

DataBaseHW1

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Part1

1.1

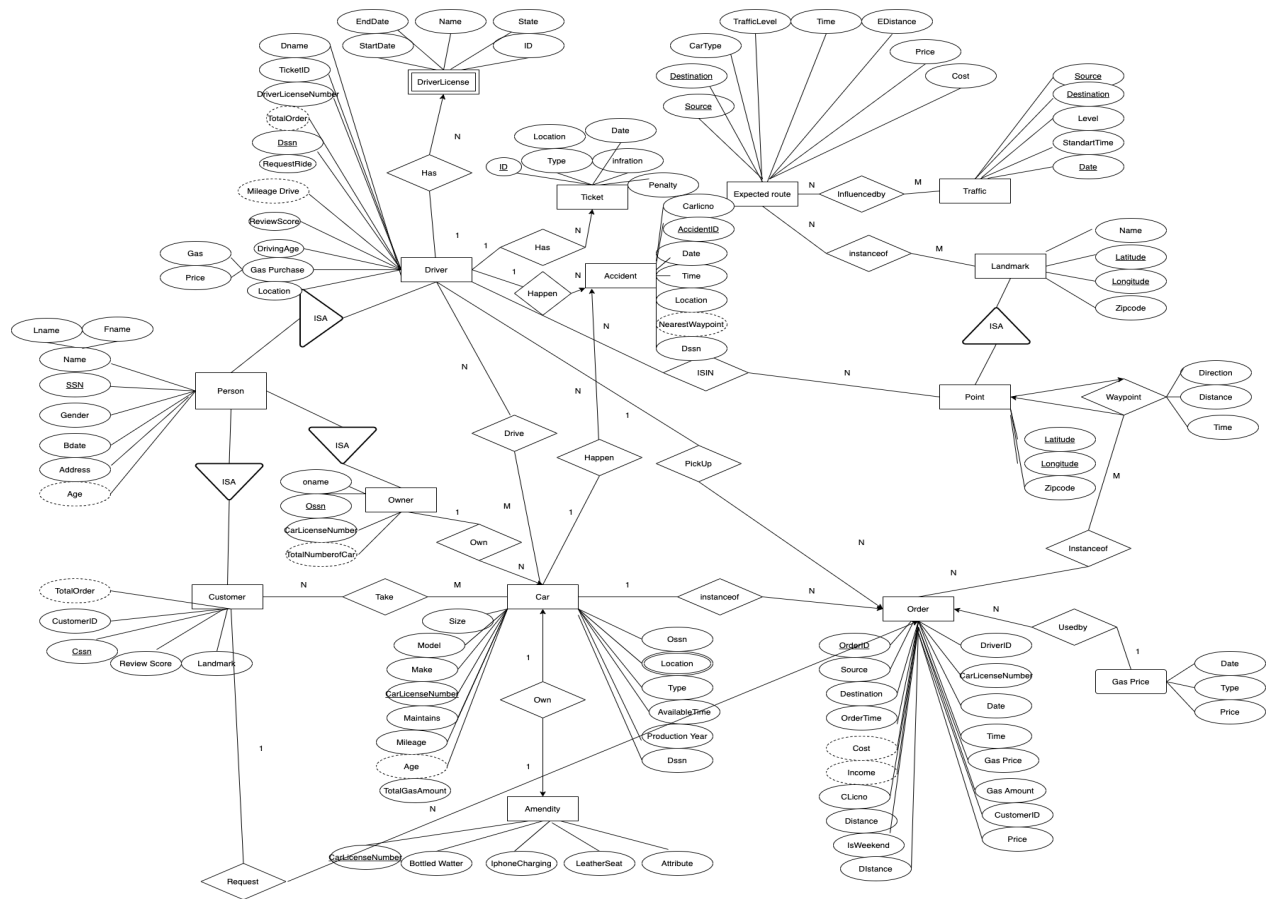


Figure 1: ERModel

Assumption:

1. Driver will fill up gas every morning, and will run out gas every evening, the gas price doesn't change in one day.
2. the cost for a driver is only something about filling up gas.
3. Driver can pick up customer at any points.
4. The traffic level is the same on a day in several weeks (Previous Wednesday = Next Wednesday)

Unusual Controversial design decisions:

I don't record when a driver starts to work and when he stops to work everyday, I just record the time and distance they are in a order. So the avg(distance/time) is not exact. For any Point, it will have 4 directions to other points, which will cause the content of this table is too much.

