Investigation of Causal Effects

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6/30/2020

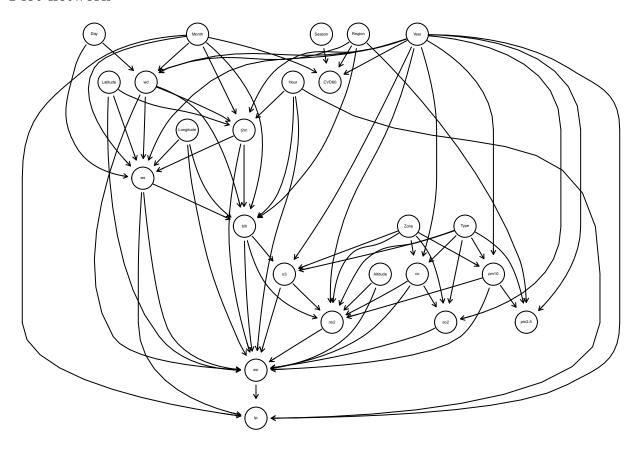
Translation of learned bayesian network to igraph

```
sample.size <- 5e+06
sample <- generate.sample(sample.size)

clustered.data <- sample %>% base::sapply(function(x) if (!is.factor(x)) {
    factor(Ckmedian.1d.dp(x, k = c(3, 5))$cluster)
} else x) %>% base::as.data.frame()

# TODO replace bn with our learned and saved model
adj.matrix.from.bngraph <- amat(bn)
igraph.from.bnmatrix <- graph_from_adjacency_matrix(adj.matrix.from.bngraph)
# TODO mark undirected edges as such</pre>
```

Plot network



Find causal effect

```
effect.igraph.from.bnmatrix <- causal.effect(y = "so2", x = "co",
    z = NULL, G = igraph.from.bnmatrix, expr = T)</pre>
```

Estimate causal effect strength

 $^{`\}sum_{Zone,Type,Year}P(so2|Zone,Type,Year,co)P(Year)P(Type)P(Zone)`$

% latex table generated in R 3.6.2 by x table 1.8-4 package % Tue Jun 30 16:12:54 2020

	so2	co	P(so2 Type, Zone, DateTime, co)P(Type)P(Zone)P(DateTime)	P(co)
1	2	1	0.00	0.08
2	2	2	0.00	0.10
3	2	3	0.00	0.11
4	2	4	0.00	0.67
5	2	5	0.00	0.04
6	3	1	0.00	0.08

causal.effect.size.so2.co(clustered.data)

[1] 5.369082e-11