': 'A1', ...
```

Terminal Objects

- Terminal objects contain the following members
 - name (str) – name of the terminal
 - id (int) – integer index of terminal as it appears in the incidence matrix (same as array index plus 1)
 - dir (int) – direction of the terminal
 - 0 – input
 - 1 – output
 - 2 – inout
 - TBD – location of terminal

```
>>> cells[7]['terms'][2]  
{ 'name': 'B', 'id': 3, 'dir': 0 }
```

Incidence Matrix Data

- The file [design]_connectivity.npz contains NumPy arrays that can be used to define SpiPy sparse matrices in coordinate (COO) format
- COO matrices can be quickly constructed and easily converted into other matrices
- Each non-zero entry in the matrix gives the terminal index
- Remember to subtract 1 to get the index for each cell's terminal array
- Note that the matrix will contain duplicate indices if a net connects multiple terminals on one instance

```
>>> import numpy as np
>>> from scipy.sparse import coo_matrix
>>> conn=np.load('counter_connectivity.npz')
>>> coo = coo_matrix((conn['data'], (conn['row'], conn['col'])), shape=conn['shape'])
>>> coo.getrow(6).getcol(3).toarray()
array([[3]])
>>> cells[design['instances'][6]['cell']]['terms'][3-1]
{'name': 'B', 'id': 3, 'dir': 1}
```

Example Incidence Matrix

- Incidence matrix for counter design shown

Terminal
Indices

Nets

```
>>> coo.toarray()
array([[1, 0, 0, 0, 0, 0, 3, 0, 0, 0, 0, 0, 2, 4, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 3, 0, 2, 0, 0, 0, 0, 6, 0, 5, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 2, 3, 0, 6, 0, 0, 0, 5, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 2, 0, 0, 0, 6, 0, 0, 5, 0, 0, 0, 0, 1, 0, 0, 0, 0, 3, 0],
       [0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 3, 0, 0, 0, 0, 0, 0, 0, 2, 0],
       [0, 0, 0, 3, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 2, 0, 0, 0, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 3, 0, 1, 0, 0, 0, 0, 0, 0, 0, 4, 2, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
       [0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 0, 0, 3, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 0, 0, 0, 0, 0, 3, 0, 0, 0, 0, 0, 0, 0, 1, 0],
       [0, 3, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 1, 0, 4, 0, 0, 0, 0, 0, 0, 0, 0],
       [0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 3, 0, 0, 0, 0, 2, 0, 0],
       [0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 0, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0],
       [0, 3, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 4, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 2, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 1],
       [0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0],
       [0, 0, 0, 0, 0, 2, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0],
       [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 2, 0, 0],
       [0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0],
       [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2]])
```

Instances

NumPy Congestion Data

```
import numpy as np
data=np.load('counter_congestion.npz')
lyr=list(data['layerList']).index('M1')
ybl=data['yBoundaryList']
xbl=data['xBoundaryList']
i,j=getGRCIndex(xloc,yloc,xbl,ybl)
data['demand'][lyr][i][j]
data['capacity'][lyr][i][j]
```

- read NumPy file
- get the index for layer M1
- get the array of y-coordinates for GRC left-edge boundaries
- get the array of x-coordinates for GRC bottom-edge boundaries
- get the array indices for the GRC containing location (xloc,yloc)
- get the demand for layer M1 at location (xloc,yloc)
- get the capacity for layer M1 at location (xloc,yloc)

- Routers report congestion information for each Global Route Cell (GRC) using the following values
 - Demand – number of routing tracks needed in the GRC
 - Capacity – number of possible routing tracks in the GRC (this number does not vary with demand)
- GRCs are organized in a rectangular grid, but not all GRCs have the same dimensions.
 - We use the x|yBoundaryList arrays to keep track of the GRC boundaries. This helps convert locations to array indices.
 - All x,y dimensions are in $\frac{1}{2}$ nm units (i.e. divide by 2000 to get microns)

getGRCIndex Function

```
def getGRCIndex(x,y,xbl,ybl):  
    j=0  
    for b in xbl[1:]:  
        if x<b:  
            break  
        j+=1  
    i=0  
    for b in ybl[1:]:  
        if y<b:  
            break  
        i+=1  
    return i,j
```

- Routers report congestion information for each Global Route Cell (GRC).
- The getGRCIndex() function used decode GRC indices from x,y coordinates (shown on the previous slide) can be defined as shown on this slide.
- Note that the **demand** and **capacity** arrays are organized with x-coordinates increasing in rows (*i.e.* last array index) and y-coordinates increasing in columns (*i.e.* next-to-last array index). Therefore, in the reference data['demand'][lyr][i][j], i references the y-location, while j references the x-location.