1.2

Unusual Controversial design decisions:

Using latitude and longitude to identify a point, which is indirect. And we only can see the difference for trips which start from a landmark and end to a landmark. (Cause we only store the information from landmark to landmark in Expected route table.

PERSON	LNAME	FNAME	<u>SSN</u>	GENDER	BDATE	ADDRESS	Name
	KUN	LIU	123456789	MALE	7/9/96	100 WEST	KUN LIU

Figure 2: Person

DRIVER	<u>DSSN</u>	DLICNO	TICKETID	REQUESTRIDE	REVIEWSCORE	DRIVINGAGE	LOCATION	Dname	ID
	123456789	5656789	1	20	90	10	41°24'12.2"N 2°10'26.5"E	Kun Liu	2008

Figure 3: Driver

CUSTOMER	<u>CSSN</u>	REVIEWSCORE	LANDMARK	CUSTOMERID
	111111111	95	Homewood	1111111

Figure 4: Customer

OWNER	<u>OSSN</u>	CLICNO	ONAME
	123456789	443443MD	Kun Liu

Figure 5: Owner

DRIVERLICENSE	NAME	ID	STATE	STARTDATE	ENDDATE
	KUN LIU	5656789	MD	8/9/10	8/9/21

Figure 6: DriverLicense

CAR	CLICNO	DSSN	SIZE	MODEL	MAKE	MAINTAINTime	MILEAGE	LOCATION	TYPE	AVAILABETIME	PRODUCTION YEAR	DSSN	TOTALGASAMOUNT
	443443MD	123456789	large	Camry	Toyota	8/9/20	100000km	41°24'12.2"N 2°10'26.5"E	Electronic car	10:20Am		2017	123456789 1000L

Figure 7: Car

AMENDITY	BOTTLEDWA	IPHONECHA	LEATHERSEA	ATTRIBUTE	CLICNO
	1	1	0	1	443443MD

Figure 8: Amenity

TICKET	ID	TYPE	INFRATION	PENALTY	Location	Date
	1	Red Light	1	500\$	41°24'12.2"N	8/9/20

Figure 9: Ticket

ACCIDENT	DSSN	LOCATION	TIME	DATE	ACCIDENTID	CARLICNO	NearestWayPoint
	123456789	41°24'12.2"N 2°10'26.5"E	10:20am	9/8/19	123456	443443MD	41°24'12.2"N 2°10'26.5"E

Figure 10: Accident

EXPTected ROUTE	SOURCE	DESTINATION	CARTYPE	TRAFFICLEVEL	TIME	EDISTANCE	PRICE	COST
	JHU Hospital	JHU Homewood	Electronic car	Low	20mins	2miles	20\$	15\$

Figure 11: Expected Route

TRAFFIC	SOURCE	DESTINATION	LEVEL	STANDARTTIM	DATE
	JHU Hospital	JHU Homewood	Low	20mins	Wednesday

Figure 12: Traffic

LANDMARK	NAME	LATITUDE	LONGITUDE	ZIPCODE
	JHU Homewood	39.3299° N	76.6205° W	21218

Figure 13: Landmark

POINT	LATITUDE	LONGITUDE	ZIPCODE
	38.2222N	77.6241W	21210

Figure 14: Point

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WAYPOINT	DIRECTION	DISTANCE	TIME	SOURCE	DESTINATION
	south	2miles	20mins	JHU Hospital	JHU Homewood

Figure 15: WayPoint

ORDER	ORDERID	DRIVEID	SOURCE	DESTINATION	ORDERTIME	CLICNO	DATE	TIME	GASPRICE	GASAMOUNT	CUSTOMERID	ISWEEKEND	PRICE	Distance
	1		2008 JHU Hospital	JHU Homewood	10:20:00 AM	443443MD		9/8/20 21mins	10\$/L		1.5	111111111	0 255	2miles

