Investigation of Causal Effects

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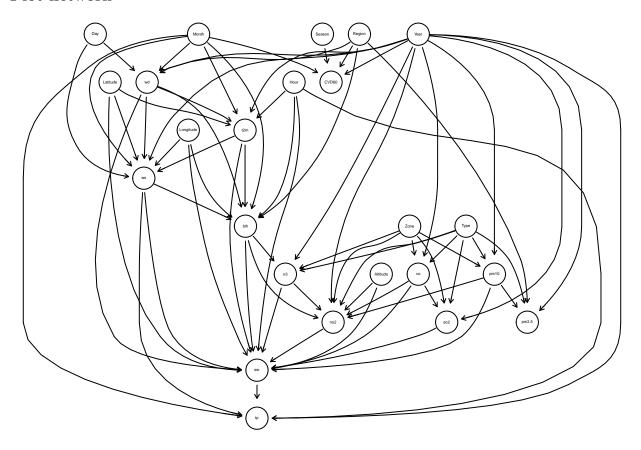
Translation of learned bayesian network to igraph

```
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

```
causal.effect.probabilites.so2.co <- function(data) {

   data %>% select(so2, co, Type, Zone, DateTime) %>% add_count(Zone) %>%

        mutate(prob_zone = n/nrow(data)) %>% select(-n) %>% add_count(Type) %>%

        mutate(prob_type = n/nrow(data)) %>% select(-n) %>% add_count(DateTime) %>%

        mutate(prob_datetime = n/nrow(data)) %>% select(-n) %>%

        group_by(Zone, Type, DateTime, co) %>% add_count(so2) %>%

        mutate(so2_cond_prob = n/nrow(data)) %>% select(-n) %>%

        mutate(total_prob = prob_zone * prob_type * prob_datetime *

            so2_cond_prob) %>% distinct() %>% group_by(so2, co) %>%

        summarise(`P(so2|Type, Zone, DateTime, co)P(Type)P(Zone)P(DateTime)` = sum(total_prob)) %>%

        inner_join(data %>% group_by(co) %>% count(co) %>% mutate(`P(co)` = n/nrow(data)) %>%

        select(-n))
}
```

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

% latex table generated in R 3.6.2 by x table 1.8-4 package % Wed Jul 01 09:57:24 2020

| - | so2 | co | P(so2 Type, Zone, DateTime, co)P(Type)P(Zone)P(DateTime) | P(co) |
|----|-----|----|--|---------|
| 1 | 1 | 1 | 0.0000000005 | 0.07581 |
| 2 | 1 | 2 | 0.0000000008 | 0.10108 |
| 3 | 1 | 3 | 0.0000000007 | 0.10969 |
| 4 | 1 | 4 | 0.0000000900 | 0.67107 |
| 5 | 1 | 5 | 0.0000000017 | 0.04235 |
| 6 | 2 | 1 | 0.0006091325 | 0.07581 |
| 7 | 2 | 2 | 0.0006955428 | 0.10108 |
| 8 | 2 | 3 | 0.0005066126 | 0.10969 |
| 9 | 2 | 4 | 0.0035689610 | 0.67107 |
| 10 | 2 | 5 | 0.0004706730 | 0.04235 |
| 11 | 3 | 1 | 0.0000083758 | 0.07581 |
| 12 | 3 | 2 | 0.0000000277 | 0.10108 |
| 13 | 3 | 3 | 0.0000000062 | 0.10969 |
| 14 | 4 | 1 | 0.0000000127 | 0.07581 |
| 15 | 4 | 2 | 0.0000000816 | 0.10108 |
| 16 | 4 | 3 | 0.0000016898 | 0.10969 |
| 17 | 4 | 4 | 0.0000001014 | 0.67107 |
| 18 | 4 | 5 | 0.0000000214 | 0.04235 |
| 19 | 5 | 2 | 0.0000000117 | 0.10108 |
| 20 | 5 | 3 | 0.0000000281 | 0.10969 |
| 21 | 5 | 4 | 0.0000014091 | 0.67107 |
| 22 | 5 | 5 | 0.0000000405 | 0.04235 |

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

[1] 0.0004713883