Figure 16: Order

GAS	DATE	TYPE	PRICE
	9/8/20	No.95	10\$/L

Figure 17: Gas

c:

$BothDC \leftarrow (Driver \bowtie_{dssn=cssn} Customer)$
 $DriveEC \leftarrow \Pi_{dssn}(\sigma_{type="ElectronicCar"}(Driver \bowtie_{clino=clino} Car))$
 $BothDCEC \leftarrow (BothDC \bowtie DriveEC)$
 $BothDCNotEC \leftarrow (BothDC - BothDCEC)$
 $Result \leftarrow \Pi_{name}(Person \bowtie_{ssn=dssn} BothDCNotEC)$

f:

$CarInAO \leftarrow (Accident \bowtie_{clino=clino} Car \bowtie_{clino=clino} Owner)$
 $CarInAD \leftarrow (Accident \bowtie_{clino=clino} Car \bowtie_{clino=clino} Driver)$
 $CarInAOwnerName \leftarrow (CarInAO \bowtie_{ossn=ssn} Person)$
 $CarInADriverName \leftarrow (CarInAD \bowtie_{dssn=ssn} Person)$
 $Result \leftarrow \Pi_{date,time,nearestwayPoint,oname,dname}(CarInAOwnerName)$

h:

$CarDO \leftarrow \sigma_{dssn=ossn}(Car)$
 $CarDnotO \leftarrow (Car - CarDO)$
 $Result \leftarrow \Pi_{make,model,year,mileage,dname,oname}(Driver \bowtie_{clino=clino} Car \bowtie_{clino=clino} Owner)$

y:

$OrderAmoreE \leftarrow \sigma_{date="9/12/20" \wedge distance > edistance}(Order \bowtie_{source=source \wedge destination=destination} ExpecteRoute)$
 $Result \leftarrow \Pi_{source,destination,dname}(OrderAmoreE \bowtie_{driveid=id} Driver)$

ss:

$DriveVisitedSLM \leftarrow \Pi_{driverid,source(lm)}(Order)$
 $DriverVisitedDLM \leftarrow \Pi_{driverid,destination(lm)}(Order)$
 $DriverDSL M \leftarrow (DriverVisitedSLM \cup DriverVisitedDLM)$
 $LMName \leftarrow \sigma_{name}(LandMark)$

$Result \leftarrow (Driver \bowtie_{id=driverid} (DriverDSLM \div LMName))$

ccc:

$Result \leftarrow \Pi_{make,model,productionyear}(\sigma_{Max(mileage \div totalgasamount)}(Car))$

Part2

Part2 Relation Algebra

2.1

$ITrump \leftarrow \sigma_{DName="Ivanka Trump"}(Drinker)$
 $ITBarDate \leftarrow \Pi_{bno,dataofvisit}(ITrump \bowtie_{dlicno=dlicno} Visit)$
 $PDLicNo \leftarrow \Pi_{dlicno}(ITBarDate \bowtie Visit)$
 $Result \leftarrow \Pi_{dname,politicalparty}(PDLicNo \bowtie_{dlicno=dlicno} Drinker)$

2.2

$BarInMD \leftarrow \sigma_{Bstate="MD"}(Bar)$
 $BarInBal \leftarrow \sigma_{Bcity="Baltimore"}(Bar)$
 $BarInMDnotBal \leftarrow BarInMD - BarInBal$
 $BarInMDnotBalServeBudLite \leftarrow (Bar \bowtie_{bno=bno} (\sigma_{beername="BudLite"}(Serves)))$
 $BarInMDnotBalNotServeBudLite \leftarrow BarInMDnotBal - BarInMDnotBalServeBudLite$
 $Result \leftarrow \Pi_{barname}(BarInMDnotBalNotServerBudLite \bowtie Bar)$

2.3

$BarInGeorgetown \leftarrow \sigma_{Bcity="Gergetown"}(Bar)$
 $PUnder30VGT \leftarrow \Pi_{dlicno}(BarInGeorgetown \bowtie_{bno=bno} Visit \bowtie_{dlicno=dlicno} (\sigma_{age < 30} Drinker))$
 $PLikeBL \leftarrow \sigma_{beername="BudLite"}(Likes)$
 $PLikeML \leftarrow \sigma_{beername="MillerLite"}(Likes)$
 $PDontLikeML = Likes - PLikeML$
 $PLikeBLDontLikeML = PLikeBL \cap PDontLikeML$
 $Result \leftarrow \Pi_{dname}(PUnder30VGT \bowtie PLikeBLDontLikeML)$

2.4

$DTrump \leftarrow \sigma_{dname="DonaldTrump"}(Drinker)$
 $DTBar \leftarrow \Pi_{bno}(DTrump \bowtie_{dlicno=dlicno} Visit)$
 $Result \leftarrow \Pi_{dname,age}(Drinker \bowtie_{dlicno=dlicno} Visit \bowtie_{bno=bno} DTBar)$

2.5

$AllBarinTowson \leftarrow \sigma_{bcity="Towson"}(Bar)$
 $PVisitBarinTowson \leftarrow (AllBarinTowson \bowtie_{bno=bno} Visit \bowtie_{dlicno=dlicno} Drinker)$
 $Result \leftarrow \Pi_{name,age}(PVisitBarinTowson \div AliiBarinTowson)$

2.6

$PVisitBar \leftarrow \Pi_{dlicno}(Visit)$
 $PSufferCOVID \leftarrow \Pi_{dlicno}Covid_diagnosis$
 $PVisitBarOrSufferCOVID \leftarrow (PVisitBar \cup PSufferCOVID)$
 $Result \leftarrow \Pi_{name,birthday}(Drinker - PVisitBarOrSufferCOVID)$

2.7

$DTrump \leftarrow \sigma_{dname="DonaldTrump"}(Drinker)$
 $DTBar \leftarrow \Pi_{bno}(DTrump \bowtie_{dlicno=dlicno} Visit)$
 $PVDTBar \leftarrow (DTBar \bowtie Visit)$
 $PVEveryDTBar \leftarrow (PVDTBar \div DTBar)$
 $JBiden \leftarrow \Pi_{dlicno}(\sigma_{dname="JoeBiden"}(Drinker))$
 $JBBar \leftarrow \Pi_{bname}(JBiden \bowtie_{dlicno=dlicno} Visit)$
 $PVisitJBBar \leftarrow (Visit \bowtie JBBar)$
 $PDontVisitJBBar \leftarrow (Visit - PVisitJBBar)$
 $Result \leftarrow \Pi_{name,age}(PVEveryDTBar \bowtie_{dlicno=dlicno} PDontVisitJBBar \bowtie Drinker)$

2.8

$PVisitBar \leftarrow \Pi_{dlicno}(Visit)$
 $PDrunkSA \leftarrow \Pi_{dlicno}(\sigma_{beername="SamAdams"}Beer_purchase)$
 $Result \leftarrow \Pi_{name}(Drinker \bowtie_{dlicno=dlicno} (PVisitBar - PDrunkSA))$

2.9

$DTrump \leftarrow \sigma_{dname="DonaldTrump"}(Drinker)$
 $DTLikesBeer \leftarrow \Pi_{beername}(DTrump \bowtie_{dlicno=dlicno} Likes)$
 $BarServeDTLikes \leftarrow (DTLikesBeer \bowtie_{beername=beername} Serves \bowtie Bar)$
 $Result \leftarrow \Pi_{bname}((Bar \bowtie Serves) - BarServeDTLikes)$

2.10

$BarinTimonium \leftarrow \sigma_{bcity="Timonium"}(Bar)$
 $BeerServeinTimonium \leftarrow \Pi_{beername}(BarinTimonium \bowtie_{bno=bno} Serves)$
 $BarInTowson \leftarrow \sigma_{bcity="Towson"}(Bar)$
 $BarInTowsonServeBeer \leftarrow (BarinTowson \bowtie_{bno=bon} Serves)$
 $BarInTowsonServeTheseBeer \leftarrow (BarinTowsonServeBeer \bowtie_{beername=beername} BeerServeinTimonium)$
 $Result \leftarrow \Pi_{bname}((Bar \bowtie Serves) - BarInTowsonServeThesesBeer)$

2.11

$DTrump \leftarrow \sigma_{dname="DonaldTrump"}(Drinker)$
 $DTLikesBeer \leftarrow \Pi_{beername}(DTrump \bowtie_{dlicno=dlicno} Likes)$
 $ITrump \leftarrow \sigma_{dname="IvankaTrump"}(Drinker)$
 $ITLikesBeer \leftarrow \Pi_{beername}(ITrump \bowtie_{dlicno=dlicno} Likes)$
 $BarServeDTLikeBeer \leftarrow \Pi_{bno}(DTLikesBeer \bowtie_{beername=beername} Serves)$
 $BarServeITLikeBeer \leftarrow \Pi_{bno}(ITLikesBeer \bowtie_{beername=beername} Serves)$
 $BarBothDTAndITLike \leftarrow \Pi_{bno}(BarServeDTLikeBeer \cap BarServeITLikeBeer)$
 $Result \leftarrow \Pi_{name}(BarBothDTANDITLike \bowtie_{bno=bno} Bar)$

2.12

$Num_serves \leftarrow \sigma_{Num_serves(bno,Numbeer)}Bno\mathcal{G}COUNT(beername)Serves$
 $Max_num_serves \leftarrow \sigma_{Max_num_serves(MaxNum)}\mathcal{G}MAX(Numbeer)Num_serves$
 $Max_bno \leftarrow \Pi_{bno}(Num_serves \bowtie_{Numbeer=MaxNum} Max_num_serves)$
 $Result \leftarrow \Pi_{barname,bcity,bstate}(Bar \bowtie_{bno=bno} Max_bno)$

2.13

$\text{Num_likes} \leftarrow \sigma_{\text{Num_likes}(\text{dlicno}, \text{Number})} \text{dlicno} \mathcal{G} \text{COUNT}(\text{beername}) \text{Likes}$
 $\text{Min_num_dlicno} \leftarrow \sigma_{\text{Min_num_dlicno}(\text{MinNum})} \mathcal{G} \text{MIN}(\text{Number}) \text{Num_Likes}$
 $\text{Min_dlicno} \leftarrow \Pi_{\text{dlicno}}(\text{Num_likes} \bowtie_{\text{Number}=\text{MinNum}} \text{Min_num_dlicno})$
 $\text{Result} \leftarrow \Pi_{\text{dname,age}}(\text{Drinker} \bowtie_{\text{dlicno}=\text{Min_dlicno}} \text{Min_dlicno})$

2.14

$\text{DTrump} \leftarrow \sigma_{\text{dname}=\text{"DonaldTrump"}}(\text{Drinker})$
 $\text{MPence} \leftarrow \sigma_{\text{dname}=\text{"MikePence"}}(\text{Drinker})$
 $\text{DTLikesBeer} \leftarrow \Pi_{\text{beername}}(\text{DTrump} \bowtie_{\text{dlicno}=\text{dlicno}} \text{Likes})$
 $\text{MPLikesBeer} \leftarrow \Pi_{\text{beername}}(\text{MPence} \bowtie_{\text{dlicno}=\text{dlicno}} \text{Likes})$
 $\text{DTLikeOrMKLike} \leftarrow (\text{DTLikesBeer} \cup \text{MPLikesBeer})$
 $\text{DrinkerLikeDTORMK} \leftarrow (\text{Likes} \bowtie_{\text{beername}=\text{beername}} \text{DTLikeOrMKLike})$
 $\text{Result} \leftarrow \Pi_{\text{name,politicalParty}}((\text{Drinker} \bowtie \text{Likes}) - (\text{DrinkerLikeDTORMK} \bowtie \text{Drinker}))$

2.15

$\text{DTrump} \leftarrow \sigma_{\text{dname}=\text{"DonaldTrump"}}(\text{Drinker})$
 $\text{AllDTLikesBeer} \leftarrow \Pi_{\text{beername}}(\text{DTrump} \bowtie_{\text{dlicno}=\text{dlicno}} \text{Likes})$
 $\text{DrinkerLikeAllDTBeer} \leftarrow (\text{Likes} \div \text{AllDTLikesBeer}) \text{Result} \leftarrow \Pi_{\text{name,politicalparty}}(\text{DrinkerLikeAllDTBeer})$

2.16

$\text{Result} \leftarrow \Pi_{\text{beername}}(\sigma_{\text{dateofpurchase} < \text{estimatedEnddata} \wedge \text{dataofpurchase} > \text{estimatedstartDate}}(\text{Beer_purchase} \bowtie \text{Covid_diagnosis}))$

2.17

$\text{DTrump} \leftarrow \sigma_{\text{dname}=\text{"DonaldTrump"}}(\text{Drinker})$
 $\text{DTVisitBar} \leftarrow (\text{DTrump} \bowtie \text{Visit})$
 $\text{DTVisitBarMoreThan1} \leftarrow \sigma_{\text{COUNT}(\text{BNO}) > 1}(\text{DTVisitBar})$
 $\text{Result} \leftarrow \Pi_{\text{barname}}(\text{DTVisitBarMoreThan1} \bowtie_{\text{bno}=\text{bno}} \text{Bar})$

2.18

$ITrump \leftarrow \sigma_{dname="IvankaTrump"}(Drinker)$
 $ITVisit \leftarrow (ITrump \bowtie Visit)$
 $ITVCovid \leftarrow (\sigma_{dateofVisit < estimatedEnddata \wedge dataofvisit > estimatedstartDate}(ITVisit \bowtie Covid_diagnosis))$
 $DrinkerVisitSameITVCovid \leftarrow (ITVCovid \bowtie_{bno=bno \wedge dateofVisit=dateofVisit} Visit)$
 $Result \leftarrow \Pi_{name, phonenumner}(DrinkerVisitSameITVCovid \bowtie Drinker)$

2.19

$VisitDuringAllBD \leftarrow (Visit \bowtie AllBnoDateofVisit)$
 $Result \leftarrow \Pi_{name, phonenumner}(VisitDuringAllBD \bowtie Drinker)$

2.20

$ALLVisitCovid \leftarrow (\sigma_{dateofVisit < estimatedEnddata \wedge dataofvisit > estimatedstartDate}(Visit \bowtie Covid_diagnosis))$
 $Result \leftarrow \Pi_{dname, barname, estimatedStartDate, estimatedEndDate}(ALLVisitCovid \bowtie_{dlicno=dlicno} Drinker \bowtie_{bno=bno} Bar)$

2.21

$JBiden \leftarrow \sigma_{dname="JoeBiden"}(Drinker)$
 $JBBar \leftarrow \Pi_{bno}(JBiden \bowtie_{dlicno=dlicno} Visit)$
 $NumBar_JB_visit \leftarrow \sigma_{NumBar_JB_visit(bno, Numvisit)} bno \mathcal{G}COUNT(bno) JBBar$
 $Num_drinker_visit \leftarrow \sigma_{Num_drinker_visit(dlicno, bno, Numvisit)} dlicno \mathcal{G}COUNT(bno) Visit$
 $DSameBarJB \leftarrow \Pi_{dlicno}(Num_drinker_vist \div NumBar_JB_visit)$
 $Result \leftarrow \Pi_{dname, age}(Drinker \bowtie_{dlicno=dlicno} DSameBarJB)$

2.22

$AllDrinkerVaccinated \leftarrow Drinker \bowtie Covid_Vaccine$
 $AllDVDiagnosis \leftarrow AllDrinkerVaccinated \bowtie Covid_Diagnosis$
 $ALLDrinkerVSameDate \leftarrow \sigma_{dateofvaccine=dateofdiagnosis}(AllDrinkerVaccinated)$
 $ALLDrinkerVaccinatedDifferentDate \leftarrow (AllDrinkerVaccinated - AllDrinkerVSameDate)$
 $Result \leftarrow \Pi_{name, age, politicalparty}(Drinker \bowtie_{dlicno=dlicno} AllDrinkerVaccinatedDifferentDate)$

2.23

$\text{BeerServedinBatBar} \leftarrow \Pi_{\text{beername}}(\sigma_{\text{barname}=\text{"Batbar"}}(\text{Bar} \bowtie \text{Serves}))$
 $\text{DrinkerPurchasedBeerSBatBar} \leftarrow (\text{BeerServedinBatBar} \bowtie_{\text{beername}=\text{beername}} \text{Beer_purchased})$
 $\text{DrinkerPurchasedAllBeerinBatbar} \leftarrow (\text{DrinkerPurchasedBeerSBatBar} \div \text{BeerServedinBatBar})$
 $\text{Result} \leftarrow \Pi_{\text{name,age}}(\text{DrinkerPurchasedAllBeerinBatbar} \bowtie_{\text{dlicno}=\text{dlicno}} \text{Drinker})$

Relation Calculus

2.1

$\{t \mid \exists d, i \in \text{Drinker}(\$
 $\exists v \in \text{Visit}(\$
 $i[\text{dname}] = \text{"IvankaTrump"}\$
 $\wedge i[\text{dlicno}] = v[\text{dlicno}]\$
 $\text{landd}[\text{dateofvisit}] = v[\text{dateofvisit}]\$
 $\wedge d[\text{dlicno}] == v[\text{dlicno}]\$
 $\wedge \wedge t[\text{dname}] = d[\text{dname}]\$
 $\wedge t[\text{politicalparty}] = d[\text{politicalparty}]))\}$

2.2

$\{t \mid \exists b \in \text{Bar}(\$
 $\exists s \in \text{Serves}(\$
 $b[\text{bstate}] = \text{"MD"}\$
 $\wedge \neg b[\text{bcity}] = \text{"Baltimore"}\$
 $\wedge \neg s[\text{beername}] = \text{"BudLite"}\$
 $\wedge b[\text{bno}] = s[\text{bno}]\$
 $\wedge t[\text{barname}] = b[\text{barname}]))\}$

2.3

$\{t \mid \exists b \in \text{Bar}(\$
 $\exists d \in \text{Drinker}(\$
 $\exists v \in \text{Visit}(\$
 $\exists l \in \text{Like}(\$
 $b[\text{bcity}] = \text{"Georgetown"}\$
 $\wedge d[\text{age}] < 30\$
 $\wedge \neg l[\text{beername}] = \text{"BudLite"}\$

$$\begin{aligned}
& \wedge \neg l[beername] = \text{"MillerLite"} \\
& \wedge b[bno] = v[bno] \\
& d[dlicno] = v[dlicno] \\
& d[dlicno] = l[dlicno] \\
& \wedge t[dname] = d[dname])])])\}
\end{aligned}$$

2.4

$$\begin{aligned}
& \{t \mid \exists d, dt \in \text{Drinker}(\\
& \exists v, pv \in \text{Visit}(\\
& dt[dname] = \text{"DonaldTrump"} \\
& \wedge dt[dlicno] = v[dlicno] \\
& \wedge d[dlicno] = pv[dlicno] \\
& \wedge pv[bno] = v[bno] \\
& \wedge t[dname] = d[dname] \\
& \wedge t[age] = d[age])])\}
\end{aligned}$$

2.9

$$\begin{aligned}
& \{t \mid \exists b \in \text{Bar}(\\
& \exists s \in \text{Serves}(\\
& \nexists dt \in \text{Drinker}(\\
& \exists dl \in \text{Like} \\
& dt[dname] = \text{"DonaldTrump"} \\
& \wedge dl[dlicno] = dt[dlicno] \\
& \wedge b[bno] = s[bno] \\
& \wedge s[beername] = dl[beername] \\
& \wedge t[barname] = b[barname])])])\}
\end{aligned}$$

2.10

$$\begin{aligned}
& \{t \mid \exists tw, tm \in \text{Bar}(\\
& \exists tws \in \text{Servers}(\forall tms \in \text{Serves}(\\
& tw[bcity] = \text{"Towson"} \\
& \wedge tm[bcity] = \text{"Timonium"} \\
& \wedge \neg tws[beername] = tms[beername] \\
& \wedge tw[bno] = tws[bno] \\
& \wedge tm[bno] = tms[bno] \\
& t[barname] = tm[barname])])\}
\end{aligned}$$